

https://theses.gla.ac.uk/

Theses Digitisation:

https://www.gla.ac.uk/myglasgow/research/enlighten/theses/digitisation/

This is a digitised version of the original print thesis.

Copyright and moral rights for this work are retained by the author

A copy can be downloaded for personal non-commercial research or study, without prior permission or charge

This work cannot be reproduced or quoted extensively from without first obtaining permission in writing from the author

The content must not be changed in any way or sold commercially in any format or medium without the formal permission of the author

When referring to this work, full bibliographic details including the author, title, awarding institution and date of the thesis must be given

Enlighten: Theses <u>https://theses.gla.ac.uk/</u> research-enlighten@glasgow.ac.uk

ENERGY EVALUATION OF GRASS SILAGE

A Thesis Submitted to the University of Glasgow for the Degree of Doctor of Philosophy in the Faculty of Science

bу

MANSOUR S.F. KRIDIS

(B.Sc. King Saud University, Riyadh, Saudi Arabia) (M.Sc. University of California, Davis, U.S.A.)

August 1989

. .

* -

....

The West of Scotland College Nutrition and Microbiology Department Animal Nutrition Unit Auchincruive Ayr, Scotland UK ProQuest Number: 10646918

All rights reserved

INFORMATION TO ALL USERS The quality of this reproduction is dependent upon the quality of the copy submitted.

In the unlikely event that the author did not send a complete manuscript and there are missing pages, these will be noted. Also, if material had to be removed, a note will indicate the deletion.



ProQuest 10646918

Published by ProQuest LLC (2017). Copyright of the Dissertation is held by the Author.

All rights reserved. This work is protected against unauthorized copying under Title 17, United States Code Microform Edition © ProQuest LLC.

> ProQuest LLC. 789 East Eisenhower Parkway P.O. Box 1346 Ann Arbor, MI 48106 – 1346



698215

.

Mesis 8218 0042



.

•

.

a provincia de provincia de la manda de provincia de la manda Receletaria de la manda de l

.

د

TO MY PARENTS, MY WIFE: ABEER,

MY SONS: ABDULLAH AND MOHAMMAD, MY

BROTHERS AND SISTERS FOR ALL THEIR LOVE

AND UNDERSTANDING WHICH MADE THIS

THESIS POSSIBLE.

TABLE OF CONTENTS

ACKNOWI	LEDGEMENTS	Page I
LIST OF	TABLES	III
LIST OF	FIGURES	VII
ABBREVI	ATIONS	Х
SUMMARY	ſ	XII
GENERAL	. INTRODUCTION	1
CHAPTER Section	R ONE – REVIEW OF LITERATURE 1 – Energy Evaluation	
1.1.1	Introduction	11
1.1.2	Historic Developments of Forage Evaluation	
1 1 7	in the UK	ΤΤ
1.1.3	ME as a measure of Nutritive value of	11
1.1.4	The Dilemma of Estimating ME for Ration	т.д
	Formulation	15
1.1.5	The Use of Digestibility to Calculate ME	34
1.1.6	Summary	35
Section	2 – Predictions of the Organic Matter	
Digestib	ility of Grass Silage	
1.2.1	Introduction	38
1.2.2	Definitions	38
1.2.3	The Measurement of Digestibility in vivo	39
1.2.4	The Importance of Digestibility as a Useful Index of Nutritive Value	11
1 7 5	Huex of Nutritive value Reatena Affosting the Dispatibility	41
1.2.0	Measurements	11
126	Laboratory Methods for Predicting the Organic	41
1.2.0	Matter Digestibility of Grass Silages	46
1.2.6.1	Introduction	46
1.2.6.2	Criteria for Selection of Laboratory Methods	47
1.2.6.3	Review of the Methods	47
	1.2.6.3.1 Chemical Methods	48
	1.2.6.3.2 Biological Methods	55
	1.2.6.3.3 A Physical Method (NIR)	69
1.2.7	Summary	77
CHAPTE	R TWO - MATERIALS AND METHODS	
2.1	Silages Studied	81
2.2	Analytical Methods	81
2.2.1	Dry Matter Base	81
2.2.2	Oven Dry Matter Determination	88
2.2.3	Asn Determination	88
4.4.4 2.2.5	Mouilled Acid Detergent Fibre	88
2.2.0	Pensin-Collulase OMD Determination	09
2.2.0	In vitro OMD Determination	90 Q1
2.2.8	In situ OMD Determination	93
2.2.8.1	The Preparation of Hay Samples	93
2.2.8.2	The Preparation of Silage Samples	93
2.2.8.3	Animals and Their Diets	93
2.2.8.4	Polyester Bag Technique	93
2.2.8.5	Analytical Methods of the Bags Residue	
	after Incubation	96

,

```
Page
```

Washing the Bags	96
Total Nitrogen Determination of Hay	00
Samples and Residues after Incubation	90
Near Infrared Reflectance Spectroscopy (NIRS)	98
Mathematical Transformations	99
Equation Output	101
Neutral Detergent Cellulase OMD Determination	103
Laboratory Methods Comparison	104
Statistical Analysis	108
	Washing the Bags Total Nitrogen Determination of Hay Samples and Residues after Incubation Near Infrared Reflectance Spectroscopy (NIRS) Mathematical Transformations Equation Output Neutral Detergent Cellulase OMD Determination Laboratory Methods Comparison Statistical Analysis

.

CHAPTER THREE - THE EFFECT OF DIFFERENT WASHING PROCEDURES ON THE LOSSES OF OM AND N FROM SAMPLES OF HAY INCUBATED IN POLYESTER BAGS WITHIN THE RUMEN OF SHEEP

3.1	Introduction	113
3.2	Methods	113
3.2.1	Treatments and Design	113
3.2.2	Zero Time Determination	114
3.2.3	Multiple Washing Experiment	114
3.2.4	Fitting the Mathematical Model	117
3.3	Results	117
3.3.1	Organic Matter Disappearance from Bags	117
3.3.2	Nitrogen Disappearance from Bags	124
3.3.3	Multiple Washing Experiment	128
3.4	Discussion	136
3.4.1	Organic Matter Disappearance from Bags	136
3.4.2	Nitrogen Disappearance from Bags	138
3.5	Conclusion	139

CHAPTER FOUR - PREDICTION OF ORGANIC MATTER DIGESTIBILITY OF GRASS SILAGES

.

4.1	Introduction	142
4.2	Results	142
4.2.1	Chemical Methods	142
4.2.2	Biological Methods	146
4.2.3	The Use of Bivariate Relationship to	
	Predict in vivo OMD	152
4.2.4	A Physical Method (NIR)	155
4.2.5	Validation of the Calibration Equations	
	Produced by the Traditional Methods and NIR	161
4.2.6	Validation of NIR 8-term Equation with	
	Irish Silages	166
4.2.7	Calibration of NIR with IVOMD	166
4.3	Discussion	172
4.3.1	Chemical Methods	172
4.3.2	Biological Methods	175
4.3.3	The Use of Bivariate Relationship to	
	Predict in vivo OMD	180
4.3.4	A Physical Method (NIR)	181
4.3.5	Calibration of NIR with IVOMD	184
4.4	Comparison of Methods	185
4.5	Conclusions	191

Page

.

CHAPTER	FIVE - THE EFFECT OF EXTERNAL FACTORS	
IN THE	RELATIONSHIP BETWEEN IN VIVO OMD AND	
ITS PRE	DICTORS	
5.1	Introduction	194
5.2	Methods	194
5.3	Results	196
5.4	Discussion	196
5.5	Conclusions	201
CHAPTER	SIX - PREDICTION OF <u>IN VIVO</u> DOMD AND DE	
Section '	1 - Prediction of DOMD of Grass Silance	
6.1.1	Introduction	203
6 1 2	Methods	204
6.1.3	Statistical Analysis	204
6.1.4	Results	209
6.1.5	Discussion	209
6.1.6	Conclusions	212
Section	2 – Prediction of the DE of Grass Silages	
6.2.1	Introduction	213
6.2.2	Methods	213
6.2.3	Statistical Analysis	215
6.2.4	Results	215
6.2.5	Discussion	218
6.2.6	Conclusions	219
CHAPTER	R SEVEN - GENERAL DISCUSSION	222
REFEREN	ICES	237
PUBLICA	TIONS	261
APPENDI	CES	263
		-

.

.

ACKNOWLEDGEMENTS

The author wishes to praise the Almightly Allah for making him able to accomplish this study.

This research was made possible by the co-operation and support of many people and organisations.

- 1 The author wishes to express his appreciation to the King Saud University, Government of The Kingdom of Saudi Arabia, without whom this study would not have been possible.
- 2 Verv special thanks and deep regards go to mv supervisor, Dr Geoffrey Barber. His devotion to this work, honest thoughts his and supervision, his encouragement and his affectionate and hospitable family were among the greatest inspirations during my stay in Scotland. The uncountable hours of informal teaching and discussions and the constructive criticism during the preparation of this thesis were among the greatest learning experiences I have ever had.

Very sincere thanks to Dr N W Offer for his help, valuable guidance and informal teaching given throughout the course of this work.

3 Thanks are also due to Mr D I Givens for his friendship, help and valuable support for this work. His provision of ADAS FEU silages is greatly appreciated.

Many thanks also go to Dr I Murray for his friendship and valuable help in NIR work.

I would like also to express my sincere thanks to Dr J H Topps, North of Scotland College of Agriculture, Mr P J S Dewey and R M Livingstone, Rowett Research Institute, and Dr E F Unsworth, Department of Agriculture for Northern Ireland, for providing silage samples and data.

Ι

- 5 I wish to thank Mr E A Hunter, Scottish Agricultural Statistics Service for his statistical advice and his interest in the work. Thanks also to Mr D Arnot for his great help in statistical analysis and computer work.
- 6 My special thanks to Dr J Dixon, head of the Analytical Chemistry Unit for his help with routine feed evaluations, Mr J Wyllie and Miss I Yuill, from the Metabolism Unit, and Mrs Alison Mann for their great help in nylon bag work.
- 7 I would like also to acknowledge the staff of the laboratories listed in Table 2.8 for the analysis of samples.
- 8 Faculty and staff members of the College are greatly acknowledged for their help, support and friendship.
- 9 Many thanks go to Miss Lorraine Maley for typing this thesis.
- 10 Finally, sincere and special thanks and regards to my wife, Abeer, for her encouragement, support and outstanding responsibility during my stay in Scotland and to my two lovely sons, Abdullah and Mohammad, for giving me the strength and courage to finish this work.

Π

.

Table		Page
1	Production of Silage and Hay in Europe: Estimates for the Countries of the European Economic Community and Scandinavia (1978–1979).	3
1.1	Methods Used to Predict ME by UK Laboratories.	17
1.2	DOMD and ME Prediction Equations for Silages.	19
1.3	Gross Energy Contents of Grass Silages.	22
1.4	The Partial Efficiency of Utilisation of Metabolisable Energy for Maintenance (k _m) Determined Calorimetrically and Calculated by the Equation of Agricultural Research Council (1980) for Eight Silages.	30
1.5	The Partial Efficiency of Utilisation of Metabolisable Energy for Fattening (k _f) Determined Calorimetrically and Calculated by the Equation of Agricultural Research Council (1980) for Fourteen Silages.	31
1.6	Efficiency of Utilisation of Metabolisable Energy of Silage Diets for Lactation.	33
1.7	Approximate Relative Variation Due to Forages and Animals.	42
1.8	Division of Forage Organic Matter by System of Analysis Using Detergents.	50
1.9	Basic Scheme of Forage Analysis Using Detergents.	51
2.1	Sources of Silages	82
2.2	Distribution of <u>in vivo</u> OMD and Chemical and Biological Predictors of the Populations Studied (%).	83
2.3	Composition of Silages for All Populations -	85
2.4	Characterisation of the Silages Studied.	86
2.5	Description of <u>in vivo</u> Trials Performed on Silages Used in this Study.	87
2.6	Composition of the Diet Fed to the Sheep.	94

Table		Page
2.7	An Example of 5 Term Equation Produced by Stepwise Multiple Linear Regression.	102
2.8	Analytical Methods Performed by Contributing Laboratories	106
2.9	Calibration and Validation Samples Used in This Study with Their <u>in vivo</u> OMD Distribution.	107
3.1	Treatments and Design.	115
3.2	Zero Time Determination.	116
3.3	Organic Matter Disappearance (%) of Hay from Polyester Bags Incubated in the Rumen of Sheep and Treated with Different Washing Techniques. Mean of 4 Sheep, one Observation per Sheep.	118
3.4	Effective Degradability (%) of Hay Organic Matter Disappearance from Polyester Bags Incubated in the Rumen of Sheep and Treated with Different Washing Techniques at Selected Outflow Rates. Mean of 4 Sheep, one Observation per Sheep.	123
3.5	Coefficient of Variation (%) of Organic Matter Disappearance of Hay from Polyester Bags Incubated in the Rumen of Sheep and Treated with Different Washing Techniques.	125
3.6	Nitrogen Disappearance (%) of Hay from Polyester Bags Incubated in the Rumen of Sheep and Treated with Different Washing Techniques. Mean of 4 Sheep, one Observation per Sheep.	127
3.7	Nitrogen Concentration (gkg ⁻¹ DM) in Bag Residues.	132
3.8	The Effect of Multiple Washing on the Losses of DM (%) from Hay Samples Incubated in Polyester Bags. Means of 4 Observations.	134
4.1	Regression Statistics for the Prediction of <u>in vivo</u> OMD of Grass Silages.	143
4.2	Regression Coefficients and Significance of Between-Population Differences in MADF, LIGA, NCOMD and PCOMD Regression Equations for the Prediction of <u>in vivo</u> OMD for Silages.	147
4.3	Regression Coefficients for the Prediction of in vivo OMD by NB48 OMD and IVOMD.	153

-

 <u>Table</u>		Page
4.4	Bivariate Regression Statistics for the Prediction of <u>in vivo</u> OMD of 122 Silages.	154
4.5	Overall Population Statistics and Significance of Between-Population Differences in NIR Regression Equations for the Prediction of Silage OMD for Silages Included in the Calibration Set.	157
4.6	Wavelength Selection by Stepwise Multiple Linear Regression for the Prediction of <u>in vivo</u> OMD of Silages by NIR.	159
4.7	Validation Statistics for the Prediction of <u>in vivo</u> OMD of 48 Silages not Included in the Calibration Set.	162
4.8	NIR 8-term Equation Produced by Stepwise Multiple Linear Regression.	164
4.9	Calibration Statistics Produced by Stepwise Multiple Linear Regression for the Prediction of IVOMD by NIR.	168
4.10	Validation Statistics for the Prediction of IVOMD of 48 Silages not Included in the Calibration Set.	169
4.11	The Distribution of DOMD Values for SAC Advisory Service Silages and the Calibration Silages Used in this Study.	173
4.12	Coefficient of Variation and Regression Statistics for NB48 OMD Uncorrected and Corrected for the Internal Standard.	177
5.1	Characteristics of Silages Studied.	195
5.2	The Effect of Different Silage Characteristics on the Regression Equations Between <u>in vivo</u> OMD and the Laboratory Methods Studied in this Work.	197
6.1	Calibration and Validation Statistics for the Prediction of <u>in vivo</u> DOMD of Grass Silages.	206
6.2	Calibration and Validation Statistics for the Prediction of <u>in vivo</u> OMD of Grass Silages.	207
6.3	Calibration and Validation Statistics for the Prediction of Ash Content of Grass Silages.	208

,

Table		Page
6.4	Predictions of <u>in vivo</u> DOMD Using Various Routes by NIR.	210
6.5	Sources and <u>in vivo</u> DE Distribution of Silages Used in this Study.	214
6.6	Overall Regression Statistics and Significance of Between-Population Differences in Laboratories Methods Regression Equations for the Prediction of <u>in vivo</u> DE.	216
6.7	Overall Population Statistics and Significance of Between-Population Differences in NIR Regression Equations for the Prediction of Silage DE.	217
7.1	Methods Used by Advisory Laboratories in the UK.	224
7.2	Unit Cost of Analysis and Laboratory Turn-Around Time for Some of the Methods Tested in this Work.	228
7.3	Unit Cost of NIR Analysis.	228

Figure		Page
1.1	Estimated Production of Conserved Grass in the UK.	4
1.2	Schematic Representations of the Changes in the Chemical Composition of Grasses which Accompany Advancing Maturity.	7
1.3	Partitioning of Food Energy within the Animal.	12
1.4	The Two Possibilities that Exist Within the Different Advisory Organisations in the UK in Calculating ME.	1.6
1.5	The Relationship between the Major Measures of Digestibility.	40
1.6	Optical Geometry for Diffuse Reflectance Measurements based on Using 0° Illumination and 45° Collection with Large Area Photocells.	70
2.1	Distribution of <u>in vivo</u> OMD in the Populations.	84
2.2	Five Polyester Bags Ready to be Inserted into a Fistulated Sheep.	95
2.3	The Statistical Models Used to Detect Significant Differences between Populations.	109
3.1	Organic Matter Disappearance (%) of Hay from Polyester Bags Incubated in the Rumen of Sheep and Treated with Neutral Detergent Reagent.	120
3.2	Organic Matter Disappearance (%) of Hay from Polyester Bags Incubated in the Rumen of Sheep and Treated with Acid-Pepsin Reagent.	121
3.3	Organic Matter Disappearance (%) of Hay from Polyester Bags Incubated in the Rumen of Sheep and Treated with Washing Powder Reagent.	122
3.4	Coefficient of Variation (%) of Organic Matter Disappearance of Hay from Polyester Bags Treated with Different Washing Technique after 24 hours of Incubation.	es 126

Figure		Page
3.5	Nitrogen Disappearance (%) of Hay from Polyester Bags Incubated in the Rumen of Sheep and Treated with Neutral Detergent Reagent.	129
3.6	Nitrogen Disappearance (%) of Hay from Polyester Bags Incubated in the Rumen of Sheep and Treated with Acid-Pepsin Reagent.	130
3.7	Nitrogen Disappearance (%) of Hay from Polyester Bags Incubated in the Rumen of Sheep and Treated with Washing Powder Reagent.	131
3.8	Nitrogen Concentration (gkg ⁻¹ DM) in Bag Residue after 72 hours of Incubation.	133
3.9	The Effect of Multiple Washing on the DM Loss (%) from Hay Samples Incubated in Polyester Bags.	135
4.1	Relationship between MADF and in vivo OMD.	144
4.2	Relationship between LIGA and <u>in vivo</u> OMD.	145
4.3	Relationship between NCOMD and <u>in vivo</u> OMD.	148
4.4	Relationship between PCOMD and <u>in vivo</u> OMD.	149
4.5	Relationship between NB48 OMD and <u>in vivo</u> OMD.	150
4.6	Relationship between IVOMD and in vivo OMD.	151
4.7	Average NIR Spectrum of All Silages Studied in this Work.	156
4.8	Relationship between OMD Predicted by an 8–term NIR Equation and <u>in vivo</u> OMD.	158
4.9	Correlation Coefficient between NIR Spectra of 122 Calibration Silages and <u>in vivo</u> OMD.	160
4.10	Validation of an 8-term NIR Calibration Equation to Predict <u>in vivo</u> OMD.	163
4.11	Validation of IVOMD Calibration Equation to Predict <u>in vivo</u> OMD.	165
4.12	Validation of an 8-term <u>in vivo</u> NIR Calibration Equation to Predict <u>in vivo</u> OMD of 40 Irish Silages.	167
4.13	Relationship between IVOMD Predicted by 5-term NIR equation and IVOMD.	170

Figure		Page
4.14	Validation of 5-term NIR Calibration Equation to Predict IVOMD.	171
6.1	The Three Optional Routes Available to Predict DOMD.	205
7.1	The Normal Distribution Curve for the Prediction of Silage ME.	229
7.2	ME Prediction Error.	231
7.3	Ration Formulation Error.	232

.

ABBREVIATIONS

ADAS	Agricultural Development and Advisory Service									
ADF	Acid Detergent Fibre									
ADL	Acid Detergent Lignin									
AP	Acid Pepsin									
API	Acid Pepsin Washing followed by rumen Incubation									
APIAP	Acid Pepsin Washing followed by rumen Incubation followed									
	by Acid Pepsin Washing.									
CP	Crude Protein									
DANI	Department of Agriculture for Northern Ireland.									
DDM	Digestible Dry Matter									
DE	Digestible Energy									
DM	Dry Matter									
DOM	Digestible Organic Matter									
DOMD	Digestible Organic Matter in the Dry Matter									
ED	Energy Digestibility									
ESCA	East of Scotland College of Agriculture									
FEU	Feed Evaluation Unit									
GE	Gross Energy									
IAP	Rumen Incubation followed by Acid Pepsin Washing									
IND	Rumen Incubation followed by Neutral Detergent									
IVDMD	In vitro Dry Matter Digestibility									
IVDOMD	In vitro Digestible Organic Matter in the Dry Matter									
IVOMD	In vitro Organic Matter Digestibility									
IWP	Rumen Incubation followed by Washing Powder Washing.									
k _f	Efficiency of Utilisation of ME for gain.									
kg	Kilo Gram									
k,	Efficiency of Utilisation of Me for milk production									
k "	Efficiency of Utilisation of ME for maintenance									
LIGA	Acetyl Bromide Lignin Method									
MADF	Modified Acid Detergent Fibre									
MAFF	Ministry of Agriculture, Fisheries and Food									
ME	Metabolisable Energy									
MJ	Mega Joule									
Mt	Million Tonne									
N	Nitrogen									
NB48										
OMD	Nylon Bag 48 hours Incubation OMD.									

NCD	Neutral Detergent Cellulase DOMD
NCDOMD	Neutral Detergent Cellulase DOMD
NCOMD	Neutral Detergent Cellulase OMD
ND	Neutral Detergent
NDA	No Data Available
NDF	Neutral Detergent Fibre
NE	Net Energy
NDI	Neutral Detergent Washing followed by Rumen Incubation
NDIND	Neutral Detergent Washing followed by Rumen Incubation
	followed by Neutral Detergent Washing.
NIR	Near Infrared Reflectance
NOSCA	North of Scotland College of Agriculture
NR	Nitrogen Concentration in Bags Residue
NS	Not Significant
ODM	Oven Dry Matter
ОМ	Organic Matter
OMD	Organic Matter Digestibility
PCA	Principal Component Analysis
PCDOMD	Pepsin-Cellulase DOMD
PCOMD	Pepsin-Cellulase OMD
POP	Population
RRI	Rowett Research Institute
RSD	Residual Standard Deviation
SAC	Scottish Agricultural Colleges
SD	Standard Deviation
SEC	Standard Error of Calibration
SED	Standard Error of Difference
SEP	Standard Error of Prediction
TDM	Toluene Dry Matter
WP	Washing Powder
WPI	Washing Powder Washing followed by Rumen Incubation
WPIWP	Washing Powder Washing followed by Rumen Incubation
	followed by Washing Powder Washing.

WSC West of Scotland College.

XI

SUMMARY

- 1 The areas covered in the literature review include:
 - a) Metabolisable energy as a measure of the nutritive value of grass silages;
 - b) The importance of digestibility as a useful index of nutritive value;
 - c) Factors affecting the digestibility measurements;
 - d) Laboratory methods for predicting the organic matter digestibility of grass silages.
- 2 The effect of different washing procedures on the losses organic matter and nitrogen from samples of hay of incubated in polyester bags within the rumen of sheep was investigated. For organic matter, post-incubation detergent washing reduces variability without altering the form of the degradation curve. For nitrogen, postincubation detergent washing might remove contaminating bacteria which could otherwise lead to the underestimation of protein degradability. Washing the bags after rumen incubation with domestic washing powder in the washing machine is both cheap and convenient.
- 3 One hundred and seventy dried samples of grass silages which had been evaluated <u>in vivo</u> for organic matter digestibility (OMD), were collected from different sources around the UK. These sources include:
 - a) Agricultural Development and Advisory Service 100 silages;
 - b) Rowett Research Institute 43 silages;

XII

c) School of Agriculture, Aberdeen - 27 silages.

All silages were subjected to seven laboratory predictors of <u>in vivo</u> OMD, including those used routinely by advisory services in the UK. These methods are:

- a) Modified Acid Detergent Fibre (MADF) [Clancy and Wilson, 1966];
- b) Acetyl Bromide Lignin (LIGA) [Morrison, 1972];
- c) Neutral Detergent Cellulase OMD (NCOMD) [Dowman and Collins, 1982];
- d) Pepsin-Cellulase OMD (PCOMD) [Jones and Hayward, 1975];
- e) In vitro OMD (IVOMD) [Alexander and McGowan, 1966].
- f) Nylon Bag 48 Hours Incubation OMD (NB48 OMD) [Kridis et al, 1989];
- g) Near Infrared Diffuse Reflectance Spectroscopy (NIR) [Norris et al, 1976].

To avoid between-laboratory differences, each method was performed by a laboratory which makes routine use of the particular method.

Of all the 170 silages, 122 silages were selected to derive prediction equations (calibration silages) and the remaining 48 silages were reserved for subsequent validation purposes.

The aim of this work was to investigate the robustness of each method as a predictor of <u>in vivo</u> OMD and then explore the possibility of establishing an improved

XIII

technique which could be used by all of the advisory services in the UK.

The <u>in vivo</u> OMD of 122 calibration silages was not precisely predicted by the MADF ($\mathbb{R}^2 = 0.34$ and $\mathbb{RSD}_{6}^{8} =$ 5.1), LIGA ($\mathbb{R}^2 = 0.52$ and $\mathbb{RSD}_{6}^{8} = 4.4$), NCOMD ($\mathbb{R}^2 = 0.54$ and $\mathbb{RSD}_{6}^{8} = 4.3$) and PCOMD methods ($\mathbb{R}^2 = 0.55$ and $\mathbb{RSD}_{6}^{8} =$ 4.2). All gave significant between-population differences in the regression equations obtained.

The <u>in vivo</u> OMD was more precisely predicted by the rumen liquor methods (NB48 OMD and IVOMD) [$R^2 = 0.68$, 0.74; RSD% = 3.6, 3.2 respectively]. In each case their application require one single regression equation to describe all silage populations.

Of the methods tested in this work, the NIR method was the best predictor of <u>in vivo</u> OMD ($R^2 = 0.85$ and SEC% = 2.5). It also gave single regression line provided that more than five terms were used in the multiple regression equation.

The best prediction of <u>in vivo</u> OMD of a blind test of 48 <u>in vivo</u> silages was obtained by the NIR method using a multiple linear regression involving eight terms ($R^2 = 0.76$, SEP% = 2.6).

The effect of external factors in the relationship between <u>in vivo</u> OMD and its predictors was investigated. The IVOMD and NCOMD methods were significantly affected by the year of harvest. Cut number was also found to significantly affect the relationship based on the MADF and LIGA methods. The method of ensiling, wilting time, additive application and nitrogen fertilisation were found not to affect the regression equations of any predictor studied in this work.

4

5

XIV

The use of the NIR method to predict <u>in vivo</u> DOMD was examined. This method can predict directly <u>in vivo</u> DOMD, however with lesser precision than predicting <u>in</u> <u>vivo</u> OMD ($R^2 = 0.64$, SEP% = 2.97 and $R^2 = 0.76$, SEP% = 2.6 respectively). The calculation of <u>in vivo</u> DOMD by NIR prediction of <u>in vivo</u> OMD and then measuring ash content, a parameter useful to indicate soil contamination, gave more precise prediction than the direct prediction of <u>in vivo</u> DOMD by NIR ($R^2 = 0.78$, SEP% = 2.43 and $R^2 =$ 0.64, SEP = 2.97 respectively).

The laboratory methods tested in this work were used to predict the <u>in vivo</u> DE of 140 grass silages. Digestible energy was not predicted with sufficient precision by the conventional methods tested in this work (for the MADF, LIGA, NCOMD, PCOMD and IVOMD methods the R² and RSD were 0.09; 1.65, 0.22; 1.53, 0.16; 1.60, 0.32; 1.43 and 0.31; 1.44 respectively). Only the NIR method makes a significant improvement in DE prediction (R² = 0.72 and SEC = 0.91). The gross energy of silages can account for much of the variation in DE prediction, suggesting the importance of gross energy in the energy evaluation of silages.

7

6

GENERAL INTRODUCTION

Agriculture is considered to be one of the oldest activities of mankind. It is not clear when people started agricultural cultivation and domestication of animals, but it is believed that it have taken place gradually from 40,000 to 10,000 B.C. may (Flannery, 1965). Before that time, people were principally food gatherers, but the time came when they settled and started to become food producers. Nowadays, food production is the world's which neither modern vital primary activity without most civilisation nor primitive cultures can survive.

Due to the huge expansion of the world's population and increased demand for food in the last 100 years, it became imperative to maximise the output of meat, milk and wool from animals - and yet minimise the costs of its production.

Traditional methods of preserving food, such as salting, drying, syruping, pickling, smoking and fermenting, were established by early civilisations. For example, ensiling the plant material in the absence of air results in the production of a fermented crop which is known as silage.

History of Silage Making

Silage is a material of high moisture content stored anaerobically and produced by a controlled fermentation. The process is called ensilage and the container or the structure is called a silo.

It is believed that silage making was practised 3,000 years ago (Schukking, 1976). The ancient Egyptians were believed to be familiar with silage making as long as 1000-1500 B.C. However, it wasn't until the latter part of the nineteenth century that interest in this process became more widespread. Practical modernisation of ensilage was undoubtedly credited to the French farmer Goffart in 1877. Five years later, silage making received interest among British farmers. In 1883 it was reported that no more than half a dozen silos were found in the whole of the UK, and as the interest increased dramatically there were about 1,605 silos found in the country by 1886 (Rew, 1888). The publication of a book "Sweet Ensilage" by Fry in 1885 was a major hindrance in accepting this technique for some 50 years. This is because of Fry's remarks of allowing the crop to heat up in the silo prior to sealing which resulted in producing a silage of a poor nutritional quality.

After the second World War, a wide general interest in silagemaking was revived. This was owing to the escalating costs of feed concentrates, necessity of more intensive animal production and improvements in mechanisation. Nowadays, silage-making is an accepted method of food conservation world-wide. The principles of ensilage are now more fully understood and the conditions necessary for obtaining a good product are well defined.

In many countries silage has superseded hay as the principal conserved forage, as the figures for European countries show (Table 1).

Conserved Forages in the UK

Ż

In the UK, ensilage of a grass has become an increasingly important method of conservation. Despite the fact that this method has been practised for many centuries, as mentioned earlier, it is less than 20 years since UK farmers started to adopt the technique. For example, in the West of Scotland silage making is now considered to be the most important method of conservation for feeding livestock in winter.

Although haymaking has long been the traditional technique of preserving forage, it was not until 1981 that the acreage of grass conserved as silage exceeded the acreage conserved as hay (Figure 1.1). According to the UK agriculture census (Henderson, 1987) in England in 1985, 667,000 hectares of grass were cut for hay, 1,018,000 hectares were cut for silage and 10,100 hectares were

2

TABLE 1

Producti	on	of	Silage	and	Hay	in	Europe:	Est	timat	es	for	the
Countries	of	the	e Euroj	pean	Econ	omi	c Commu	unity	and	Sc	andi	navia
		()	1978 -	1979))	[W:	ilkinson,	198	0]			

Country	Silage	Hay	Silage of Total (%)
Belgium	1.9	1.7	52.8
Denmark	1.1	0.4	73.3
Finland	0.9	1.7	34.6
France	16.0	22,3	41.8
Germany (FRG)	10.5	9.1	53,6
Iceland	2.4	3.6	40.0
Italy	5.9	9,3	38.8
Luxemburg	0.2	0.2	50.0
Netherlands	4.1	1.3	75.9
Norway	1.1	0.7	61.1
Sweden	1.0	4.6	17.9
United Kingdom	5.5	6.9	44.4

Figures expressed as x 10^6 tonnes of dry matter.

FIGURE 1.1

Estimated Production of Conserved Grass in the UK (From Wilkinson, 1985)



cut for artificially dried fodder. This represented an increase of 18.7% in the area cut for silage over the previous year. There was also an increase of 32.9% in other crops harvested for silage.

Silage making has its own advantages over hay making. Woolford (1984) indicated that silage making is less dependent on the weather, the crop is harvested at a less mature stage of growth than for hay and thus at the outset, is more nutritious and two be taken for silage, with hay it three cuts per year can or is usually one. In addition, silage is generally superior to hay in terms of its content of energy and crude protein (Wilkinson et al, 1976) and apparently superior to the original crop in terms of energy (Alderman et al, 1971).

The Nutritive Value of Grass Silage

Grass silage is an important forage in feeding ruminant livestock in the UK. The determination of its nutritive value is of primary concern to farmers so that ration formulation can be achieved. While the nutritive value of silage is a measure of its ability to satisfy the energy and protein needs of the animal, much of the effort has been directed to characterise silage as an energy source.

In the UK, metabolisable energy has been used to estimate the nutritive value of silages. While this parameter is difficult to measure, digestibility is more frequently used as a useful index of the nutritive value.

Factors Affecting the Digestibility and Energy Value of Grass Silages

A Composition of The Parent Crop

There are many factors which affect the composition of the harvested crop. These include, species and strain of the

plant, stage of growth, application of fertilisers, climate, type of soil and growth condition. The stage of growth is the most important factor influencing the composition and nutritive value of the standing crop. As plants grow, there is a greater need for structural tissue, and therefore the structural carbohydrates (cellulose, hemicellulose) and lignin increase. This increase in cell wall components is coupled with a parallel decrease in the digestible cell contents. Figure 1.2 demonstrates how the cell components change as the plant advances in maturity.

B The Effect of Ensiling on Digestibility and Metabolisable Energy of Grass Silages

Digestibility is considered to be a useful index of a forage nutritive value. It is well documented that the digestibility of a good quality silage is similar to the digestibility of the crop from which it was made (Harris and Raymond, 1963). More recent studies by McDonald and Edwards (1976) confirmed these findings. They found that the digestibilities of 36 different silages and grasses from which they were made were 0.768 and 0.767 respectively.

Wilting generally results in a small reduction in digestibility (Harris <u>et al</u>, 1966; Marsh, 1979). The depression in digestibility is likely to be a reflection of changes occurring in field drying rather than during ensiling (Thomas and Thomas, 1985).

Formic acid may increase digestibility slightly (Castle, 1972; Wilson and Wilkins, 1973; Thomas and Thomas, 1985), although Waldo (1978) claimed a 7% increase in digestibility of the organic matter of silages studied. Formalin may reduce OMD by 2-4 units (Wilkins <u>et al</u>, 1974). This depression of digestibility by the addition of formalin may be due to the disturbance of the rumen function with the consequence that cellulose digestion is reduced.

6

٠,

Schematic Representations of the Changes in the Chemical Composition of Grasses which Accompany Advancing Maturity. All Values are in Percentages (From Osbourn, 1980).



Metabolisable energy values for silage vary widely and this is dependent on the type of fermentation during ensilage. For example, silages in which there has been extensive butyric may have much higher ME values, sometimes fermentation exceeding 13 MJkg⁻¹DM (Woolford, 1984). Wilkins (1974) quoted 8 comparisons between ME values of fresh grass and ensiled grass and found that the mean value was 10.45 MJkg⁻¹DM for MJkg⁻¹DM for 10.41 silage. grass However. and in an experiment by Donaldson and Edwards (1976) in which lactate, wilted and formic acid-wilted silages were made from the same ryegrass herbage, they found that the ME value of the lactate silages were significantly higher than that of the fresh grass and those of the other two silages in which fermentation had been restricted. In both studies the loss of energy as methane was estimated. Because wilting restricts fermentation in the silo, it is not expected that wilting silage will change the metabolisable energy values from the original crop. This view is supported by the study of Donaldson and Edwards (1976) where they found that the metabolisable energy values of grass and wilted silages made from the same grass were 11.6 and 11.4 MJkg⁻¹DM respectively.

Prediction of Silage Digestibility and Metabolisable Energy in the UK

It is well known that the measurement of digestibility and metabolisable energy of forages using animal trials is the ultimate standard of measurement. However, this method is expensive, time consuming and requires special facilities.

In the UK therefore, different advisory laboratories have used different methods to predict these two parameters. Up to now, these methods have not been compared properly. For the advisory purposes, economic consideration plays an increasingly important role for the selection of the advisory method(s), however, nutritional considerations should not be ignored. It is these aspects which form the theme of this work.

8

The aims and objectives of this work are as follows:

- 1 Test the possibility of improving the reliability of the nylon bag technique by washing the bags after or before rumen incubation with different washing reagents.
- 2 Examine the ability of various laboratory methods, including those used routinely by advisory services in the UK, to predict <u>in vivo</u> OMD of a large population of grass silages. These silages were collected from different sources around the UK.
- 3 Explore the possibility of establishing an improved predictive technique which could be used by all advisory laboratories in the UK.
- 4 Investigate the effect of environmental and other factors on <u>in vivo</u> digestibility – laboratory method relationship for each predictive technique.
- 5 Study the ability of these predictive techniques to predict the digestible energy of grass silages.

CHAPTER ONE

REVIEW OF LITERATURE

Energy Evaluation

1.1.1 Introduction

When food is burned completely in a bomb calorimeter, energy is released and this is termed as the gross energy (GE) of the food. A scheme showing the manner in which the GE in herbage is partitioned and finally used for productive purposes in ruminants is shown in Figure 1.3.

1.1.2 Historic Developments of Forage Evaluation in the UK

The Starch equivalent system of Kellner and Kohlen (1900) was the basis of expressing the energy value of ruminant feeds in the UK. This system was introduced by Wood (1917) and adopted in the UK by 1921.

After World War II, calorimetric measurements began to be made again under the direction of Kenneth Blaxter at the Hannah Dairy Research Institute (Blaxter et al, 1954). He was the first person to propose a new approach to energy systems based on the determination of ME, to describe the values of feeds and the requirements of ruminants (Blaxter, 1962). His approach was accepted by the UK Agriculture Research Council (ARC, 1965) and subsequently adopted as the official advisory method for allocating energy allowances for ruminants at a Joint Conference on "Nutrient Standards for Ruminants" held in London on 12 April 1972. Thereafter, an Energy Requirements Working Party was set up to implement the system and concluded that the new system was superior to the Starch equivalent system and recommended its adoption, but in a modified form more suitable for advisory purposes. This led to the publication of a practical manual, Technical Bulletin No 33 (MAFF, 1975), describing the derivation and use of the modified ME system in detail.

FIGURE 1.3

Partitioning of Food Energy within the Animal (after MAFF, 1975)



12

۰.

• •

A major problem which faced the advisers when it was decided to change to an ME system in 1972, was which sources of information should be used to obtain an estimate of the ME values of feedstuffs.

At that time, the Working Party preparing the Technical Bulletin 33 felt that while there was no correct estimate of ME values, factors may be used to convert the digestible proximate constituents of a food to ME values as indicated in equation No 2 in Technical Bulletin 33. These were factors proposed by workers at the Oskar Kellner Institute at Rostok, West Germany:

ME (MJ kg⁻¹DM) = 0.0152 DCP + 0.0342 DEE + 0.0128 DCF + 0.0159 DNFE

Where:

DCP	=	Digestible	Crude	Protein	g/kg	
DEE	=	Digestible	Ether	Extract	g/kg	
DCF	=	Digestible	Crude	Fibre	g/kg	
DNFE	=	Digestible	Nitrog	en-Free	Extractives	g/kg

The use of such an approach was accepted in the UK by the advisory organisations as a means of smooth and rapid change-over to be made in 1975 from a starch equivalent system to an ME system. Eventually, however, the weakness of such an approach began to be realised because it was questionable whether the ME values obtained from nineteenth century German studies would be applicable to all British foodstuffs grown lately under modern conditions. These uncertainties provided some of the reasons for the establishment of two Feed Evaluation Units at Rowett Research Institute in Scotland in 1972 and the ADAS Nutrition Chemist's Unit in England in 1976. The aim of these units was to provide feed values for the different British feeds, but more importantly, it was
asked to provide a simple way of estimating ME from readily determined parameters.

1.1.3 ME as a Measure of Nutritive Value of Grass Silage

The ME of food is measured in a trial similar to the but additional steps have be digestibility trial to determined. The urine must be collected and its GE value measured. More importantly, the determination of the losses in combustible gases, mainly methane, is required and calls for the use of an expensive and complicated apparatus, respiration chamber, and involves much а labour as well as technical skill accordingly. The measurement of methane is an essential part of the determination of the ME content of feed for ruminants.

Ideally, the ME values used by the advisers should be directly determined by the animal accounting to all of the energy losses of GE. However, such an approach is labour intensive, expensive and requires special facilities and consequently, this direct method is not likely to be a practical proposition.

The use of the table of feed composition to assign a single ME value for the ME content of silage is not feasible. Edwards (1986) has argued against such a proposition and concluded that for the comparatively small number of relatively uniform samples examined in the Rowett Research Institute Studies (Mean = 11.44 MJ kg⁻¹ DM, SD = 0.787), it may be calculated that 52% of samples would differ from the mean by more than 0.5 MJ kg⁻¹DM and 20% by more than 1 MJ kg⁻¹DM.

The search for a simple laboratory method to accurately predict the ME of grass silage has been continued vigorously for many years, in particular at the two feed evaluation units mentioned previously. The studies involved a comprehensive measurement of in vivo ME

with silages together appropriate values of grass using different laboratory prediction equations measurements. These studies have been described in detail in the Rowett Research Institute (1975), Wainman et al (1978 and 1984), Barber et al (1984) and Givens et al (1989a).

1.1.4 The Dilemma of Estimating ME for Ration Formulation

Since the implementation of the ME system, the different advisory laboratories in the UK have used different routes in which ME values of grass silages can be predicted. Figure 1.4 shows the two optional routes that existed in the UK for calculating ME values and subsequently the use of these values for ration formulation.

Scottish Agricultural Colleges (SAC) used and are still using the first route where they predict the OMD of silage, measure ash, calculate DOMD and then apply a constant factor to calculate ME. In the past, ADAS used the first route and by 1983 they changed to use the second route (Barber <u>et al</u>, 1984; Givens, 1986, Givens <u>et</u> <u>al</u>, 1989a). In addition to the different methodologies of estimating ME, different laboratory methods have been used to predict ME values (Table 1.1).

Scientifically, route (II) is the preferred approach. However, two aspects have to be clarified before such a conclusion can be made:

- A The Reliability of the Current ME Measurements!B What is the Efficiency of ME Utilisation?
- A <u>The Reliability of the Current ME Measurements</u> Since the ME system was adopted in the UK, up to now there have been relatively small numbers of grass silages which have measured <u>in vivo</u> ME. The only documented in vivo ME values were the ones

. , , ,

The Two Possibilities that Exist Within the Different Advisory Organisations in the UK in Calcuating ME.

.



TABLE 1.1

Methods Used to Predict ME by UK Laboratories

		UK Laboratories				
Methods	ADAS	WSC	NOSCA	ESCA	DANI	Reference
IVOMD	x	x				Tilley & Terry, 1963 (ADAS); Alexander & McGowan, 1966 (WSC)
MADF	x	х		x	x	Clancy & Wilson, 1966
LIGA			x			Morrison, 1972
NCD	x					Downan & Collins, 1982

Institute (RRI) publicised by Rowett Research reports (RRI, 1975, Wainman et al, 1978 and Despite the fact that ADAS feed evaluation 1984). unit (FEU) reported а measured in vivo ME. however, the methane production was predicted by the Blaxter and Clapperton (1965) equation due to unavailability of a respiration the chamber to measure energy loss as methane gas.

Barber et al (1984) reported prediction equations for silages based on the ADAS FEU data, these equations are listed on Table 1.2. They concluded that the associations relatively poor of the different predictions of ME must be caused by variation in calculating in vivo ME values due to other factors and the accuracy of prediction of DOMD was somewhat better and was comparable to those for As more silages become fresh grass and hay. available, these equations were updated by Givens (1986) and Givens et al (1989a).

Edwards (1986) analysed the equation for predicting ME content from MADF and NCD proposed by Givens The values of standard error of prediction (1986).(SEP) for predicted silage ME of 11.5, 10.8, 10.0 and 9.3 MJ kg⁻¹TDM by using MADF equation were 0.97, 0.87 0.99, and 1.0 respectively. He concluded that at a predicted ME of 10 MJ kg⁻¹TDM, some 57% of estimates will differ from the true value by more than \pm 0.5 MJ kg⁻¹TDM and 25% by more than + 1 MJ kg⁻¹TDM. Similarlv. at a predicted ME of 10 MJ kg⁻¹TDM using the proposed NCD equation some 60% of estimates will differ from the true value by more than + 0.5 MJ kg⁻¹TDM and 31% by more than + 1.0 MJ kg⁻¹TDM.

TABLE 1.2

0.968 0.969 0.939 0.944 RSD (%) 1.03 3.23 3.84 l.03 2.93 2.70 3.64 3.97 (%) 62.6 67.3 76.9 53.6 46.3 48.4 25.6 34.6 37.5 32.4 20.3 19.2 \mathbb{R}^2 68 55 56 79 77 79 68 55 56 79 77 79 Ħ MADF% % 8.10 + 0.881 IVD % % ≫ - 0.583 NDF % 0.883 ADF % 4.35 + 0.092 NCD % + 0.132 IVD % 0.175 MADF% 0.084 NDF % 0.122 ADF % $= 25.04 \pm 0.605$ NCD + 0.100 CD = 26.55 + 0.651 CD 98.31 - 1.04 95.20 -I ۱ 1 1.90 96.31 4.51 = 15.97 = 14.55 = 14.91 11 H It It II 11 II ME (MJ k g^{-1} DM) ME (MJ k g^{-1} DM) ME (MJ k g^{-1} DM) ME (MJ kg⁻¹DM) ME (MJ kg⁻¹DM) ME (MJ kg⁻¹DM) DOMD % DOMD % DOMD % DOMD % DOMD % DOMD %

DOMD and ME Prediction Equations for Silages (Barber et al. 1984)

The use of MADF to predict the ME content of 48 silages of RRI (RRI, 1975; Wainman <u>et al</u>, 1978 and 1984), was also commented on by Edwards (1986). The residual standard deviation (RSD) of the prediction equation was 0.76. He concluded that at a predicted value of 11.5 MJ kg⁻¹DM, 51% of estimates will differ from the true value by more than \pm 0.5 MJ kg⁻¹DM and 19% by more than \pm 1 MJ kg⁻¹DM.

From the foregoing discussion it is clear that the current prediction of ME using route (II) is unsatisfactory. An attempt to examine some of the problems associated with the measurement of the ME content of silage and the areas which need further research will be presented below.

1 GE of Silage

It is commonly assumed that food in general has a $MJ kg^{-1}DM$ GE value of 18.4 (MAFF, 1975). However, it is unlikely that this common value is true for silages and many workers have commented on the apparently high GE of silages. Alderman et al (1971) reported a mean value of 45 grass silages of 20.17 MJ kg⁻¹DM. This value is 10% higher than the usual value of 18.4 MJ for herbage. McDonald and Edwards (1976) obtained GE values of 18 silages of 20.0 MJ kg⁻¹DM and 18.4 MJ kg⁻¹DM of the original grass, representing an increase during ensiling of 8.7%.

The reason for these high GE values of silages stems from the fact that ensilage is an energy concentrating process and that it comes through the replacement of low energy substrates by high energy fermentation products. Several of the products of fermentation are of higher GE value than the original substrates. For example, butyric acid and ethanol have GE values 24.9 and 29.8 MJ kg⁻¹DM respectively, compared with glucose and fructose which have values of 15.6 and 15.7 MJ kg⁻¹DM respectively.

The GE of silages is not only high compared to the original material before ensiling, but also variable. Table 1.3 illustrates the wide range of gross energy contents of grass silages accumulated from different institutions and demonstrates the danger of applying a fixed value for all silages. For these 238 grass silages, the GE mean was 19.1 MJ kg⁻¹DM with a standard deviation of 0.71 and a range of 17.0-21.1 MJ kg⁻¹DM.

It is not surprising that the GE of silages is widely variable (see Table 1.3) considering the nature of the fermentation process that takes place inside the silo and the production of fermentation end products with variable GE contents. It is this characteristic which makes it difficult to measure GE of silages.

Sampling problems is a major contributor to the variation of GE of silages. A recent study bv Norcross et al (unpublished) where she determined sampled the GE silage of by а variety of techniques, indicated that there were significant differences (P < 0.05) between the treatments studied. The nature and quantity of volatiles are in part responsible for the variation in the GE value of silage, therefore any loss of volatiles from the exposed silages result in a change in GE value. Other losses of energy when silage is exposed to air through the oxidation processes are of aerobic deterioration.

TABLE 1.3

Gross Energy Contents of Grass Silages (MJ kg⁻¹DM*) [Edwards, 1986]

Source	N	Mean	SD	Range
Edinburgh School of Agriculture	60	19.2	0.67	17.8-20.7
MAFF/ ADAS	126	19.0	0.80	17.0-21.1
Rowett Research Institute	48	19.2	0.49	18.1-20.2
Hannah Research Institute	4	19.0	0.93	18.1-20.0

* Dry matter determined by toluene distillation with a correction for ethanol.

Conventionally, the GE of silages is measured with an adiabatic bomb calorimeter. However, different varieties of techniques have been used with regard to the primer and currently polythene is the most Reproductibility of the determination of common. the GE of silage has been found to be difficult, therefore a trimmed mean approach was used to exclude high GΕ figures (Norcross et al, unpublished).

The measurement of <u>in vivo</u> ME in a ring test involving ADAS FEU, DANI and NOSCA indicated that, although there was good between-centre agreement with regard to <u>in vivo</u> digestibility, the ME values obtained varied between laboratories. The problems in the measurement of GE in each centre was stressed (Unsworth, unpublished).

These variations in GE determination of silages focused the attention on the use of the GE values as a bivariate in ME prediction equations. Givens and Brunnen (1987) and Givens et al (1989a) stressed this aspect strongly and indicated that the GE of silages accounts for 50% of the variability in ME concentration and yet it is ignored in present laboratory prediction methods. For example, the accountable variance $(R^2 \)$ for the prediction of in vivo ME concentration for 115 silages increased from 23.9% when IVOMD was used as a single predictor to 77.4% when GE was included as a bivariate (Givens et al, 1989a). However, it is important to realise that this approach is unlikely to be adopted for advisory purposes because of the difficulties in measuring GE for silages and also the lack of a quick and easy method to measure GE for advisory purposes.

2 The Expression of the DM Base

concentration of silages is Routinely, the ΜE currently expressed on an oven dry matter (ODM) basis which involves drying the silage samples in 105°C for 24 hours. This method an oven at however, has been recognised for a long time to be inaccurate for fermented products because it fails to of consider the loss drv matter as volatile substances during the normal drying process.

These volatiles are part of silage dry matter and many workers have drawn attention to the importance of taking into consideration the volatiles lost on oven drying in order to assess the true dry matter content of silage (Haigh and Hopkins, 1977). Different methods have been suggested by many workers to determine the true dry matter content of silages, but a toluene distillation method developed by Dewar and McDonald (1961) has been recommended for the routine determination of dry matter content in silage.

The losses of these volatile compounds can be significant during the normal oven drying procedure. For example, oven drying at 100°C can lead to dry matter losses of up to 16%, depending on the quantity of organic acids present (Minson and Lancaster, 1963). Compared with the toluene dry matter method, the losses of the dry matter of silages can be 9.7, 6.8, 8.6 and 10.7% when the dry matter was determined by freeze drying, microwave drying and oven drying at 70 and 100°C respectively (Aerts et al, 1974).

Barber <u>et al</u> (1984) stressed the importance of including the weight of volatile compounds and indicated that by necessity, the ME concentration

equations must be expressed on true dry matter Table 1.1]. More recently, (TDM) [see Givens (1986) confirmed this aspect and concluded that the expression of silage ME values on an ODM basis is both mathematically and biologically incorrect and gives rise to erroneously high values. Therefore he replaced the old prediction equation which was used to predict ME on an ODM basis by a new equation Accordingly, this equation based on a TDM basis. has been used by ADAS Nutrition Chemists from mid 1985.

The new proposed ADAS FEU equations (Barber et al, 1984) were found to give a predicted ME value 7-12% higher than that predicted from equations in current use. This difference depends on the amount of volatile compounds present and also the fermentation quality of the resulting silage. These high ME values raised suspicion among advisers about the applicability of these new equations ration in formulation and evaluation. Assessment studies performed by Barber et al (1980) and Barber et al (1983), suggested that the current ME equation which is based on ODM gave no bias prediction of animal performance as compared with the new proposed equation. An over-prediction of 7-10% of milk yield has been caused by the use of the new equation. Possible explanations suggested for this discrepancy between actual and predicted yield of dairy cows caused by the new ME values include loss of silage volatiles during the normal feeding conditions on the farm which are initially included on the calculated ME value of the silage and the over-estimation of the efficiency of utilisation of ME.

At present, no routine method can be used for advisory purposes to measure the true dry matter of

The method of Dewar and McDonald (1961) silages. is lengthy and complicated and therefore, prediction equations were suggested to predict silage volatiles. 150 advisory West of Scotland College, At the silages were examined during 1982 to establish regression equations which can be used to predict silage volatiles, however no satisfactory relationship has been found for this purpose (Barber, et al. Barber et al (1984) and Givens (1986) 1983). suggested a single fixed correction factor of 2.3% convert ODM to TDM. However, and 1.9% to considering the nature of silage fermentation and the variability of the volatiles raises questions about the validity of this approach.

For practical reasons, the advisory laboratories in the UK has been forced to ignore the volatiles contained in the silages unless an accurate and routine method can be discovered to measure these important compounds.

3 Measurement of Methane Energy

To determine in vivo ME, faecal. urinary and methane energy needs to be measured. The faecal and urinary energy can be easily determined using conventional metabolism cages. On the other hand, methane energy loss measurement is more complicated and requires special equipment. Understandably, many workers tried to predict this parameter from the knowledge of the amount and type of the food eaten by the animal (Bratzler and Forbes, 1940; Swift et al, 1948; Blaxter and Clapperton, 1965).

Several workers measured methane production and found it to be remarkably similar to the figure of 8% of GE commonly assigned in general for foods, suggesting that methane production differs very little from this common number. For example, Ekern and Sundstol (1974) found that methane production for two silages was 7.5 and 7.9% and for two hays was 8.1 and 8.3% made from the same grass. Sundstol and Ekern (1976) found methane production for fresh grasses, hays and silages made from the same grasses was 7.5, 7.0 and 6.6% respectively. The examination of methane production figures reported by RRI (RRI, 1975; Wainman <u>et al</u>, 1978 and 1984) revealed that the losses of energy as methane for 39 dried grasses, 26 hays and 48 silages were 7.4, 7.9 and 7.7% respectively (Edwards, 1986).

In the UK, the regression equation derived by Clapperton (1965) Blaxter and at the Hannah Research Institute is used routinely to predict methane production in cattle and sheep. It has been used extensively at the ADAS FEU so that in vivo ME values can be calculated (Givens, 1986). This equation was derived by using 48 different diets and in excess of 2500 determinations of methane production by cattle and sheep. It is defined as:

Cm = 3.67 + 6.2D

where Cm is the percentage energy lost as methane per 100 units of GE and D is the apparent digestibility of GE.

Recently, doubts have been raised about the validity of the Blaxter and Clapperton equation to predict methane production for grass silages. Edwards (1986) compared the observed energy losses as methane with those calculated by the Blaxter and Clapperton equation. He found that for the 48 silages reported by RRI (RRI, 1975; Wainman <u>et al</u>, 1978 and 1984) the mean difference (observed -

calculated) was -0.48% with SD of 0.711. This that by using this equation. methane suggests over-estimated bv 0.48%. He production is concluded that this bias is not of major importance since it could be easily corrected and moreover, it is only 0.1 MJ of GE. The latter worker however, did not report the regression statistics obtained by regression equation using the Blaxter and the Clapperton's equation variables, therefore the same exercise was repeated here. The bias of -0.48 and SD 0.711 was confirmed. however the of no relationship can be found between methane production and the apparent digestibility of GE ($R^2 = 0.0$). This suggests that for silages the use of Blaxter and Clapperton's equation is not justified and may be no better than the mean determined value of 7.7%.

B What is the Efficiency of ME Utilisation?

Ŋ

ś

In progressing from GE to NE, measurement becomes increasingly difficult, but in compensation more accurate estimates of the energy required for maintenance and production are achieved. Therefore, DE can be considered a first approximation to nutritive value and NE the most precise (Ulyatt, 1973). It is NE which both the nutritionist and the farmer is interested in, thus an accurate estimate of this parameter is of primary concern to them. The ratio of NE to ME contents represents the efficiency of ME utilisation denoted as k.

The methods of measuring the efficiency of ME utilisation have been described in detail by Blaxter (1962). Two types of methods are basically used, comparative slaughter and calorimetry experiments. Comparative slaughter experiments require little equipment, but are expensive in terms of animals. In calorimetry, the animals are kept in calorimeters and these call for elaborate and expensive equipment and labour input. For silages, few calorimetric studies have been made to determine the efficiency of ME utilisation. Consequently, a means of predicting the efficiency using a suitable regression equation was used. In the UK, the equations derived by the ARC (1980) were used for this purpose. These equations are:

 $k_{\rm m} = 0.35 q_{\rm m} + 0.503$ $k_{\rm f} = 0.78 q_{\rm m} + 0.006$ $k_{\rm l} = 0.35 q_{\rm m} + 0.42$

where \boldsymbol{q}_m is the metabolisability (ME/GE).

For advisory ration formulation purposes, ${\rm k_m}$ and ${\rm k_1}$ were assigned single values of 0.72 and 0.62 respectively, and for ${\rm k_f}$ the following relationship was given (MAFF, 1975):

 $k_{f} = 0.0435 \text{ M/D}$

where M/D is the ME per kg DM.

Thomas and Thomas (1985) reported k_m values for 8 silages and these were compared with calculated values using the ARC (1980) equation (Table 1.4). These workers found that the calculated k_m values were always higher than k_m values determined calorimetrically. The mean difference between the two values were 0.04 and the two means were significantly different (P < 0.001).

Similarly, McDonald (1983) showed that the calculated k_f values using the ARC equation (1980) were on average, higher than the k_f values determined calorimetrically (Table 1.5). The mean difference between the two values were 0.09 and the two means were significantly different (P < 0.05). For k_f values, we can see from the table that the size of the difference between calculated and determined values was much greater than the k_m values. Some of the individual differences were extremely large, which causes a lot of concern.

TABLE 1.4

The Partial Efficiency of Utilisation of Metabolisable Energy for Maintenance (k) Determined Calorimetrically and Calculated by the Equation of ^mAgricultural Research Council (1980) for Eight Silages (after Thomas and Thomas, 1985)

Silage	Determined k m	Calculated k * m
1	0.69	0.75
2	0.71	0.74
3	0.68	0.70
4	0.66	0.70
5	0.71	0.71
6	0.68	0.70
` 7	0.65	0.74
8	0.66	0.74
Mean	0.68	0.72
SD	0.023	0.022
SE of D	ifference 0.0	011

*k_m = 0.35 ME/GE + 0.503

TABLE 1.5

The Partial Efficiency of Utilisation of Metabolisable Energy for Fattening (k_f) Determined Calorimetrically and Calculated by the Equation of Agricultural Research Council (1980) for Fourteen Silages (after McDonald, 1983)

Silage	Determined k _f	Calculated k_{f}^{*}	Silage	Determined k _f	Calculated k_{f}^{*}
1	0.21	0.44	8	0.43	0.49
2	0.51	0.56	6	0.39	0.47
ę	0.42	0.42	10	0.34	0.53
4	0.41	0.45	11	0.23	0.61
2	0.49	0.43	12	0.61	0.61
9	0.54	0.51	13	0.52	0.61
<u> </u>	0.40	0.53	14	0.45	0.58
Mean k _f d	etermined 0.43				
SD -	0.112				
Mean k _f c	alculated 0.52				
SD	0.069				
SE of the	Difference 0.031				

* Calculated from $k_f = 1.32 \text{ ME/GE} - 0.318$

Several workers commented about the low k_f values obtained for Thomas sometimes silages. and Chamberlain (1982) suggested that the very low figures obtained for k_f by some experiments (see Table 1.5) could partly be explained by technical difficulties in obtaining correct estimates, but in the other experiments the low k_f values represent a true inefficiency in energy utilisation of silages. Thomas Thomas (1985)indicated that these low and efficiencies mav arise because of а nutritional imbalance in silage diets, resulting from the poor utilisation of silage nitrogen in the rumen. The work of Donaldson and Edwards (1977 and 1979) supported this view where they found high rumen ammonia concentrations when the animals were fed silage diets only. Furthermore, Kelly and Thomas (1978) found that the low k_f values were only apparent when the silage was given alone, and therefore this might not be of a major concern under farm conditions where silages are given as part of a mixed diet.

Thomas and Thomas (1985) also suggested that silages might contain specific substances which could disturb the energy metabolism inside the animal. They further added that dried silage extracts can increase the basic metabolic rate of rats, due apparently to the presence of certain flavenoids (McLaren et al, 1964; Qasim and Stelzing, 1973).

calorimetric determinations Verv few have been for carried out to measure grass silages. k₁ Unsworth and Gordon (1985) reported a k_1 value of 4 silage diets supplemented with 40% concentrate. The comparison was made between determined k and predicted k_1 using the ARC (1980) equation (Table The discrepancy between determined k_1 and 1.6).

TABLE 1.6

•

Efficiency of Utilisation of Metabolisable Energy of Silage Diets for Lactation (after Unsworth and Gordon, 1985)

Silage	k,	L
	Determined	Calculated*
a. Unwilted	0.56	0.65
Wilted	0.58	0.64
b. Unwilted	0.50	0.64
Wilted	0.53	0.65

* $k_1 = 0.35 q_m + 0.42$ (ARC, 1980)

SE of Difference = 0.018

predicted k_1 was thought to be caused by the fact that the ARC (1980) equation was derived by using a non-silage diet (Unsworth and Gordon, 1985).

For advisory purposes, k₁ is assumed to be constant at 0.62 (MAFF, 1975). The work of Barber <u>et al</u> (1980) and Barber <u>et al</u> (1983) suggested that the use of this constant figure may be unjustified. These workers found that over-prediction of milk yield was detected when a new high ME value was calculated using a new prediction equation (Barber <u>et</u> <u>al</u>, 1984) based on a TDM basis. They concluded that the cause of this biased prediction of animal performance may be due to the fact that silage ME is utilised less efficiently than what it is thought to be.

1.1.5 The Use of Digestibility to Calculate ME.

It is possible ME to calculate from digestibility measurements and this approach has been widely practised by most advisory organisations in the UK. In this case, ME is calculated from the multiplication of DOMD% by a constant factor. This factor is based on the assumption that the energy value of the DOM is relatively constant between foods (Switt, 1957), together with a constant ratio of ME/DE (Blaxter, 1964).

MAFF (1975) suggested the use of a multiplication factor of 0.154 for various foodstuffs (19.0 MJ kg⁻¹DOM * 0.81). However, this factor is unlikely to hold for silages because of the presence of high energy volatile compounds. Indeed, several investigators suggested the use of higher coefficients to account for these volatiles. Thomas and Chamberlain (1982) stated the use of a multiplication factor of 0.163 (19.35 MJ kg⁻¹DOM * 0.84) for well preserved high digestibility silages. In a comprehensive study reported by Givens et al (1989a), they calculated a factor of 0.161, however the GE of the DOM and ME/DE ratio were somewhat different than the previous study (19.9 MJ kg⁻¹DOM and 0.81 respectively). Based on the 48 RRI silages (RRI, 1975; Wainman et al, 1978 and 1984), a factor of 0.166 has been found (20.14 MJ kg⁻¹DOM * 0.826) which is higher than the factors previous workers (Givens, stated by the personal communication). We might observe that the ME/DE ratio for RRI silages is relatively higher (0.826) than the one reported by Givens et al (1989a) [0.81].This discrepancy is possibly related to the lower methane losses measured calorimetrically for RRI silages compared to those predicted by the Blaxter and Clapperton equation (1965) for Givens et al (1989a) silages.

This latter aspect has been confirmed by a cow study performed in Northern Ireland where an ME/DE ratio was found to be 0.84 (Unsworth, personal communication).

It is clear from the previous findings that the calculated factor for grass silages is subject to considerable variation, an area which needs further study and investigation.

1.1.6 Summary

Silage is no longer considered as roughage feed which provides the animal with maintenance requirements only. The principles of making a good quality silage are now more fully understood. Considering the nature of the fermentation that takes place inside the silo, it is not surprising that silage is not an homogeneous entity. Therefore, a correct estimate of its nutritive value is of major concern to nutritionists and farmers.

The ME is the basis for allocating energy allowances for ruminants in the UK. Up to now, there has been a lack of accurate, fully determined ME values from measurements of faecal, urinary and methane energy losses. There is doubt about the applicability of Blaxter and Clapperton's equation to predict methane production, at least for silages and may be it is no better than the mean determined figure.

been known to have higher GE values Silages have compared to the original herbage. Also due to fermentation, silages contain а variable quantity of volatile substances. Taking these two aspects into account when calculating ME intake is preferred. However, practical considerations should not be ignored, namely the difficulties of measuring the GE of silages and also finding a suitable routine method which accounts for the volatiles. As a result of these difficulties, advisers have ignored these important factors.

The recent proposed ME prediction equations have been found to give a higher ME value than that calculated from equations in current use. This results from the inclusion of the extra energy of the volatile substances. А practical assessment of the new proposed regressions showed that their use will result in an over-prediction of animal performance when compared to the current prediction equations.

The fate of the volatile substances prior to consumption by the animal and also the the lack of justification for the use of the assigned efficiency values, could explain this bias prediction of animal performance.

The current situation about the efficiency of ME utilisation is not fully understood. The use of ARC (1980)to predict the efficiency values for silages may be unjustified. This is because firstly. none of these equations are based on silage diets and secondly, the

efficiency of ME utilisation is subject to numerous variations and it is not only related simply to metabolisability (q_m) . More research is needed in this area.

above uncertainties will prevent, in part, The the immediate adoption of route II as shown in Figure 1.4. The advisers will prefer an approach which is based on simple and sound practical advice than one which appears scientifically correct, but more complicated and not fully understood. Until these uncertainties are removed, route I will provide an alternative simple approach which is sound practically and more understandable. It is therefore important to find a good method for the prediction of silage OMD.

SECTION 2

PREDICTION OF THE ORGANIC MATTER DIGESTIBILITY OF GRASS SILAGE

1.2.1 Introduction

The nutritive value of forages cannot be considered as a single parameter, but it can be classified into three general components: digestibility, intake and the efficiency of utilisation of digested feed (Raymond, 1969). In this section, some related aspects of digestibility will be discussed and the methods used to predict this parameter will be reviewed.

1.2.2 Definitions

.

Some related definitions of digestibility which are used commonly are going to be mentioned here. These will include:

1 Dry matter digestibility (DMD) % =

DM consumed - DM excreted ----- x 100 DM consumed

2 Organic matter digestibility (OMD) % =

OM consumed - OM excreted OM consumed x 100

3 Digestible organic matter in the dry matter (DOMD)
or (D) value (%)=

OM consumed - OM excreted DM consumed x 100

OR

Digestible Organic Matter (DOM) consumed

This equation is used frequently for forages, particularly silages where soil contamination can occur during crop harvesting. Since this term is expressed on the DM consumed, it is always less than OMD by an amount depending on the extent of soil and mineral content of the herbage.

The relationship between the major measures of digestibility is shown in Figure 1.5.

1.2.3 The Measurement of Digestibility in vivo

Digestibility is considered to be a simple nutritional balance where a measurement of the amount of food eaten (or a particular nutrient) and the quantity of faeces produced from that food (or the particular nutrient in the faeces) is made over a period of time. The difference between the two quantities is the amount of food (or nutrient) which has been digested by the animal and when this is expressed on the amount of food (or nutrient) consumed, this is defined as the apparent coefficient of digestibility.

To measure digestibility, the experimental ration is fed in measured quantities for a long period in order to ensure that a steady state of faecal excretion is reached, and the faeces excreted are collected over a measured interval of time. A simple trial lasts 2-3 weeks and animals are kept in individual pens to facilitate the measurement of feed intake and faeces collection. For precise measurement, a preliminary test period of at least 10 days is recommended. The work of Blaxter et al (1956) showed that the necessary length of this period varied with type and level of feed, but it is usual for it to extend over 7-14 days and 10 days may be used in almost all cases.

The Relationship Between the Major Measures of Digestibility (after Osbourn, 1980)



40

.

1.2.4 The Importance of Digestibility as a Useful Index of Nutritive Value

Historically, digestibility trials were known as early as 1860 (Schneider and Flatt, 1975) and many feed evaluation units considered it to be an important part of their This parameter is regarded as the most useful programs. and basic index of nutritive value available at present. It is a reproducible characteristic and can be measured easily and accurately in many experimental stations around the world. However, digestibility cannot be regarded as criterion which describes nutritive value the only (Raymond, 1969).

Despite the fact that nutritive value is a product of three components, namely digestibility, intake and efficiency 1.2.1), most of the research has been more. (see successful in predicting digestibility of forages than the other two components. This is because digestibility offers considerable less animal variations as compared with intake and efficiency (see Table 1.7). Intake is largely influenced by the characteristics of the animal, therefore it is not surprising that intake can be predicted with less confidence as compared with digestibility. Castle (1982) indicated that for most of the trials conducted at the Hannah Research Institute, DOMD was the major factor affecting milk production when various silages were offered ad libitum. He concluded that on average, an increase of $10g \text{ kg}^{-1}$ in the DOMD concentration of the silage increased daily milk yield by 0.24 kg. Similar effects have been found for live weight gain (Wilkinson, 1985).

1.2.5 Factors Affecting the Digestibility Measurements

1 Species of Animal

Several investigators have commented about the use of either sheep or cattle when performing the digestibility trials (Schneider and Flatt, 1975).

Table 1.7

Approximate Relative Variation Due to Forages and Animals (Van Soest, 1982)

	Coefficient	of Variation (%)
	Forage	Animal
Digestibility	30	3
Intake	50	30
Efficiency*	50	20

* Use of digested energy for productive purposes.

Wainman (1977) and Aerts <u>et al</u> (1984) reported some differences in the digestive capacity between cows and sheep, however these differences may be small and it is questionable whether it bears a practical significance. Aerts <u>et al</u> (1984) found that there were no systematic differences in the digestive capacity of cows and sheep for 26 maize silage, 24 grass silage and 18 grass hay with an OMD of 65% or higher.

Cows appear to digest low quality rations better than sheep and on the contrary sheep appear to digest very high quality rations better than cows (Aerts <u>et al</u>, 1986). For medium quality rations, which form the major part of the rations for ruminants, the differences in the digestive capacity between sheep and cows are small and thus corrections may be unnecessary (Aerts et al, 1986).

Two possible reasons were given by the latter workers for the better digestion by cows. Firstly, low quality rations may have longer retention time in the rumen compared with sheep. Secondly, minerals better recycled via cows' are saliva therefore efficient microbial more activity is achieved.

Different breeds of sheep have not been found to have differences in their digestive capacity. A study by Blaxter <u>et al</u> (1966) at the Hannah Research Institute indicated that there were no differences in the digestive capacity between six breeds of sheep fed dried grass.

In conclusion, for dairy and beef cattle nutrition, it seems logic to use cows as the experimental animals. However, this proposition may be impractical since digestion trials with cows are expensive and difficult to operate. Therefore, for evaluation purposes a simpler and cheaper digestion trial using sheep appears to offer an acceptable alternative.

2 Level of Feeding

For diets containing forages, it is not surprising that as level of feeding increases, the rate of passage through the gut is also increased and accordingly, digestibility falls down (Wainman, 1977). Although this aspect seems to be simple and straight forward, the effect of feeding levels on digestibility is variable and complex and various interactions can occur between feeding level and other factors such as physiological status of the animal, diet composition, animal species and duration of adaptation to the diet (Aerts et al, 1986).

Several workers commented about the feeding level effect on digestibility. Brown (1966) reported that when dairy cows were fed on grain-hay diets (2:1 ratio) each increase in feeding level of one unit of maintenance, dry matter digestibility decreased 2.02 percentage units. More recently this has been Khidir and Vestergaard (1983) confirmed by EL where they found that digestible organic matter decreased when daily intake increased as a multiple of maintenance requirements. They concluded that 70% of the decline in digestibility of organic matter was caused by a decrease in the digestibility of the cell wall constituents.

For most evaluation programs in many parts of the world, digestibility trials are commonly carried out at maintenance level of intake since this is easier and the results obtained are reproducible and not liable to be influenced by the level of feeding.

3 Associative Effects

For livestock production, it is commonly practiced that suitable concentrates are fed along with forages so that high levels of production are maintained.

In this situation, the influence of one food on the digestibility of another might occur. However, this influence should be considered together with the term "balanced ration" where all nutrients are quantities which do not limit present in the utilisation of other nutrients (Wainman, 1977). For example, the digestion of cellulose can be decreased if there is a lack of nitrogenous substances or essential minerals in the ration. In addition, the existence of large proportions of easily digested such as starch carbohydrates and molasses can digestibility of depress the the fibre. because the simpler carbohydrates micro-organisms attack first rather than attacking the fibre constituents (Schneider and Flatt, 1975).

Various experiments have been cited which suggest no associative effects between feeds, while some other experiments indicated the opposite. A study performed at RRI Feed Evaluation Unit (Wainman et 1976) failed to show significant associative al, effects when oats, barley and wheat were fed in differing ratios with silage and dried forages. This latter work has been confirmed lately by Wainman et al (1979) when maize silage and barley were fed together. However, American workers indicated the contrary where Byers et al (1976) reported negative associative effects when corn silage and corn were Wainman et al (1979) attempted to fed together. explain this discrepancy and attributed the difference between the two experiments to the physical form in which the grains were fed and in

the nature of the starch contents.

4 Individual Variations

has been done on the effect of Little work individual variations on digestibility measurements. However, in а study based on 1328 individual digestion measurements with 79 different sheep and individual digestion with 24 518 measurements different cows, Aerts et al (1986) reported that the significant difference between individual animals appears to be small. However, they stressed the importance of using 4 animals so that reproducible and reliable digestibility measurements can be obtained. They concluded that digestibility can be determined with а standard error of 1 - 2digestibility units when using 4 animals.

1.2.6 Laboratory Methods for Predicting the Organic Matter Digestibility of Grass Silages

1.2.6.1 Introduction

There is no doubt that the measurement of digestibility using animals is the ultimate standard in determining nutritive value of forages. However, for practical feed evaluation, this proposition may not be feasible since this approach is expensive, labour intensive and cannot be used on a large number of samples. As a result, simple and economical laboratory methods which can be used to predict this parameter are an absolute necessity. In the UK, the search for such method(s) has been a major advisory concern of the different organisations, in particular the two feed evaluation units established in England and Scotland. Recently, notable attempts at developing prediction equations for OMD of grass silages were reported by Aerts et al (1977), Barber et al (1984) and Givens et al (1989a).

1.2.6.2 <u>Criteria for Selection of Laboratory Methods</u> The need for laboratory methods to predict <u>in vivo</u> digestibility of forages is well recognised. Various methods have been suggested in the literature and before any method(s) can be selected, some basic principles should be considered. These principles include:

- 1 The analytical procedures involved must be capable of being performed routinely on large numbers of samples with reasonable speed. Its execution should be reproducible, economical and not require a high technical skill.
- 2 The method(s) should have a high predictability of the standard in vivo, maximise R^2 and minimise prediction error.
- 3 The method(s) must be versatile, that is the equations needed for different feed types should be as few as possible.
- 4 The method(s) must be transferable between laboratories and should need minimal change and updating.

1.2.6.3 Review of the Methods

The predictive methods which are most widely used for evaluating forages may be categorised as:

- I Chemical methods, which measure some chemical parameter thought to be related in some way to the digestibility of the sample;
- II Biological methods, which attempt in some way to mimic the natural processes of digestion and

III A physical method (Near Infrared Reflectance, NIR), which makes more fundamental, non-destructive, measurements, related to the energies absorbed from the incident radiation by molecular groups in the sample.

1.2.6.3.1 Chemical Methods.

For more than 130 years, the Weende system of Henneberg and Stohmann has been used extensively to describe chemical composition of feed in many parts of the world.

This system involves the following procedures:

- (1) Oven drying at 100°C for moisture determination.
- (2) The dried residue is ether extracted to determine lipids.
- (3) The remaining residue is refluxed 30 min with
 1.25% sulphuric acid followed by 30 min with
 1.25% sodium hydroxide. The insoluble organic matter residues are reported as crude fibre.
- (4) Nitrogen and ash are determined in separate samples.
- (5) Nitrogen-free extract is calculated as:
 NFE = 100 Ether extract Crude Fibre Ash Crude Protein (N * 6.25).

This system is based on the assumption that crude fibre represents the indigestible part of the feed. For this reason, therefore, the early evaluation systems were based on regression equations relating digestibility to crude fibre (see Kivimae, 1960; Alderman <u>et al</u>, 1971). The latter assumption has been found to be incorrect, in particular for ruminants since these animals have the capability to digest a large part of the crude fibre.

Many attempts have been made to replace the Weende system with a more fundamental approach and the detergent system of Van Soest (1967) is such an approach.

This system divides the forage organic matter into two main fractions: 1 - Cell contents which are soluble in neutral-detergent and 2 - Cell wall constituents which are insoluble in neutral detergent. Table 1.8 illustrates the division of forage organic matter using the detergent system and Table 1.9 shows the basic scheme of analysis.

introduction of the detergent system, many Since the workers have used its parameters to predict the nutritive An extensive review by Barnes (1973) value of forages. documented various experiments in which neutral detergent fibre (NDF), acid detergent fibre (ADF) and acid detergent lignin (ADL) were used to predict in vivo digestibility of various forages. O'Shea et al (1972) found that ADF was a poor predictor of in vivo OMD of 31 grass silages ($R^2 = 0.05$). However, more recently Givens et al (1989a) reported that NDF and ADF were better predictors of in vivo OMD of 86 grass silages (R^4 , SEP were -0.53, 4.5% and -0.37, 5.2% respectively) than crude fibre (R^2 = -0.33 and SEP = 5.2%).

In an attempt to improve the predictive capacity of ADF, Clancy and Wilson (1966) studied the effect of increasing acid strength and the length of boiling time from 1 to 3 hours. Two hours hydrolysis has been found to increase the R^2 of 131 herbage samples from -0.49 to -0.79 and decreased the standard deviation from \pm 5.6 to \pm 3.6. The use of this procedure includes oven-drying at 95°C as a preliminary step. This treatment prevents the use of acid detergent as a means of assaying for heat damaged and unavailable protein, which is one of the most valuable applications of ADF (Van Soest, 1982).

The method of Clancy and Wilson (1966) is well known as modified acid detergent fibre (MADF) and has been used
TAULE LO

Division of Forage Organic Matter by System of Analysis Using

Detergents (Van Soest, 1967)

	ly digestible	stible accordi
Availability	Almost complete not lignified	Partially diges the degree of
Components	Lipids Sugars, organic acids and water- soluble matter Pectin, starch Non-protein N Soluble protein	Hemicellulose Fibre-bound protein Cellulose Lignin Lignified N
Fraction	Cell contents (soluble in neutral-detergent)	<pre>Cell wall constituents (fibre insoluble in neutral-detergent) l Soluble in acid-detergent 2 Acid-detergent fibre (ADF)</pre>

.

А

Table 1.9

Basic Scheme of Forage Analysis Using Detergents (Van Soest, 1982)

Fraction	Reagent	Treatment	Yield
Neutral-detergent residue (NDF)	Na lauryl Sulfate, EDTA pH 7.0	Boil l h	Plant cell wall less pectins
Acid-detergent fiber (ADF)	Cetyl trimethlyammonium bromide in 1 N H ₂ SO ₄	Boil I h	Lignocellulose + insoluble mineral
Unavailable nitrogen	Acid-detergent	Kjeldahl N deter- mination on ADF	Maillard products and ligni- fied N
Lignin	72% H ₂ SO ₄ treatment on ADF	3 h , 20°C	Crude Lignin
Cellulose	None	Ash residue from lignin step	Loss in weight
Silica (SiO ₂)	Conc. HBr (48%) treat- ment of ADF ash	Treat ash dropwise 1 h, 25°C	Residue is SiO ₂
Hemicellulose	None	Calculate as NDF-AD	Έι

.

extensively for many years to predict nutrient value of various forages in many laboratories in the UK. Its usage was more pronounced at the ADAS laboratories and has been considered the official method of calculating the forages (see Barber et al, 1984). value of energy However, more recently, the weaknesses of this method were evident and suspicions about its capability to predict in vivo digestibility, for silages in particular, started to arise (Givens, 1986).

Among the fibre methods, Lignin techniques were the most methods promising chemical in predicting in vivo Lignin digestibility of forages. is а completely indigestible substance known to be the main factor limiting digestibility of forages. The manner in which Lignin reduces digestibility is not fully understood, even though several theories have been proposed. Lignin is believed to link carbohydrates and make them indigestible by reducing their susceptibility to enzymatic attack. This is supported by the evidence obtained from alkali-treated straws where ester linkages between Lignin and carbohydrate are easily cleaved by alkali (Van Soest, Another theory was that Lignin accumulates at the 1982). surface of plant cells and therefore protecting the underlying polysaccharide from microbial attack (RRI, 1983).

have used Lignin predict in vivo Manv workers to For 18 grasses and legumes mixtures, Van digestibility. Soest (1963) found an R^2 of 0.61 between in vivo OMD and total Lignin when the ADF fraction was extracted by 72% sulphuric acid. Aerts et al (1977) compared the ability of seven laboratory methods to predict in vivo OMD of 56 He found an R^2 of 0.77 between the Van Soest silages. and Wine (1968) Lignin method and in vivo OMD with a coefficient of variation (CV) of 5.6%.

Morrison (1972) dissolved the cell wall fraction in 25% bromide solution and measured Lignin acetyl spectrophotometrically at 280 nm. Subsequently, Morrison (1973) used his method to predict in vivo DOMD of 20 grass silages and reported an R^2 of 0.71. Recently. Givens et al (1989a) quoted an R^2 of 0.63 with an SEP of 3.7% when they used the acetyl bromide lignin method to predict the in vivo OMD of 99 grass silages. They concluded that this method performed considerably better than any other chemical methods tested in their work.

The Morrison (1972) method has been used routinely by NOSCA to predict <u>in vivo</u> DOMD of grass silages and subsequently to calculate the energy value for ruminants (Murray, 1986).

Despite the fact that Lignin was found to be a better predictor of <u>in vivo</u> digestibility of forages than any other chemical methods, it has not been used widely in practical feed evaluations. Several possible reasons can be attributed to this:

1 Various different methods of Lignin analysis have been used which suggest that there is no agreed and standard procedure to determine the Lignin content of forages. This fact has been studied by McLeod and Minson (1971) when they compared the magnitude of the error in predicting in vivo DOMD from four different methods of analysis for Lignin. They found that the error of prediction was greatly variable between methods and reported to be from + 3.1 to + 5.0. More recently, Givens et al (1989a) found a striking contrast between the prediction power of acetyl bromide Lignin and permanganate Lignin when these methods were used to predict in vivo OMD of 99 and 86 grass silage samples respectively.

The R^2 and SEP for each method were found to be 0.63, 3.7% and 0.22, 5.7% respectively.

- 2 The differences among species of plants in Lignin content are variable. The most striking difference is between grasses and legumes where the latter relatively higher amounts of Lignin than have grasses of the same digestibility (Van Soest, 1982). species differences complicate the use of the These Lignin's prediction equations where the equation for Legumes does not fit either grass data or data for mixed samples. This aspect was confirmed by Minson (1982) when he showed large differences in the relationship between Lignin content and digestibility for a variety of forage species and particularly between grasses and legumes.
- 3 The methods of Lignin analysis are complicated, tedious and require skillful labour to operate. It requires a very small amount of the sample which means sampling is a very important consideration, otherwise precision might be affected.

The inclusion of a Lignin analysis with other prediction methods in a multivariate relationship usually improves the predictive power of the regression equation. This is not surprising since we know the role of Lignin in forage digestibility.

Kridis <u>et al</u> (1987) revealed that the combination of IVOMD plus acetyl bromide lignin method in a bivariate relationship improved the prediction power of <u>in vivo</u> OMD of 72 grass silages from an R^2 of 0.74 and RSD of 3.4% to an R^2 of 0.80 and RSD of 3.0%. This aspect has been confirmed in a similar study by Givens et al (1989a).

1.2.6.3.2 Biological Methods.

For the purposes of this review, these methods will be classified into three main groups:

- 1 In vitro rumen liquor methods
- 2 Enzymic methods
- 3 In situ method

1 In vitro rumen liquor methods

As the name implies, these methods involve the use of live rumen micro-organisms to digest a known amount of forage samples anaerobically in a test tube under a controlled environment of temperature and pH for a specific period of time.

The known and unknown factors that might affect the disappearance of the incubated samples will be subjected to rumen micro-organisms. This is in contrast to the chemical methods where the analyst analysed the plant cell wall in an attempt to identify a chemical functional group(s) which might be related to digestibility. Therefore, it is not surprising that the rumen liquor method will perform better in predicting in vivo digestibility of forages than the chemical methods. Van Soest (1982)emphasised that the superiority of the in vitro method over the chemical methods relies on its ability to reflect biological events.

The use of these methods was early reported as a one-stage process. Asplund et al (1958) predicted in vivo DMD digestibility of 11 hays with an R^2 of 0.56 3.5% and RSD of when using 24hours Subsequently, Tilley et al (1960) incubation time. found that in vivo DMD of 20 cocksfoot samples can be predicted with an R^2 of 0.83 and RSD of 3.6%

when using 48 hours incubation.

The addition of a 48 hours second-stage with acidpepsin improved considerably the coefficient of determination lowered the standard error and compared to results obtained from the one-stage (1960)technique. Tillev et al reported an improvement in in vivo DMD of 20 cocksfoot from an $\mathbf{R}^{\mathbf{Z}}$ of 0.83 and RSD of 3.6% using one-stage process R² of 0.96 and RSD of 2.0% using two-stage to process.

This method has been universally known as <u>in vitro</u> two-stage technique and found to be superior as a predictor of <u>in vivo</u> digestibility over the chemical methods (Tilley and Terry, 1963).

The rumen micro-organisms in the first stage attempts to measure the digestible fibrous fraction of the plant cell wall while the acid-pepsin in the second stage solubilises the microbial dry matter and nitrogen resulting from the first stage. Therefore, the entire system is rather a simulation of the digestive process that occurs inside the ruminant animal. Van Soest et al (1966) replaced the acid-pepsin stage by neutral detergent which solubilises bacterial cell wall and other indigenous materials. They reported such systems as in vitro true digestibility rather than the apparent digestibility of Tilley and Terry. Even though the Van Soest system is two days shorter, it has not been found more accurate than the original method (Cottyn et al, 1986).

Several modifications to the Tilley and Terry (1963) original procedure were proposed. Alexander and McGowan (1966) noted that the centrifugation at the

end of each stage was not suitable to apply this method for routine use. Therefore, they suggested direct acidification at the end of the first stage followed by immediate filtration at the end of the second stage. By doing so, Alexander and McGowan (1966) were able to determine 250-300 samples per week without loss of accuracy.

To improve the accuracy of prediction of in vivo silages, Alexander and digestibility of McGowan (1969) proposed a scheme of analysis where fresh homogenised silage samples can be used instead of the dried samples so that volatile compounds usually silage found in can be included in the determinations. This scheme improved the prediction of in vivo OMD of 18 grass silages from an $R^2 = 0.27$ to 0.74 and lower RSD from 3.6% to 2.2%. The latter approach was adopted at the West of Scotland College as the routine method to evaluate silages.

The Tilley and Terry (1963) method has been cited as the best and the most accurate technique of predicting in vivo digestibility of forages over a variety of other techniques (Oh et al, 1966; McLeod and Minson, 1978; Abrams et al, 1981; Morgan and Stakelum, 1987; Coelho et al, 1988). For silages, limited work has been reported, however the work of Aerts et al (1977), Barber et al (1984) and Givens et al (1989a) were the most notable. A11 these researchers showed an accurate prediction of in vivo OMD of 56, 56 and 117 grass silages respectively, when using Tilley the and Terrv (1963) method.

Being a biological system, the <u>in vitro</u> rumen liquor method is liable to many sources of error which

The make this technique difficult to standardise. most important single source of error is the rumen Several factors were found to influence the liquor. rumen liquor activity. These include the diet of the donor animal, method of collection, sampling, and processing of the rumen liquor (Barnes, 1973). The activity of rumen liquor collected from cattle or sheep has also been found to be different (Van Dyne, 1962). Other sources of variations like grinding procedures, maintaining adequate pH, sample size, temperature, adequate anaerobiosis and length of incubation time have been found to be important. obtain satisfactory results (Barnes, 1973). to Alexander and McGowan (1966) included standard samples in each in vitro run so that between-run drifts can be corrected for.

Despite the fact that the in vitro rumen liquor substantially the method improved of accuracy predicting in vivo digestibility, this method has not been widely accepted to evaluate forages routinely. Beside the uncontrolled variations mentioned above, difficulties of this technique major are worth mentioning. These include its dependence on fistulated animals, expensive, slow and difficult to operate. More importantly, the objection to the use of fistulated animals by the animal welfare organisations is strong. These reasons will undoubtedly make the using of this method limited and it is likely to decline in the near future. These difficulties have led to a search for more simple methods which can replace the rumen liquor, and the use of the cellulase enzyme may offer an acceptable alternative.

2 Enzymic methods

These methods are empirical procedures that

solubilise forage dry matter with little attempt to the actual chemical involved groups (a define distinction from the chemical methods) as long as results are correlated with the in vivo the parameters (Barnes, 1973). They involve the use of enzymes at moderate temperature.

Commercially produced cellulase from fungi were widely used for this purpose. Jones and Hayward (1975) tested 4 different fungal cellulases in their ability to solubilise forage and cellulose paper. They found that the Trichoderma viride cellulase was the most active in solubilising both substrates. Accordingly, this source of enzyme was used subsequently by other workers (Terry et al, 1978; Adamson and Terry, 1980; DeBoever et al, 1988). It is now the most widely accepted source and has been used successfully in many laboratories around the world.

Donefer et al (1963) were the first to examine the use of cellulase preparations to predict forage digestibility. Subsequent work by Jarrige et al (1970), Jones and Hayward (1973), Hartley et al (1974) and McQueen and Van Soest (1975), all reported а high correlation between in vivo digestibility and the solubility of forage dry or organic matter with cellulase preparations.

A second stage with acid pepsin has been proposed to follow the cellulase stage in an attempt to improve the predictability of the method (Jones and Hayward, 1973). However, these authors found that this procedure did not improve the correlation with <u>in vivo</u> digestibility. Jones and Hayward (1975) noted that the pretreatment of forage samples with acid pepsin for 24 hours followed by 48 hours cellulase incubation considerably improved the prediction of <u>in vivo</u> digestibility of 19 grasses. This has been confirmed subsequently by Goto and Minson (1977) and McLeod and Minson (1978). This improvement has been suggested as being due to not only the removal of protein but also the forage cell wall became more accessible to cellulase attack (Jones and Hayward, 1975).

Limited work has been reported about the use of the cellulase method to predict in vivo pepsin digestibility of grass silages. The work of DeBoever et al (1988) and Givens et al (1989a) were recent and comprehensive studies the most The latter workers reported an R^2 of available. 0.71 and RSD of 3.2% for 55 grass silages.

Several workers attempted to replace the acid-pepsin stage with neutral detergent (Van Soest and Wine, 1967) so that the time required could be reduced from 24 hours to 1 hour and also forage cell walls could be prepared for cellulase attack. Hartley et al (1974), Roughan and Holland (1977) and Dowman and Collins (1982) all reported high correlation between this method and in vivo digestibility of forages. Jones (1986) criticised the pretreatment with neutral detergent and concluded that while the time required to complete the analysis was greatly reduced when compared with the pretreatment with acid pepsin, it was difficult to remove the traces of detergent from the residues. He further added that the neutral detergent acts as a potent inhibitor of cellulase, therefore, a prolonged and effective washing is required.

McLeod and Minson (1982) compared the effects of the pretreatment of acid-pepsin or neutral detergent

upon the accuracy of predicting in vivo DMD of grasses and legumes by the cellulase method. Thev concluded that, for grass samples, the gain in reducing analytical time with neutral detergent is larger error in predicting in vivo. offset by a This larger error is confirmed by digestibility. Givens et al (1989a) where he reported that the use of the pepsin-cellulase method is relatively more the neutral detergent _ accurate than cellulase method for grass silages.

safelv It can be suggested that the cellulase techniques may offer another alternative to the rumen fermentation techniques (Jones, 1986). The cellulase techniques have been found to be precise and reproducible and the replicates of determinations were significantly lower than the rumen liquor Beard, 1977). method (Clark and For these pepsin-cellulase reasons. method has the been proposed to be used routinely as a predictor of in vivo digestibility of hays (Adamson and Terry, 1980) and grass silages (Givens et al, 1989a). However, Adamson and Terry (1980) stressed the importance of checking the activity of the cellulase batches since they found a significant difference between batches at 0.1% level. This confirmed earlier findings by Clark and Beard (1977). Close activity monitoring of the of cellulase between batches and from different suppliers is therefore required to ensure constant enzyme activity.

Terry <u>et al</u> (1978) compared the accuracy of rumen liquor and pepsin-cellulase methods in predicting the <u>in vivo</u> DMD of 48 grasses and 25 legumes. They showed that one regression equation can be used for both grasses and legumes when using the rumen liquor method whereas the pepsin-cellulase method failed to satisfy this criteria. They further added that the pepsin-cellulase can be of value for grasses and, within species, for legumes because of its speed, precision, accuracy and convenience when compared to the rumen liquor method. These latter findings may raise suspicion about the applicability of the pepsin-cellulase method for practical evaluation of forages of mixed species.

3 In situ method

The in situ method involves placing small amounts of feed inside bags made of indigestible fabric (such as silk, dacron, nylon or polyester) and suspended in the rumen of a fistulated animal for specific incubation periods. The technique measures the disappearance of feed dry matter or specific nutrients from the bags. This is in contrast with the previous technique where the incubator was the test tube.

Quinn <u>et al</u> (1938) were the first to suspend silk bags in the rumen of fistulated sheep and study the disappearance of feeds. McAnally (1942) suspended silk squares in fistulated sheep and studied the disappearance of wheat and oat straw. Balch and Johnson (1950) placed cotton threads in the rumen of fistulated cows and observed the rate of the thread's breakdown.

Since these initial studies, the in situ method has been used subsequently to enhance our understanding of forage degradation and rumen functions. Erwin and Elliston (1959)measured the rate of disappearance of concentrate and hays from nylon bags. The rate of digestion of the cellulose in four forages was measured by Hopson et al (1963) using dacron bags. Chenost et al (1970) characterised the

dry matter disappearance of 84 samples of leaves and stems of 9 forages using nylon bags. Shoeman et al (1972) used the in situ method to study the formaldehyde the degradation of effect of on different protein samples. et al (1978) Ørskov studied the effect of diets supplemented with fat on Rook et al (1983) determined the fibre digestion. rate and extent to which various minerals were released from polyester bags containing grass silages incubated in the rumen of cows. Silva and Ørskov (1988) used the in situ method to study the effect of five different supplements on the degradation of straw in sheep given untreated barley straw.

The in situ method was also proven to be useful in forage evaluation. Lusk et al (1962) reported an R^2 of 0.69 between 72 hours dry matter disappearance from nylon bags and in vivo DMD of 8 hays. Chenost et al (1970) found a highly significant correlation between in vivo digestibility hours rumen disappearance from various and 48 forages including silages. Aerts et al (1977) compared the ability of seven laboratory methods in predicting in vivo OMD of 56 grass silages. Thev concluded that the in situ technique predicted in vivo digestibility more accurately than the in vitro rumen liquor or chemical methods. The most distinguished study about the use of this method to evaluate grass silages was reported by Cottyn et al (1986) at the National Institute of Animal Nutrition in Belgium. They revealed that the nylon bag method can predict in vivo OMD of grass silages better than in vitro rumen liquor methods. For 100 grass silages, they reported an R^2 = 0.85 and CV = 3.2%.

results subject to Variability of in situ were to improve its considerable study and attempts reliability were suggested. The length of incubation time in the rumen could affect the results obtained. of the data suggests that dry matter Most disappearance from bags tends to increase rapidly to a point, then reaches a plateau with little increase thereafter.

Van Keuren and Heinemann (1962) found a significant increase in dry matter disappearance of various forages from the bags with each 24 hour increase in length of incubation time. In the same study, larger variations between bags were found associated with short incubation times. These variations diminished as the incubation times increased up to 72 hours. This has been confirmed by Mehrez and Ørskov (1977) who recorded a high variability between bags when rolled barley was incubated in nylon bags for 6, 9, 12, 15, 18 and 24 hours. Chenost et al (1970) showed that the in vivo OMD of various forages was better correlated with 48 hours than with 12 and 24 hours incubation time. This time has been found to be adequate for the complete digestion of forages in the rumen.

The diet of the fistulated animals can influence the rate of the dry matter disappearance from the bags. Ørskov et al (1980) stressed this effect and indicated that a high concentrate diet could influence the cellulolytic activity in the rumen. Lindberg (1981b) showed that the dry matter disappearance from bags containing hay samples was significantly decreased when the amount of cereals increased in the basal diet. They concluded that this decrease in dry matter disappearance was due to changes in rumen micro-organisms from fibre digesting organisms

to amylolytic and saccharolytic organisms. These findings were in agreement with previous reports (Van Keuren and Heinemann, 1962).

pore size is an important factor The bags in regulating the flow of the rumen content out of and Lindberg and Knutsson (1981) showed into the bags. that when grass hays were incubated in nylon bags, significant difference in the loss a highly of particulate dry matter between bags with 10 and 20 um pore size occurred. The use of bag pore sizes of 20 and 35 um were found to limit gas release from the bags (Uden et al, 1974). However, Lindberg and Varvikko (1982) reported that the bag pore size had little effect on dry matter disappearance if the sample was incubated long effective washing of enough. An bags between experiments is recommended so that the pores are cleared out (Ørskov et al, 1980).

The preparation of samples before incubation in the rumen could have an effect on the in situ results. Ideally, extrusa collected from the oesophagus of the animal can be used (Playne et al, 1978). However. this approach may be impractical. Alternatively, a laboratory hammermill fitted with different screen sizes has been used. However, it is recommended that one screen size is used so that consistent results can be obtained. Grinding forage samples through 20-, 40- or 60-mesh screen was found to have little effect on the dry matter disappearance (Van Keuren and Heinemann, 1962). Similar findings were reported by Lindberg (1981a) when silage samples were milled to pass through 1 and 4.5 mm screens. They concluded that fineness of grind little effect has on drv matter disappearance during longer incubation times.

However, the loss of particulate matter was more than doubled using a 1-mm screen compared to the (Lindberg and 4.5-mm screen Knutsson. 1981). Ørskov et al (1980) recommended the use of 2.5-3.0 screens for dry forages in the nylon bag mm studies. The use of fresh forage could eliminate the problem of grinding, however this is an area where further research is needed.

Ļ

Rapid fibre digestion can be obtained if the bags are incubated in the ventral rumen sac rather than the dorsal sac (Balch and Johnson. 1950). Rodriguez (1968) attached different weights to nylon bags and observed a significant increase in dry matter disappearance from bags as compared with bags which did not have attachments. However, the variation between bags was higher for the attached In the same study, he further found that the bags. variation between bags was reduced when they were tied to 50 cm of string rather than to 30 cm. He suggested that the longer string allowed free movement of the bags within the rumen and thus minimised variation.

It is necessary to incubate an adequate sample size inside the bags so that enough material can remain after incubation for subsequent analysis. However, this depends on bag size. Many workers showed a reduction in dry matter disappearance as the sample size, for a given bag size, was increased (Van Keuren and Heinemann, 1962; Mehrez and Ørskov, 1977; Lindberg, 1981a). Ørskov (1980) recommended a sample size of 2 g air dry ground straw, 3 g good hay or dried grass, 5 g of concentrate and 10-15 g of fresh herbage for a bag size of 140 x 90 mm.

major source of variation which could influence А the in situ results may be the washing procedure after bags are removed from the rumen. However, few reports have been cited to study this effect (Chenost et al, 1970; McManus et al, 1972; Aerts et al, 1977; Cottyn et al, 1986; Kridis et al, 1989). Aerts et al (1977) and Cottyn et al (1986) commented about the superiority of the in situ method in predicting in vivo OMD of grass silages when bags were treated with acid-pepsin after removal from They concluded that the this pepsin rumen. treatment permits a more effective washing and leads to a better agreement between bags, probably by removing forage residues and adherent bacteria.

The results obtained from the in situ method should be treated with caution and Ørskov et al (1980) suggested three aspects which need to be considered about the use of this method. Firstly, the sample in the bag is not exposed to physical breakdowns due to mastication and rumination. Secondly, forage samples will be degraded to smaller particles until they become small enough to leave the bag through the pores. Thirdly, the in situ method measures the disappearance of the incubated sample from the bag not the complete breakdown to chemical Taking these reservations in compounds. mind. Ørskov and Mcdonald (1979) developed a technique to estimate the actual degradability within the rumen. They proposed an exponential expression of the form $p = a + b(1 - e^{-Ct})$ where "p" is the actual degradability after time "t", "a" is zero hour degradation or the readily soluble materials, "b" represents the amount of sample which will be degraded within the rumen given sufficient time and "c" the rate constant for the degradation of "b". The combination of a + b gives the expression of

the potential degradability and represents the amount of sample which can be degraded within the rumen given sufficient time.

Ørskov and McDonald (1979) linked the degradation rate to an estimate of particle outflow rate from the rumen to give an expression of the effective degradability (P):

where a, b and c are the constants obtained from the previous equation and k is the outflow rate. This expression represents the amount of sample which will actually be degraded in the rumen and its value varies according to the outflow rate.

This approach has been used with some success at the Rowett Research Institute to study protein degradability.

From the foregoing discussion, it is clear that the in situ method suffers from inherent limitations which make it very difficult to standardise, and Donaldson et al (1980) attempted hence, reproduce. standardise this technique between to Scottish Agricultural Colleges, however considerable discrepancies reported. has were This been confirmed by a subsequent ring test involving 13 laboratories (Evans and Cottrill, 1984). As with the in vitro rumen liquor method, its dependence on fistulated animals major problem. is а These limitations will undoubtedly make this method less attractive for routine use. However, for initial forage evaluation and trying to establish calibration sets, the in situ method may prove to be a powerful technique.

1.2.6.3.3 A Physical Method. (NIR)

Near infrared reflectance is a non-destructive physical method capable of predicting the chemical composition of a solid sample rapidly with little sample preparation. After making the measurement, the sample can be retained intact for future or subsequent analysis. This technique is relatively new technology and has emerged in the last 10 years as a rapid method in forage testing.

The near infrared is that part of the electromagnetic spectrum which lies between the visible and infrared regions and covers the wavelength range between 800-2500 nm. For most quantitative analysis, the most useful region \mathbf{is} from 1200-2500 nm (Norris, 1985). The properties of the hydrogen bearing functional groups like -NH (Kaye, 1954) and their -CH. -OH and hydrogen bonding environment are considered to be useful to reveal the biological availability of energy and other compositional information. Any effect which could alter bond strength such aromaticity, as polar groups or hydrogen bonding to adjacent atoms in the sample, is going to change the spectral characteristics (Murray, 1986).

When the sample is submitted to the instrument a narrow beam of light strikes the sample and the reflected radiation from the surface of the sample is collected by detectors [photocells] (see Figure 1.6). The signals from the detectors are converted to a digital signal at defined and stored in a coupled computer intervals as the $\frac{1}{R}$). logarithm of reciprocal reflectance (Log Each Log represents a wavelength and the wavelength value interval can be from 0.1 to 10 nm, with 1 to 2 nm being the most selected interval (Norris, 1985). The computer performs stepwise multiple linear regression analysis to

FIGURE 1.6

Optical Geometry for Diffuse Reflectance Measurements based on Using O° illumination and 45° collection with Large Area Photocells (from Norris, 1985).



relate optical data to the calibration set at its disposal and then select wavelengths which best fit the component being predicted. The best selected prediction equation takes the following form:

Predicted Composition = $B_0 + B_1 W_1 + B_2 W_2 + B_3 W_3 + \dots$ $B_n W_n$

Where B_0 is the regression constant, B_1 to B_n are partial regression coefficients and W_1 to W_n are reflectances or derivatives of these at wavelengths $\lambda 1$, $\lambda 2$, ..., λn .

kinds of NIR commercial instrumentation There are 3 namely available, the monochromator, scanning filter instruments and discrete filter instruments (Shenk, 1986). The monochromator has the capability of scanning the entire NIR region of spectra and thus more spectral information can be collected compared to the other two The monochromator used in this project scans the types. NIR region between 1100 to 2500 nm and can generate 700 Log $\frac{1}{R}$ values (data points) at 2 nm intervals. On the other hand, the scanning filter instruments, and in particular the discrete filter instruments, have less NIR coverage and hence less data points. Therefore, it is not expected that the performance obtained from these instruments will match the performance of the monochromator. In forage analysis, it has been found that the scanning and the discrete filter instruments gave less satisfactory results when compared to the monochromator (Minson et al, 1983).

Before calculating regression equations, spectral data are subjected to mathematical transformations which were found to be useful in reducing noise due to factors unrelated to forage chemistry and the extraction of as much chemical information from the sample as possible (Westerhaus, 1985a).

Derivative techniques have been used to give a smoothing effect and sharpen the detail along the NIR spectral curve (Savitsky and Golay, 1964). Norris et al (1976) were the first to suggest that the use of this technique could improve the predictive ability of NIR when analysing forages by reducing the particle size effect which is a feature of ground forage samples. They found that the performance improved when using the second NIR derivative of Log $\frac{1}{R}$ values rather than Log $\frac{1}{R}$ values per They further examined the use of first and third se. derivatives, however no improvement was found over the second derivative. This technique has been used to transform forages spectral data with relative success by many workers (Barton and Burdick, 1981).

Recently. interest has been renewed in the use of alternative spectral data manipulation. The principal component analysis (PCA) is such a development. This technique reduces the number of variables by creating synthetic variable called "principal components" which are related to the original variable and then selecting the most significant wavelengths (Robert et al, 1986). Bv doing so, each principal component is independent of each other and therefore, reduces the possibility of intercorrelation between spectral data which might occur in the derivative method (Downey et al, 1987). Another feature of this approach is to permit a close investigation of each component and therefore allow chemical interpretation of the selected wavelengths (Downey et al, 1987). Robert et al (1986) compared the predictions of 87 in vivo grasses obtained by multiple linear regression of Log $\frac{1}{D}$ values and the PCA technique and found that the standard errors of calibration (SEC) and prediction (SEP) were not significantly different between the two techniques.

Calibration of NIR instruments is an important concept so that this technology can be applied successfully for

predictive purposes. The NIR technique cannot be expected to give satisfactory results beyond the range of available information at its disposal, therefore a careful calibration is an absolute requirement. This is a serious problem when it comes to a heterogeneous population, the grass silage population in the like, for example, whole of the UK. A calibration based defined on boundaries (ie: limited range in chemical, physical and composition) is not expected to botanical predict a population with undefined boundaries. Murray (1986) stressed the importance of this concept and indicated that the calibration samples should have a wide range and even distribution in composition, precise analytical data and typical properties of all future unknown samples. An adequate in vivo calibration set for grass silage should include for example, different species, harvests for different years, wilting time, stage of maturities, growing environments, methods of ensiling and different silage additives.

It is difficult to decide how many samples should be included in the calibration set, however this may depend on the heterogeneity of the population. Trying to predict animal response (for example, digestibility) may require a calibration set with large numbers of samples (Windham and Coleman, 1985). Very few studies have been reported about the number of samples recommended for calibration, however the work of Abrams et al (1987) may be the most useful. They studied the effect of increasing the number of calibration samples for various assays, including IVDMD, the standard error on of prediction. They observed that SEP was reduced as the number of samples in the calibration set was increased and concluded that a minimum of 100calibration samples were needed to minimise SEP and bias.

Sample selection is an important factor in successful NIR calibration. Three methods of data selection were namely; structured, random and spectral suggested, characteristics (Abrams, 1985). The first method depends on the available information, like stage of maturity, which can be used to describe the sample. This information is classified and then proportional sampling from each class This will ensure that each class is equally is selected. represented in the calibration set and thus bias will be avoided. Random selection is appropriate when there is no prior knowledge about the population. The third method requires obtaining the spectral data for all the samples intended to be included in the calibration set. The spectral characteristics are then grouped into clusters and random samples selected from each cluster are used to obtain the calibration set. This approach is scientifically sound because the samples selected represent the entire characteristics of spectral the population available. However, Abrams et al (1987) found that this method did not have any advantage over random selection as a means to select samples.

It is important to validate the equations generated by the multiple linear regressions so that a proper equation(s) can be selected to predict the unknown samples. The equation(s) which has minimum SEP, minimum bias and slope closest to unity should be selected (Westerhaus, 1985b; Murray, 1986).

Near infrared reflectance techniques have been used initially to predict protein and oil in grains (Hymowitz <u>et</u> <u>al</u>, 1974) and fat, protein and moisture in fresh meat (Kruggel <u>et al</u>, 1981). Norris <u>et al</u> (1976) were the first to use NIR to predict the nutritive value of forages. They reported an R^2 of 0.90, 0.78 and SEC of 2.64 and 3.58 for predicting IVDMD and <u>in vivo</u> DMD respectively, when 76 forage mixtures were analysed by monochromator. In this study, they used a multiple linear regression equation involving 9 terms. Barton and Burdick (1981) used monochromator to predict chemical and biological parameters for hay samples. They reported an SEC% and SEP% of 1.78 and 2.54 respectively for predicting in vivo Based on broad and large calibration sets, Abrams DMD. et al (1987) demonstrated the accuracy of this technique biological in predicting chemical and parameters of diverse hay samples. These results were obtained by using a monochromator and 9 terms calibration equations. Equations based on principal component analysis were successful to predict in vivo OMD of dried grasses with an R^2 of 0.95 and SEP% of 2.51 (Robert <u>et al</u>, 1986).

Few investigators used NIR to predict the nutritive value of grass silages. Using commercial farm silages, O'Keefe <u>et al</u> (1987) reported an accurate prediction of crude protein (CP) [SEP = 0.63%], but IVDMD was not predicted with sufficient accuracy (SEP = 2.96%). The low accuracy of predicting IVDMD could be due to the use of fixed filter instruments fitted with 19 filters only. Downey <u>et</u> <u>al</u> (1987) reported better accuracy when predicting IVDMD of grass silages using a monochromator (SEP = 2.7%).

The Use of NIR to Evaluate Silages in the UK

The need for economical, fast and accurate laboratory methods to predict the nutritive value of grass silage is well recognised by the advisory organisations in the UK. Since commercial NIR instruments became an accepted methodology in forage evaluation, recently ADAS, SAC and DANI adopted this technique in their silage evaluation programs as a means of rapid and, more importantly, economical method to predict nutritive value of grass silages. However, instead of direct calibration on <u>in vivo</u> digestibility, those laboratories have used NIR to predict MADF, LIGNIN and MADF respectively. Digestibility and ME are then calculated using appropriate regression equations. Two fundamental errors can be noted when using this indirect approach, namely, the error resulting from NIR prediction of these chemical parameters and then the error resulting from the use of these parameters to predict digestibility. These errors are additive which ultimately, reduces the accuracy of <u>in vivo</u> digestibility prediction.

In this thesis, it is intended to investigate this problem and explore the possibility of calibrating NIR directly with a large <u>in vivo</u> digestibility data set. If this can be achieved, it would be appropriate to recommend this technique to be used by all the advisory laboratories in the UK in a more sound and fundamental way.

Conclusions

The near infrared technique is a fast and economical method and could provide some potential for silage evaluation. Provided that NIR is calibrated adequately, it might show a reliable alternative to the conventional Because the parameter intended to be predicted methods. is determined in a multivariate way, this could lead to accurate prediction greater than any traditional method. Simplicity is another attractive feature of this technique, since little sample preparation is required other than simple grinding and drying.

Given these features however, the NIR does pose some problems. The cost of the instrument are relatively high, and the calibration procedures and subsequent equation selection and testing demands a skillful and experienced with access to powerful computers. personnel The development of a calibration set is sometimes expensive, in particular with biological parameters, for example digestibility. In addition, each calibration is specific for the component to be predicted, therefore several

calibrations are required for other components. Spectral data treatments are rather complicated and need to be standardised among users. The noise level is a serious problem of NIR and the need for constant monitoring and instrument to control this noise is updating of the required. The NIR instruments are sensitive to temperature and humidity and these must be maintained at the recommended level to ensure adequate performance.

1.2.7 Summary

Reliable silage evaluation is an important program in providing satisfactory farm advice and conducting adequate Digestibility is an important aspect of nutritive research. value and its determination is essential for practical feed evaluation and ration formulation. Ideally, the use of animals to determine digestibility is the ultimate measurement. however, this approach is expensive and cannot be used on large numbers of samples. Therefore, a rapid, simple and comparatively inexpensive laboratory method to predict this parameter \mathbf{is} absolute an requirement.

The use of chemical methods to predict in vivo digestibility was the first to be suggested and speed, reproducibility and cheapness were the most attractive features of these techniques. Since these methods rely on pure statistical association between digestibility and chemical entities (cellulose, lignin), it should not be expected that these techniques will accurately describe this complex biological parameter. While these methods, in particular lignin methods, are of value between species of grasses, it is found to be inaccurate for a mixture of species, a feature found in real farm situations. This demands the need of separate regression equations for each species if these methods are to be implemented. However, this might be impractical for silage evaluation.

These inaccuracies led to a search for methods which might reflect biological reality and the rumen liquor and in situ techniques were such a development. The prediction superiority of these methods over the chemical methods may be due to the sensitivity of the rumen micro-organisms to the determined and undetermined factors which might affect digestibility. The in situ method benefits from a lesser variation due to rumen liquor activity when compared with in vitro methods, since bags are incubated in the rumen itself (Aerts et al, 1977). This feature along with its simplicity, makes the in situ method a powerful technique for the evaluation of forages knowledge and improving our of the processes of degradation which take place inside the rumen.

However, these methods were found to be difficult to More importantly, their use demands free standardise. access to fistulated animals and in many situations this might be impractical and difficult to achieve. The pressure from animal welfare organisations against the use of surgically prepared animals is apparent. These limitations make these techniques unfavourable and their applications for routine use is likely to diminish in the future.

The use of cellulase enzyme methods were suggested as a means of replacing the rumen liquor. These methods profit from their speed, precision and convenience relative to the in vitro rumen liquor technique. They are found to be accurate in predicting in vivo digestibility of individual species, however for a mixture of species it was not sufficiently accurate and in this case, the need for separate regression equations was found to be essential. The activity of the cellulase enzyme requires constant monitoring between batches and from different suppliers \mathbf{SO} that method's the accuracy can be maintained.

With increasing economic pressures to find an accurate and reliable method for predicting <u>in vivo</u> digestibility, NIR might be of potential value for routine silage evaluation. Provided that this method is calibrated properly, it might offer a reliable and acceptable alternative to the conventional techniques.

ŗ

CHAPTER TWO

60

-

.

MATERIAL AND METHODS

2.1 Silages Studied

One hundred and seventy dried samples of grass silage which had been evaluated <u>in vivo</u>, were collected from different sources around the UK. Table 2.1 shows the sources of the silage populations used in this work.

The distribution of <u>in vivo</u> OMD and chemical and biological predictors of the populations studied is shown in Table 2.2. The distribution of <u>in vivo</u> OMD for all silages studied is shown in Figure 2.1.

The composition of silages studied for all populations is shown in Table 2.3.

All silages were obtained from <u>in vivo</u> trials conducted during 1974 to 1986 at the different institutions shown in Table 2.1. The known characteristics of the silage used in this study are shown in Table 2.4. The <u>in vivo</u> determinations of OMD have been described in Table 2.5.

2.2 Analytical Methods

2.2.1 Dry Matter Base

Where analytical results from the centres were expressed on a toluene dry matter basis, they were converted to an oven dry matter basis according to the conversion factors described by Barber <u>et al</u> (1984). This is done because the toluene dry matter determinations for population 4 was not available.

A Analytical data such as proximates and fibres can be converted to an ODM basis using the following formula:

Chemical Parameter ODM = -----ODM/TDM

Table 2.1

.

Sources of Silages

Source	Population	Number	Drying Temperature (°C)	Mill Screen (mm)
ADAS Feed Evaluation Unit - Pre-1980		28	60	0.60
ADAS Feed Evaluation Unit - Post-1980	7	72	100	1.00
Rowett Research Institute	ς	43	100	1.00
North of Scotland College of Agriculture	4	27	08	1.00

-

.

Table 2.2

Distribution of in vivo OMD and Chemical and Biological Predictors of the Populations Studied (%)

		In vivo QMD (%)	MADF (%)	LIGA cm Lg-1	NCOMD (%)	PCOMD (%)	NB48 OMD (%)	IVOMD (%)
Pop 1 n = 28	Mean Range SD	70.3 59.2-78.1 5.2	33.3 25.4-43.7 4.6	2.58 1.88-3.17 0.340	75.5 59.1-84.9 6.9	59.4 51.4-70.8 5.6	- - -	68.6 57.6-73.7 4.0
Pop 2 n = 72	Mean Range SD	70.5 52.8-82.8 6.6	34.1 23.0-42.4 4.0	2.56 1.77-3.83 0.402	66.5 43.9-84.7 9.2	63.5 36.8–81.2 8.6	84.3 61.5-96.8 7.0	66.0 47.5-72.9 6.0
Pop 3 n = 43	Mean Range SD	72.4 60.9-81.5 4.5	37.8 30.9-45.3 3.4	3.00 2.40-3.50 0.247	61.6 48.5-73.2 7.8	60.1 45.5-73.4 7.9	86.8** 76.1-95.1 4.6	68.2 55.0-78.2 5.7
Pop 4 n = 27	Mean Range SD	70.1 55.2-86.7 7.1	36.5 31.7-44.1 3.2	2.60 1.80-3.38 0.412	67.9 53.8-79.8 6.3	54.0 44.6-64.3 6.5	85.3 73.8-95.4 5.7	66.9 56.4-76.2 6.8
TOTAL n = 170	Mean) Range SD	71.0 52.8-86.7 6.0	35.3 23.0-45.3 4.2	2.68 1.77-3.83 0.403	67.5* 43.9-84.9 8.8	60.5 36.8–81.2 8.3	85.0*** 61.7-96.8 6.3	67.1 47.5-78.2 5.7

* NCOMD data applies to 165 samples, 5 values were missing, two from population 1, one from population 2 and two from population 4.

** NB48 data applies to 24 samples only, since 19 samples contained insufficient material.

*** NB 48 data applies to 123 samples only, since population 1 was too finely milled for this method and in the case of population 3, 19 samples contained insufficient material.



Population number

[Populations								
		1		2		3		4	1	lotal
Composition	n	Values	n	Values	n	Values	n	Values	n	Values
ODN -1 Mean (gkg ⁻¹) Range SD	28	295 213-461 81.2	72	275 159-580 94.1	43	215 163-358 41.7	27	226 165-361 49.7	170	255 159-580 81.0
TUM -1 Mean (gkg ⁻¹) Range SD	28	321 235-489 80.4	72	299 183-585 89.5	43	238 175-392 45.2		NDA	143	285 175-585 82.9
Volatiles Mean as % of Range TDM SD	28	8.7 2.9-16.4 3.3	72	9.1 0.0-26.2 5.2	43	9.6 4.7-13.1 2.2		NDA	143	9.1 0.0-26.2 4.1
Ash - 1 Mean (gkg ⁻¹ DM) Kange SD	28	112 87-168 16.2	72	90.6 54-130 15.5	43	93 62-129 16.4	27	81.2 62-103 11.8	170	93.3 54-168 17.8
NDF Nean (%) Range SD	6	52.7 42.1-64.5 9.9	71	57.4 42.0-85.5 7.9	43	61.3 45.5-73.2 5.8		NDA	120	58.6 42-85.5 7.6
ADF Mean (%) Range SD	6	32.6 25.2~40.5 6.5	71	34.5 16.3-43.3 4.9	43	40.1 34.0-47.5 3.3		NDA	120	36.4 16.3-47. 5.2
pli Mean Range SD	28	3.9 3.2-5.0 0.45	71	4.3 3.4-6.0 0.59	43	4.2 3.6-4.9 0.34	27	3.9 3.6-5.3 0.34	169	4.2 3.2-6.0 0.50
CP Nean (gkg ⁻¹ DN)Range SD	28	188 115-342 50.3	72	170 93-259 40.3	43	162 108-223 24,5	27	131 93-185 25.4	170	165 93-342 40.3
Vol. N Nean as % of Range TN SD	28	10.4 6.1-18.8 3.2	65	$ 11.4 \\ 3.7-31.1 \\ 5.3 $	43	12.7 5.6-258 4.4	27	11.0 5.4-66.8 11.5	163	11.5 3.7-66.8 6.3
Lactate Nean (gkg DM)Range SD	28	63.7 8.9-132.6 32.5	70	69.5 7.9~203 44.1	43	61.3 17.7-104 24.2	8	66.0 2-150 52.6	149	65.8 2-203 37.5
Acetate Mean (gkg DM)Range SD	28	25.6 7+54.4 13.5	67	30.0 1.9-254 40.0		NUA		NDA	95	28.7 1.9-254 34.2
GE Nean (MJkg ⁻¹ DN)Range SD	28	20.8 19.1-23.1 1.0	72	21.0 15.4-25.7 1.5	43	21.3 19.6-22.0 0.67		NDA	143	21.0 15.4-25.7 1.2
ME - 1 Mean (NJkg-1DM)Range SD	28	11.7** 8.1-14.3 1.5	72	12.0** 6.1~15.3 1.7	43	12.6* 9.5-14.7 1.04		NDA	143	12.1 6.1-15.3 1.5

Table 2.3									
Composition	of	Silages	for	A11	Populations	Studied			

NDA * **

.

No Data Available Nethane losses are measured Methane losses are predicted from Blaxter and Clapperton equation (1965) Number of samples for which the analytical results are available. n
Table 2.4

Characterisation of the Silages Studied*

	Characteristics	Number of Silages
1	Method of Ensiling - Clamp - Big Bale	72 19
2	Cut Number - First Cut - Second Cut - Third Cut	74 31 7
3	Wilting Time (hours) - ≤ 12 - > 12-40 - > 40	26 72 18
4	Age of Ley (years) - ≤ 1 - > 1-5 - > 5 - Permanent Pasture	17 32 17 16
5	Additive Application - None - Add-F - Sylade + F100	54 40 26
6	Fertilisation (Kg N/ha) - ≤ 100 - > 100-200 - > 200	28 45 9

* Not all the 170 silages have known characteristics

86

Description	of	in	vivo	Trials	Performed	on	Silages	Used	in	this	Study
-------------	----	----	------	--------	-----------	----	---------	------	----	------	-------

Populations	Number of Sheep	Diet Fed With Grass Silage	Adaptation Period (days)	Collection Period (days)
1	4	None	10	10
2	4	None	10	10
3				
- first report	4	Oats + Barley	10	10
- second report	4	Wheat + Wheat Offals	10	10
- fourth report	4	Maize	10	10
4				
- Conn o lly (1984)	6	None	NS*	8
- Siddigi (1985)	5-6	None	10	8
- Helliwell (1986)	4	None	10	10

*NS = Not Stated

•

B Digestibility values such as <u>in vivo</u> DOMD can be converted to ODM basis using the following formula:

DOMD (ODM Basis)%=

Note that where the digestibility values were on OMD, they must be converted to DOMD first using:

DOMD% = OMD fraction x OM %

2.2.2 Oven Dry Matter Determination

A known quantity of sample was dried in a forced draught oven at the temperature indicated in Table 2.1. After a constant weight was reached, the dry matter was expressed as a percentage of fresh weight.

2.2.3 Ash Determination

Approximately 3g of dried sample was placed in a preweighed dry crucible and ashed in a muffle furnace at 500°C overnight. The crucibles were cooled in a dessicator then re-weighed and the ash content calculated.

2.2.4 Modified Acid Detergent Fibre

The MADF content of the samples was determined according to the method of Clancy and Wilson (1966).

1 g of dried silage was transferred into a beaker or flask. 100 ml of sulphuric acid - cetyltrimethylammonium bromide solution was added to the beaker or flask and then a condenser was fitted. The content was rapidly boiled and then gently boiled for 2 hours. The digest was then filtered hot through a weighed filter crucible using gentle suction. Excessive foaming in the filtrate was reduced by the addition of 1-2 drops of octane-2-ol. then washed with 3 portions The residue was of approximately 50 ml of almost boiling water and then with The crucible and content was then dried in an acetone. oven at 102°C overnight. The crucible was allowed to cool in a dessicator and then weighed. The increase in weight was multiplied by 1000 and the results were reported as g/kg of MADF in the sample DM.

2.2.5 Acetyl Bromide Lignin

The lignin content of the samples was determined according to the method of Morrison (1972).

The silage samples were dried at 80°C for 18 hours and then ground to pass through 0.8 mm screen. The dried sample (50 mg) of known ash content was heated with 20 ml distilled water at 70°C with occasional shaking. It was then filtered hot through a glass fibre filter paper GFA 2.5 (Whatman cm). The residue was washed thoroughly on the filter with hot water, 6 times with 5 ml ethanol, 4 times with 5 ml acetone and 4 times with 5 ml diethyl ether. The cell wall preparation and filter paper was transferred by tweezers to a glass stoppered test tube (Quickfit MF 24/3) and heated at 100°C for 15 minutes to remove traces of solvent.

Five ml of 25% acetyl bromide was then added to the tube and stoppered and heated for 30 minutes at 70°C. The tube and its contents were then cooled to room temperature. Five ml of the digest was taken and mixed 20 ml of acetic acid and then stoppered and with shaken. An aliquot of 1 ml added to 1.2 ml of 2.06% sodium acetate trihydrate in a 10 ml stoppered centrifuge tube, 7.5 ml of ethyl alcohol and 0.3 ml of 0.5 M hydroxyl ammonium chloride solution was also added and the tube stoppered and shaken. The tube was centrifuged

89

for 10 minutes at 1500 rpm and then was allowed to stand for one hour. The samples were read on a Cecil spectrophotometer using 10 mm manual flow cell at 280 nm. In every set, reagent blanks were included. Absorption value A was calculated from the equation:

> A = ODs - ODb C

Where:

ODs = optical density of sample ODb = optical density of blank C = concentration of dry OM in the final solution gl^{-1} .

2.2.6 Pepsin-Cellulase OMD Determination

The pepsin-cellulase OMD of the samples were determined according to the method of Jones and Hayward (1975).

The cellulase enzyme derived from fungi <u>Trichoderma</u> <u>viride</u> (BDH Ltd, Poole, Dorset, England) was used. Pepsin (1:10,000) was obtained from Zimmerman and Hoppes Ltd, Milton Keynes, England.

200 mg of the ground dried silage sample was mixed with 20 ml of acid pepsin solution (0.2% of pepsin in 0.1 M HCl) in a screw cap plastic tube. The tube was placed in a water bath at 40°C and incubated for 24 hours. The content of the tube was swirled occasionally. After incubation, the suspension was centrifuged (maximum speed for 5 minutes). By means of suction, the supernatant was removed as completely as possible leaving the residue inside the tube. Using a dispenser, 20 ml phosphate citrate buffer (pH 4.6) was then added to the tube to wash the remaining traces of the acid pepsin solution. The stopper was firmly placed and the residue was suspended by vigorous shaking. The suspension was

recentrifuged (maximum speed for 5 minutes). The supernatant was then removed by suction leaving the residue inside the tube.

Twenty ml of buffered cellulase solution (125 mg cellulase in 20 ml phosphate citrate buffer pH 4.6) was added to the tube. The content was thoroughly mixed and the tube was then placed in a water bath at 40°C and incubated for The content of the tube was swirled a further 48 hours. occasionally during this period. The residues were recovered at the end of 48 hours incubation in the presence of a filter-aid (hyflo supercel) by filtration through a fibreglass paper. The fibreglass papers and residues were dried at 100°C, cooled and weighed. They were then ignited at 480°C for 16 hours, cooled and weighed again. A parallel estimate of the total OM by ignition of 1 g of sample at 480°C for 16 hours and the inclusion of control tubes with buffer and cellulase only, allow the digestibility coefficients for the samples to be calculated:

Pepsin-Cellulase OMD% = Wt of OM in - OM of sample - OM of control Sample residue residue Wt of OM in Sample

2.2.7 In vitro OMD Determination

The IVOMD of silages were determined according to the method developed by Tilley and Terry (1963) and modified by Alexander and McGowan (1966).

Exactly 0.5 g DM of the ground sample was weighed in duplicate directly into glass tubes. One litre of rumen liquor was obtained from each of 3 fistulated sheep and then filtered as quickly as possible through 2 layers of

muslin, swept with CO₂ and was added to four times its volume of McDougall's buffer (1948) previously saturated with CO₂ to a pH of 6.9. One ml of molar ammonium sulphate solution per 50 ml of the rumen liquor buffer mixture was added. The sample inside the tube was inoculated by adding 50 ml of the rumen liquor/buffer mixture. The tube was swept with CO₂ and then capped with rubber bungs fitted with bunsen valves. The tubes were then placed in a water bath at 38.5°C. The pH of the digests were adjusted electrometrically to 6.9 after 24 hours of incubation. The digestion was carried out for a period of 48 hours during which a gentle swirling was performed from time to time. To terminate the 48 hours stage, 1.5 ml of 6 M HCl followed by 2.5 ml of 6 M HCl were injected into each tube. Five ml of aqueous pepsin containing 0.12 g 1:10,000 pepsin (Zimmerman and Hoppes, Milton Keynes, England) were injected into each tube after electrometric adjustment of the pH to 1.2. The content in each tube was then further incubated for 48 hours at 38.5°C during which the tubes were gently swirled from time to time. At the end of the pepsin digestion stage, the residues were recovered in the presence of filter-aid (hyflo supercel) by filtration through a fibreglass paper. The fibreglass papers and residues were dried at 100°C, cooled and weighed. They were then ignited at 480°C for 16 hours, cooled and weighed again. A parallel estimate of the total OM by ignition of 1 g of sample at 480°C for 16 hours and the inclusion of control tubes with buffer and rumen liquor only, allow the digestibility coefficients for the samples to be calculated:

IVOMD 🖁 =

Wt of OM in - OM of sample - OM of control Sample residue residue '----- x 100 Wt of OM in sample

2.2.8 In situ OMD Determination

2.2.8.1 The Preparation of Hay Samples.

One sample of timothy hay was used. The hay was first chopped twice to a length of 5-6 cm in a "GHL" straw chopper. Then it was mixed through a "Hobart" mixer slicer attachment 3 times until it reached a length of \leq 1 cm.

2.2.8.2 The Preparation of Grass Silage Samples.

The milled grass silage, samples were sieved through a 45 um sieve to remove fine particles. Approximately 5.5 g sample was weighed into each bag. The bags were incubated in batches of five at one time and then replicated in 3 sheep. Four bags represent 4 different silage samples and the fifth one was an internal standard (alfalfa nuts milled through 1 mm screen). This standard allowed adjustment for variation between rumens (between sheep and with time).

2.2.8.3 Animals and Their Diets.

Four mature 8 year old Suffolk wethers were used. They were housed indoors in loose pens and each fitted with a 55 mm rumen cannula with a screw top. Once the top was removed a bung of the correct diameter could be inserted to which were attached the 5 bags for incubation. Each sheep was fed a maintenance ration of 800 gd⁻¹ of meadow hay and 200 gd⁻¹ of Ewbol concentrate in two meals offered at 9.00 am and 5.00 pm. The composition of the diet fed to the sheep is shown in Table 2.6. Fresh water was given ad libitum.

2.2.8.4 Polyester Bag Technique.

The bags were made of synthetic polyester fibre (Sericol Group Ltd, London) with pore size of 40-50 um. They are measured 10 x 21 cm with a round base to prevent the sample building up in the corners (Figure 2.2). Approximately 5.5 g air-dry samples were put in

Table 2.6

Composition of the Diet Fed to the Sheep

820

Ι

Composition of Hay DM (gkg⁻¹)

OM (gkg ⁻¹ DM)	930
CP (gkg ⁻¹ DM)	82
<u>In vitro</u> D-value (%)	53.2
ME (MJkg ⁻¹ DM)	8.2

II - Composition of Concentrates (302 Ewbol Pencils)*

				Vita	min IV l	1 دو	Mine	rals
Protein (%)	Fibre (%)	0il (%)	Ash (%)	A	D ₃	E	Selenium (mgkg ⁻¹)	Magnesium (gkg ¹)
14	13.5	3.0	9.8	5000	2000	7.5	0.2	12.0

* BOCM, UK.

Figure 2.2

Five Polyester Bags Ready to be Inserted into a Fistulated Sheep



The bags were sewn with 2 ply 50 pre-dried bags. denier polyester thread, 10/70 ball point needle, lock stitch and approximately 10 stitches per cm. The neck of the bag was twisted tightly then bound with approximately 115 cm length of strong nylon string, leaving loose length of 50 cm after a knot had been tied. Then the strings were threaded through a 17 cm piece of polythene tubing and then through a holed rubber bung. Further knots were tied to ensure the strings did not slip back through the hole. The ends of the string were melted over a bunsen flame to prevent fraying. For the washing experiment (Chapter 3), four bags were tied into each The bags were pushed into the sheep's rumen bung. through the cannula so that the bung was sealing it. The polythene tubing prevents tangling of string and spreads the bags out in the rumen.

2.2.8.5 Analytical Methods of the Bags Residue after Incubation.

The incubated material still in the bag was oven-dried at 60°C for 48 hours and this was then stored in a dessicator until analysed. Dry matter disappearance was determined by weight difference. Organic matter disappearance was calculated after ashing the undigested residue at 500°C overnight. Dry matter and OM of the original sample was determined by oven-drying at 100°C and ashing at 500°C respectively overnight.

2.2.8.6 Washing the Bags.

After incubation, the bags were then washed in a domestic washing machine (Zanussi Z915T) with hot detergent wash (Kridis <u>et al</u>, 1989). Each wash involved a batch of 25 bags so that each bag could be washed evenly with the detergent.

96

2.2.8.7 Total Nitrogen Determination of Hay Samples and Residues after Incubation.

The total-N content in hay and the dried residues after incubation was determined in triplicate by the Kjeldahl method according to the procedure described by Egan <u>et</u> al (1981).

Reagents.

- 1 The digestion mixture was made up by mixing slowly with cooling the following:
 - a) 40 g selenium oxide in 100 ml distilled water b) 2 l concentrated H_2SO_4
- 2 Buffer 5 g NaOH + 3.74 g anhydrous Na_2HPO_4 + 31.8 g Na_3PO_4 . 12H₂O + 10 cm³ sodium hypochlorite (10-14% av Cl) in 2 litres distilled water.
- 3 Caustic phenol 2.4 g NaOH + 0.1 g sodiumnitroprusside + 20g phenol in 1.6 litres distilled water.
- 4 Ammonia standards ranging from $0.05-0.25g1^{-1}$ ammonia nitrogen were made from a stock solution containing 4.7168g $(NH_4)_2SO_4$ in 1 litre 10% H_2SO_4 (1 g $1^{-1}N$).

The different ammonia-N concentrations were made by appropriate dilution in 10% H_2SO_4 .

Procedure.

a) Digestion

Approximately 0.1-0.15g of the sample was weighed into a 75 ml digestion tube. Using a dispenser, 8 ml of the digestion mixture was added to each tube. Three 1 ml volumes of hydrogen peroxide (100 vol) and one piece of sintered glass were then added. The tubes were heated to 35°C for 2 hours in a block digester (Tecator Ltd, Bristol, England). After cooling, the digest was diluted with 50 ml distilled water and then the volume was made up to exactly 75 ml with distilled water. The digest was thoroughly mixed and then left to stand at room temperature for 1 hour.

b) Analysis for Ammonia-N

Ammonia was measured colorimetrically using the Indo-phenol blue method.

Aliquots of 0.1 ml from the digests and standards were transferred into 50 ml test tubes. Twenty ml of buffer solution and 8.0 ml of caustic phenol were then added to each tube. The tubes were swirled gently and allowed to stand at room temperature for 1 hour for colour development.

Using a spectrophotometer model SP8-500 (Pye-Unicam Ltd, Cambridge, England), absorption was measured at 585 nm. A regression equation was developed between blanks and standards and absorbance from which the N content of unknown samples was calculated. From these measurements, N disappearance from bags was calculated.

2.2.9 Near Infrared Reflectance Spectroscopy (NIR)

All silages were scanned using a Pacific Scientific Co. Neotec 6100 scanning monochromater linked to a Digital Equipment Corp. LSI 11/03 mini-computer. Calibrations were obtained by running the PSU/USDA software of Shenk et al (1981).

The samples were packed into a sample holder which holds the sample between a clear window made of quartz

98

and a pressure pad to ensure a good physical contact between the sample and the window. The sample was submitted to the instrument and then it was illuminated by a narrow beam of light through the window and the reflected radiation was collected by lead-sulphide detectors equally spaced around the incident beam (see Figure 1.6).

The signals from the detectors were converted to digital signals and stored into the coupled computer as the logarithm of reciprocal reflectance ($\text{Log } \frac{1}{R}$). Each sample scan is the mean of 64 individual NIR scans conducted in 55 seconds. The stored sample spectrum consisted of 700 data points ($\text{Log } \frac{1}{R}$ values) taken at 2 nm intervals between 1100 and 2500 nm. After each scanned sample, the quartz glass was wiped thoroughly with clean soft tissue to remove adhering forage particles resulting from the previous scan.

The computer then performed stepwise multiple linear regression analysis to relate optical data to the calibration set at its disposal and then select wavelengths which best fit the component being predicted.

2.2.9.1 Mathematical Transformations

Before calculating regression equations, spectral data are subjected to mathematical transformations. The derivative technique described by Norris <u>et al</u> (1976) was used in this study.

Derivative or difference spectra used in this work were either first or second derivative. A first difference spectrum is the difference between two segments of the spectra (A - B)

99



while a second difference spectrum is computed from (A – B) – (B – C) which is the same as A – 2B + C. For example, a derivative segment of 1262 nm using a 1, 16, 4, 4 math treatment is not a single wavelength but it is a Log $\frac{1}{R}$ segment (W value) computed as:



Similarly, a derivative segment of 1600 nm using a 2, 24, 4, 4 math treatment is a Log $\frac{1}{R}$ segment computed as:



W Value

Therefore, 1600 nm segment includes the change in Log $\frac{1}{R}$ values from 1574 to 1626 nm. A 2, 24, 4, 4 math treatment means that a segment involved was computed by a second derivative (A - 2B + C) with 24 nm gap and a first and second running smooth of 4 nm.

2.2.9.2 Equation Output

After the computer performed a stepwise multiple linear regression ,the equation produced took the form shown in Table 2.7.

The expressions presented in Table 2.7 can be explained as:

- B(O) is the regression constant.
- B(1) B(5) are partial regression coefficients.
- F ratios are variance ratios which indicate how much the given term contributed to the equation.
- Wavelengths 1262 1230 are derivative wavelength segments at Log $\frac{1}{R}$.
- Math treatment was described above.
- Standard error is denoted as the standard error of calibration (SEC) which indicates the errors associated with regression calibration data on optical information. When testing an independent calibration equation against unknown samples and then comparing the observed values against the predicted values, SEC becomes SEP (Standard Error of Prediction).
- R² (ADJ) is the coefficient of determination adjusted for degrees of freedom.

TABLE 2.7

An Example of 5 Term Equation Produced by Stepwise Multiple Linear Regression

The Mean = 71.090 The Standard Deviation = 6.300 N = 122

Standard Error = 2.896 . R-Squared (ADJ) = 0.789

	4	4	4	4	4		ę*	* °	4*
atment	4	4	4	4	4	•H •	0.20	0.32	0.18
Math Trea	16	24	24	24	24	∎ E⊷i a`	0.64	-1.77	0.90
		7	7	7	7				
length	262	662	266	646	230	Res	1.65	-4.23	2.35
Wave	н	1	2	1	1	Pred	80.44	65.61	76.21
μ	37.76	183.22	68.18	16.44	40.06	Actual	82.09	61.38	78.56
Coefficient	67.510 -5938.944	4410.001	917.587	-1108.577	-4583.659	Sample	73	74	83
	11 11	11	H	R	11				
	B(0) B(1)	B(2)	B(3)	B(4)	B(5)				

.

The lower portion of the equation shows unusual observations found in the calibration set:

- 'T' statistics indicates goodness of fit between calibration data and optical data for a particular sample.
- 'H' statistics indicates how much the spectral information of a given sample differs from the spectra of the other samples used in the calibration set.

2.2.10 Neutral Detergent Cellulase OMD Determination The neutral detergent cellulase OMD of the samples was determined according to the method of Dowman and Collins (1982).

The cellulase enzyme derived from fungi Trichoderma viride (BDH Ltd, Poole, Dorset, England) was used. Pepsin (1:10,000) was obtained from Zimmerman and Hoppes Ltd, Milton Keynes, England.

Neutral detergent solution: 93g of disodium ethylene diamine tetraacetate dihydrate (EDTA) and 34 g of sodium borate ($Na_2B_4 \sigma_7 10H_2 O$) were dissolved in distilled water by gentle heating. To this solution 150g of sodium lauryl sulphate and 50 ml of 2-ethoxy ethanol (ethylene glycol monoethyl ether) were added; 22.8 g of disodium hydrogen sulphate was dissolved in distilled water separately and the two solutions mixed and diluted to 5 litres. A check was made to ensure that the pH was in the range 6.9-7.1.

0.5g of sample ground through 1 mm screen was digested for 1 hour with 50 ml of neutral detergent solution, 0.25g of sodium sulphite plus 1 ml of antifoam. The digest was filtered through a porosity 1 sintered glass crucible and

The washed thoroughly with boiling distilled water. transferred while still damp into residue was а polyethylene weighing bottle and excess water removed with a filter stick. The residue was incubated with 30 ml of buffered cellulase (20g litre⁻¹ buffer) for 24 hours During this period the digests were shaken at 40°C. twice. Finally, the residue was filtered through a sintered glass crucible, washed with hot porosity 2 distilled water and then acetone. The contents were dried and weighed. The residue was ignited at 520°C for 4 hours, cooled and reweighed so that the organic matter in the residue could be determined. Independently, the OM in a separate sample was determined by igniting at 520°C for 4 hours.

Neutral Detergent Cellulase OMD% =

Wt of OM in Sample - OM of Sample Residue ----- x 100 Wt of OM in Sample

2.3 Laboratory Methods Comparison

In order to perform a proper comparison between the prediction techniques described previously, it was felt that 3 requirements needed to be satisfied, these include:

- 1 The predictive technique needs to be tested against <u>in vivo</u> calibration set having the following characteristics:
 - i) It should contain at least 100 samples. Abrams <u>et al</u> (1987) recommended that for NIR calibration, a minimum of 100 samples were needed to minimise SEP and bias.
 - ii) The sample must have been evaluated using a standardised in vivo procedure.

- iii) The calibration should include a wide range of sample composition so as to fully represent the range found in practice.
 - iv) The samples must be fully characterised in terms of origin, plant species, preparation and chemical composition.
- 2 The <u>in vivo</u> samples need to be the same and also equal in number for each test.
- 3 То between-laboratory differences, each avoid method must be performed at a single laboratory. In each case, the analysis should be determined by а laboratory which makes routine use of the particular method.

The laboratory methods and laboratories involved in the work reported in this study are listed in Table 2.8.

All 170 silages from four populations were subjected to the predictive methods shown in Table 2.8. To compare the performance of these predictive methods against the population of silages obtained for this study, the calibration and validation samples were selected as described in Table 2.9.

The criterion that was used to select the calibration samples from the total population was that a complete set of predictive methods should exist for each of the chosen samples. This ensured that comparison of the techniques made on a fair and equal basis. was In practice. application of this selection criterion meant that 122 samples were chosen for calibration purposes. This number was found to be sufficient, in particular with NIR calibration. In an NIR study, Abrams et al (1987) recommended that a minimum of 100 calibration samples

Analytical Methods Performed by Contributing Laboratories

Method	Laboratory
MADF (Clancy and Wilson, 1966)	ADAS, Wolverhampton
LICA (Morrison, 1972)	North of Scotland College of Agriculture
NCOMD (Dowman and Collins, 1982)	ADAS, Wolverhampton
PCOMD (Jones and Hayward, 1975)	West of Scotland College
IVOMD (Alexander and McCowan, 1966)	West of Scotland College
NB48 (Kridis <u>et al</u> , 1989)	West of Scotland College
NIR Spectroscopy (Norris <u>et al</u> , 1976)	North of Scotland College of Agriculture

Table 2.9

Calibration and Validation Samples Used in This Study with Their in vivo OMD Distribution

	Calibra	ttion Samp	les		Validat:	ion Samp	les	
Population	No of Samples Used	Mean	Range	ർ	No of Samples Used	Mean	Range	ୟ
Population 1	None	I	F	1	28	70.3	59.2-78.1	5.2
Population 2	72	70.5 5	2.8-82.8	6.6	None	I	I	I
Population 3	23	72.9 6	3.0-78.7	4.1	20	71.8	60.9-81.5	5.1
Population 4	27	70.1 5	5.2-86.7	7.1	None	J	ı	I
Total	122	71.1 5	2.8-86.7	6.3	48	70.9	59.2-81.5	5.2

were needed to minimise SEP and bias.

The remaining 48 samples were used for validation and consisted of samples for which no <u>in situ</u> measurements could be made. This is because the samples in population 1 (No 28) were too finely milled for this method and in the case of population 3, 19 samples contained insufficient material and one sample was being a red clover silage.

2.4 Statistical Analysis

The relationship between <u>in vivo</u> OMD and each predictive technique was calculated using regression techniques. MINITAB Statistical Package (Ryan <u>et al</u>, 1985) was used for this purpose.

Analysis of variance was used to compare the regression lines of the different populations and then detect significant differences between populations in intercept and/or slope. For this purpose, GENSTAT statistical package (1983) was used to calculate 3 statistical models (see Figure 2.3).

The approach was to test significant differences between these 3 models. The test was performed as follows:

I Test Model 1 VS Model 3 -----> NS -----> one regression line to describe all data. F is calculated as:

F = ESS for Model 1 - ESS for Model 3 EDF for Model 1 - EDF for Model 3

> where ESS is Error Sum of Squares, EDF is Error Degrees of Freedom and EMS is Error Mean square.

> > 108

Figure 2.3

The Statistical Models Used to Detect Significant Differences Between Populations



If the test is significant at P < 0.05 then:

- III Test Model 2 VS Model 3 ----> Significant -----> Regression equations differ in slopes. F is calculated as:

If test II and III are significant then regression equations are different in both intercepts and slopes.

the Shenk et al (1981) software performed For NIR. stepwise multiple linear regression analysis which relates optical data to the variable being predicted. In this the software was allowed to select equations study. containing 1-9 wavelengths using all the data from all 122 samples in the calibration set. The derivatised Log R segments (W values) from each population found in the calibration set were then calculated using the central wavelengths and math treatments selected by the software. It was then possible to relate the W values specific for each population to the in vivo calibration data and then produce regression equations for each population. The regression equations for each population were then compared by analysis of variance. The following example illustrates how the calculations were performed in the

case of a 5-term equation:-

- 1 A stepwise multiple linear regression of the 122 calibration samples was carried out and gave the 5-term equation shown in Table 2.7.
- 2 The 5 W values for each sample in this population were calculated from the Log $\frac{1}{R}$ values using the math treatment appropriate to each of the 5 selected wavelengths. Details of this calculation are shown in 2.2.9.1. The computer program used for the calculation is shown in Appendix 1.
- 3 For each population within the 122 calibration set, multiple linear regressions were carried out between the 5 W values obtained for each sample and the <u>in</u> <u>vivo</u> data.
- 4 Differences between slopes and/or intercepts of the regression equations produced in this way for each population were assessed by analysis of variance.

CHAPTER THREE

.

· •

THE EFFECT OF DIFFERENT WASHING PROCEDURES ON THE LOSSES OF OM AND N FROM SAMPLES OF HAY INCUBATED IN POLYESTER BAGS WITHIN THE RUMEN OF SHEEP.

٩.,

.

3.1 Introduction

been known that a limitation of the It has in situ technique is poor between-bag reproducibility. A major source of variation may be inconsistent post incubation Aerts et al (1977) and Cottyn et al (1986) washing. showed that additional washing in acid-pepsin reduced between-bag variability and improved the prediction of forage digestibility, probably by removing forage residues bacteria. and adherent For protein, bacterial contamination of bag residues by microbial matter mav cause an underestimation of protein degradability by the bag method (Mathers and Aitchison, 1981).

Recently, domestic washing machines have been used to wash the bags after incubation (Ørskov <u>et al</u>, 1988). In this chapter, it is intended to investigate the possibility of combining machine wash with a range of different washing reagents to further improve the reliability and repeatability of the in situ results.

3.2 Methods

The use of chopped hay samples would be a good choice for this type of experiment. Compared to ground silage samples, chopped hay samples have the benefit of stability with time and additionally, the need to remove fine materials out of the samples will be avoided. This will ensure that the polyester bag reproducibility test will not be affected by the amount of fines present in the samples.

The preparation of hay samples, animals and their diets, polyester bag technique and the analytical methods are described in 2.2.8.1, 2.2.8.3, 2.2.8.4 and 2.2.8.5 respectively.

3.2.1 **Treatments and Design**

The effect of different washing techniques on the

disappearance of organic matter and nitrogen was studied. This was conducted in three consecutive periods and each period utilised a different washing reagent (see Table 3.1). The bags in period 1 received a wash with neutral detergent solution (ND) (Van Soest and Wine, 1967), in period 2 they received a wash with acid-pepsin (AP) (Tilley and Terry, 1963) and in period 3 they received a wash with commercial washing powder (WP) (non-biological Persil brand).

The different washes were applied before- (NDI, API and WPI); after- (IND, IAP and IWP) or before and after incubation (I) in the sheep (NDIND, APIAP and WPIWP). All bags received a final cold water wash in the washing machine, control bags received only this (see Table 3.1).

Each treatment was replicated in 4 sheep and incubated for 3, 8, 24 and 72 hours making a total of 192 bags.

3.2.2 Zero Time Determination

This procedure was set to determine the disappearance without incubation in the rumen and using the three different washing reagents (see Table 3.2).

3.2.3 Multiple Washing Experiment

This experiment was performed to test the effect of multiple washing on the losses of hay DM from bags.

Approximately 5.5 g of a hay sample were weighed into each bag. From one to four washes in the automatic washing machine were tested. Four bags in each wash were used and either the cold or hot cycle was was utilised. After washing, the bags were oven-dried at 60°C for 48 hours and the DM loss from the bags was calculated by difference.

114

Table 3.1

Treatments and Design ^a

Washing Reagent	Period	Treat. Code	Washing Procedure
None Neutral.		A NDI	Control. Post-incubation cold water machine washing. Pre-incubation treatment with neutral detergent.
Detergent	H	UNICUN	Pre- and Post-incubation treatment with neutral detergent. Post-incubation treatment with neutral detergent.
None		A	Control. Post-incubation cold water machine washing.
Acid		API	Pre-incubation treatment with acid pepsin.
Pepsin	2	APIAP	Pre- and Post-incubation treatment with acid pepsin.
		IAP	Post-incubation treatment with acid pepsin.
		<u>.</u>	
None		A	Control. Post-incubation cold water machine washing.
Washing		IdN	Pre-incubation treatment with washing powder.
Powder	£	JMT JM	Pre- and Post-incubation treatment with washing powder.
		IMP	Post-incubation treatment with washing powder.

.

a Refer to Appendix 2 for further details about each treatment.

.

Zero Time Determination^a

	Washing Procedure
5 Bags	Regular Machine washing with cold water (control)
5 Bags	Treatment with neutral detergent + machine washing
5 Bags	Treatment with neutral detergent + machine washing + treatment with neutral detergent and machine washing
5 Bags	Treatment with acid-pepsin + machine washing
5 Bags	Treatment with acid-pepsin + machine washing + treatment with acid pepsin and machine washing
5 Bags	Treatment with washing powder + machine washing
5 Bags	Treatment with washing powder + machine washing + treatment with washing powder and machine washing

a For machine washing see A in Appendix 2 For neutral detergent treatment see NDI in Appendix 2 For acid pepsin treatment see API in Appendix 2 For washing powder treatment see WPI in Appendix 2

3.2.4 Fitting the Mathematical Model

The fitted OM disappearance values for all washing treatments were obtained using the Ørskov and McDonald (1979) exponential equation which calculates the "line of best fit" through the disappearance values:

 $p = a + b (1 - e^{-ct})$

where a, b and c are constants fitted by an iterative least squares procedure. "p" is the actual disappearance after time "t", "a" is zero hour degradation or the readily soluble materials, "b" is the amount of sample which will have disappeared within the rumen given sufficient time and "c" is the rate constant for the degradation of "b".

Ørskov and McDonald (1979) linked the degradation rate to an estimate of the solid particle outflow rate from the rumen to give the "effective degradability" P.

> bc P = a + ----c + k

where a, b and c are the constants obtained for the previous equation and k is the fractional outflow rate from the rumen per hour. P represents the amount of sample which will actually be degraded in the rumen and its value varies according to the outflow rate.

3.3 Results

3.3.1 Organic Matter Disappearance from Bags

The percentage disappearance of OM from the bags is shown in Table 3.3.

Table 3.3

.

Organic Matter Disappearance (%) of Hay from Polyester Bags Incubated in the Rumen of Sheep and Treated with Different Washing Techniques. Mean of 4 Sheep, one Observation per Sheep.

A Neutral Detergent Wash.

Incubation Times (Hours)

Treatments	0	3	8	24	72	
Control NDI NDIND IND	11.7 21.9 23.7 21.9	18.7 28.4 28.9 27.1	32.0 40.5 43.2 40.0	58.0 62.8 63.6 60.1	71.0 75.6 78.3 74.2	
SED	_	0.837	1.695	1.933	0.872	

B Acid-Pepsin Wash.

Incubation Times (Hours)

Treatments	0	3	8	24	72
Control API APIAP IAP	11.7 25.0 27.7 25.0	16.9 30.6 32.3 27.2	31.1 44.1 47.3 39.8	58.9 66.3 67.5 60.4	73.9 79.7 81.2 76.8
SED	_	0.837	1.695	1.933	0.872

C Washing Powder.

Incubation Times (Hours)

Treatments	0	3	8	24	72	
Control WPI WPIWP IWP	11.7 28.2 35.4 28.2	16.9 32.8 37.0 30.0	29.7 41.3 46.7 41.7	54.9 61.1 66.5 63.3	75.4 80.9 82.4 78.3	
SED	-	0.837	1.695	1.933	0.872	

All treatments significantly (P < 0.001) increased the OM 3 and 8 hours of incubation compared disappearance at with the control. With the exception of IND and IAP disappearance treatments. the OM was significantly (P < 0.05) increased by all treatments compared to the 24 hours of incubation. At 72 hours of control at all increased OM disappearance incubation, treatments significantly (P < 0.05) above the control level.

There is a general pattern in that the OM disappearance was greater from bags which received before incubation washing (NDI, API and WPI) compared to bags which received after incubation washing (IND, IAP and IWP) at all incubation times. An exception to this was the WPI and IWP treatments at 8 and 24 hours of incubation. Before and after incubation washing (NDIND, APIAP and WPIAP) had the greatest effect upon OM disappearance at all incubation times.

The exponential equation (Ørskov and McDonald, 1979) (see 3.2.4) was used to obtain the fitted OM disappearance values for all washing treatments. These disappearances have been shown graphically in Figures 3.1, 3.2 and 3.3 for neutral detergent, acid-pepsin and washing powder treatments respectively. These figures demonstrate that while the 3 washing treatments increased OM disappearance from bags above the control level, the forms of the degradation curves were unaltered.

The dynamics of ruminal activity have already been recognised and feed entered into rumen could either continue to disappear or pass out at any time (Ørskov and McDonald, 1979). In order to allow for the effect of rumen retention time and passage rate on the feed and thus simulating normal rumen conditions, the data in Table 3.4 are the effective degradability at selected outflow rates after subjection of data in Table 3.3 to the Ørskov

FIGURE 3.1

Organic Matter Disappearance (%) of Hay from Polyester Bags Incubated in the Rumen of Sheep and Treated with Netural Detergent Reagent.

HAY OM DISAPPEARANCE



Time in rumen (hrs)

FIGURE 3.2

Organic Matter Disappearance (%) of Hay from Polyester Bags Incubated in the Rumen of Sheep and Treated with Acid-Pepsin Reagent.

HAY OM DISAPPEARANCE Acid pepsin


Organic Matter Disappearance (%) of Hay from Polyester Bags Incubated in the Rumen of Sheep and Treated with Washing Powder Reagent.



122

Table 3.4

Effective Degradability (%) of Hay Organic Matter Disappearance from Polyester Bags Incubated in the Rumen of Sheep and Treated with Different Washing Techniques at Selected Outflow Rates. Mean of 4 Sheep, one Observation per Sheep.

A Neutral Detergent Wash.

Effective Degradabilities (%)

Treatments	ED 2%	ED 4%	ED 8%
Control NDI NDIND IND	56.8 61.8 64.3 60.5	47.0 53.0 55.5 52.0	36.0 43.3 45.3 42.3
SED	0.837	1.046	1.09

B Acid-Pepsin Wash.

Effective Degradabilities (%)

Treatments	ED 2%	ED 4%	ED 8%
Control API APIAP IAP	57.8 66.0 67.8 61.0	47.3 57.3 58.8 52.0	35.8 47.5 49.3 42.8
SED	0.837	1.046	1.09

C Washing Powder.

Effective Degradabilities (%)

Treatments	ED 2%	ED 4%	ED 8%
Control WPI WPIWP IWP	57.0 65.0 68.0 64.0	45.5 54.5 59.5 55.0	34.0 45.3 51.0 45.5
SED	0.837	1.046	1.09

and McDonald (1979) model (see 3.2.4). The effective degradability of the OM followed the same general pattern as with the OM disappearance.

The coefficient of variation (CV%) of OM disappearance is shown in Table 3.5. Figure 3.4 shows the histogram representation of CV of OM disappearance at 24 hours of incubation. All different washing techniques decreased CV% below the control level at 24 hours of incubation with the exception of IND treatment. The coefficient of variation was further decreased at 72 hours of incubation with the exception of bags receiving neutral detergent and acid pepsin treatments (Table 3.5).

Analysis of variance on the data for organic matter disappearance (%) and effective degradability are given in Appendices 3 and 4 respectively. The individual OM disappearances (%) and the effective degradabilities (%) for all washing treatments are shown in Appendix 5.

3.3.2 Nitrogen Disappearance from Bag

The percentage disappearance of Nitrogen from bags is shown in Table 3.6.

A11 treatments (except NDI, 24 hours) significantly (P < 0.001) increased nitrogen disappearance at all incubation times compared with the control. NDI treatment 24 at hours of incubation time was significant at Ρ < 0.05.

The effect of either before or after incubation washing on nitrogen disappearance from bags has been reversed compared with organic matter disappearance. At all incubation times. the bags which received an after incubation washing (IND, IAP and IWP) clearly have a greater nitrogen loss compared with bags which received before incubation washing (NDI, API and WPI). At 3

Table 3.5

Coefficient of Variation (%) of Organic Matter Disappearance of Hay from Polyester Bags Incubated in the Rumen of Sheep and Treated with Different Washing Techniques.

A Neutral Detergent Wash.

Incubation Times (Hours)

Treatments	3	8	24	72
Control NDI NDIND IND	3.7 8.2 2.7 3.8	11.5 2.2 3.2 3.2 3.2	4.8 4.3 1.8 7.6	2.9 4.7 3.5 4.4

B Acid-Pepsin Wash.

Incubation Times (Hours)

Treatments	3	8	24	72
Control	7.9	8.4	10.8	0.91
API	10.1	3.5	2.2	2.3
APIAP	3.9	1.3	2.0	1.8
IAP	4.5	6.9	5.2	1.2

C Washing Powder.

Incubation Times (Hours)

Treatments	3	8	24	72
Control	2.8	5.8	7.3	2.8
WPI	1.8	3.8	4.3	1.5
WPIWP	2.3	5.8	2.4	1.9
IWP	2.8	10.3	4.0	1.4

)

Coefficient of Variation (%) of Organic Matter Disappearance of Hay from Polyester Bags Treated with Different Washing Techniques after 24 hours of Incubation.



Table 3.6

Nitrogen Disappearance (%) of Hay from Polyester Bags Incubated in the Rumen of Sheep and Treated with Different Washing Techniques. Mean of 4 Sheep, one Observation per Sheep.

A Neutral Detergent Wash.

Incubation Times (Hours)

Treatments	0	3	8	24	72	
Control NDI NDIND IND	16.4 37.1 44.9 37.1	27.5 46.7 56.9 48.4	41.9 58.3 73.0 65.4	64.8 68.6 80.9 79.9	79.3 84.3 91.9 89.3	
SED		1.916	2.561	1.629	0.622	

B Acid-Pepsin Wash.

Incubation Times (Hours)

Treatments	0	3	8	24	72
Control API APIAP IAP	16.4 77.9 83.9 77.9	24.0 76.4 82.0 79.2	39.9 75.0 88.6 81.9	64.5 72.7 88.0 86.6	79.0 83.4 89.7 88.1
SED		1.916	2.561	1.629	0.622

C Washing Powder.

Incubation Times (Hours)

Treatments	0	3	8	24	72
Control WPI WPIWP IWP	16.4 46.2 63.1 46.2	23.0 54.9 70.0 55.7	35.7 67.0 77.8 70.8	63.2 78.5 90.6 87.5	83.7 88.7 95.3 93.3
SED ·	_	1.916	2,561	1.629	0.622

hours of incubation the differences between before or after incubation washing were not significant. However, at 8 the differences were significant at hours of incubation 0.05 with the exception of WPI and IWP treatments. P < At 24 and 72 hours of incubation, these differences were highly significant (P < 0.001). The before and after WPIWP) the incubation wash (NDIND, APIAP and has at all effect nitrogen disappearance greatest upon incubation times.

The graphical representation of nitrogen disappearance from bags is shown in figures 3.5, 3.6 and 3.7 for neutral acid-pepsin and washing powder treatments detergent, respectively. То demonstrate the effect of washing reagents upon the removal of nitrogen from bags, Table 3.7 shows the concentration of nitrogen in bag residues. Figure 3.8 shows the histogram representation of nitrogen concentration in bag residues after 72 hours of incubation. At 24and 72 hours of incubation. the nitrogen concentration in residues were significantly bags (P < 0.001) reduced by after incubation washing (IND, and IWP) compared to either control bags or before IAP incubation washing (NDI, API and WPI).

Analysis of variance on the data for nitrogen disappearance (%) and nitrogen concentration on bag residues are given in Appendices 6 and 7 respectively. The individual Ν disappearance (%) and the nitrogen concentration in bag residues for all washing treatments are shown in Appendix 8.

3.3.3 Multiple Washing Experiment

The percentage DM loss of hay samples from bags is shown in Table 3.8 and Figure 3.9.

There is highly significant (P < 0.001) DM loss from bags which received two washes as compared to one wash.

Nitrogen Disappearance (%) of Hay from Polyester Bags Incubated in the Rumen of Sheep and Treated with Neutral Detergent Reagent.



129

Nitrogen Disappearance (%) of Hay from Polyester Bags Incubated in the Rumen of Sheep and Treated with Acid-Pepsin Reagent.



HAY N DISAPPEARANCE

Nitrogen Disappearance (%) of Hay from Polyester Bags Incubated in the Rumen of Sheep and Treated with Washing Powder Reagent.





Persil

Time in rumen (hrs)

Table 3.7

Nitrogen Concentration ($gkg^{-1}DM$) in Bag Residues.

Treatments	0	3	8	24	72	
Control. NDI NDIND IND	9.8 8.4 7.6 8.4	9.3 7.9 6.6 7.5	8.9 7.5 5.1 6.1	8.7 9.0 5.6 5.5	7.3 6.8 3.9 4.3	
SED	_	0.292	0.437	0.408	0.252	

Incubation Times (Hours)

A Neutral Detergent Wash.

B Acid-Pepsin Wash.

Incubation Times (Hours)

Treatments	0	3	8	24	72	
Control API APIAP IAP	9.8 3.1 2.3 3.1	9.5 3.6 2.8 3.0	9.1 4.7 2.3 3.2	8.9 8.3 3.8 3.5	8.2 8.3 5.7 5.3	
SED	-	0.292	0.437	0.408	0.252	

C Washing Powder.

Incubation Times (Hours)

Treatments	0	3	8	24	72
Control WPI WPIWP IWP	9.8 7.9 6.0 7.9	9.7 7.1 5.1 6.7	9.6 6.0 4.5 5.3	8.4 5.9 3.0 3.6	6.8 6.3 2.8 3.3
SED	-	0.292	0.437	0.408	0.252

Nitrogen Concentration (gkg⁻¹DM) in Bag Residue after 72 hours of Incubation.



•

Table 3.8

The Effect of Multiple Washing on the Losses of DM (%) from Hay Samples Incubated in Polyester Bags. Means of 4 Observations

Number	DM Loss (%)					
of Washes	Cold Wash	Hot Wash				
1	15.2	19.9				
2	17.6	22.2				
3	17.9	23.4				
4	19.1	24.0				
SED ⁺	0.28	0.374				

.

:

The Effect of Multiple Washing on the DM Loss (%) from Hay Samples Incubated in Polyester Bags.

÷



Thereafter, the losses were gradual indicating that much of the loss occurred in the first and second washes.

3.4 Discussion

In recent years, the in situ method has been used extensively as a measure of the rate of degradation of Many factors inherent in this feedstuffs in the rumen. method have been recognised which may affect the rate of An important factor which has not been degradation. studied carefully is the washing procedure after bags are washing removed from the rumen. The different techniques described in this chapter were intended to investigate the following:

- 1 What is the effect of these washing techniques in reducing between-bag variability?
- 2 What is the ultimate consequence of these washing procedures on altering the form of the degradation curves?
- 3 For protein degradation measurements, can these procedures be regarded as a useful addition to the bag technique in removing adhering bacteria from bag contents after incubation in the rumen?

3.4.1 Organic Matter Disappearance from Bags

The results presented in Table 3.3 and Figures 3.1, 3.2 and 3.3 clearly indicate that the three washing reagents increased OM disappearance from bags. This is in close agreement with the findings reported by Chenost <u>et al</u> (1970) and Playne <u>et al</u> (1978). They showed that subjecting the bags after removal from rumen to a second stage digestion with acid-pepsin identical to that used by Tilley and Terry (1963) increased DM disappearance. When the exponential equation of Ørskov and McDonald (1979) was applied to the disappearance of OM from the bags, the forms of the degradation curves were unaltered which suggest that the results are as biologically meaningful as those obtained with a cold water wash only (control) (see figures 3.1, 3.2 and 3.3).

•

The bags which received before incubation washing (NDI, noticed to OM API and WPI) were have greater disappearance compared to bags which received after incubation washing (IND, IAP and IWP) (see Table 3.3). This observation may be explained by firstly, the washing reagent, either ND, AP or WP, may modify the cell wall and make it more accessible to microbial attack during the This observation has been subsequent rumen incubation. reported previously by Jones and Hayward (1975) and Roughan and Holland (1977) where they indicated that the matter solubility was increased when a herbage drv treated with the cellulase enzyme is preceded with acidpepsin or neutral detergent respectively.

Secondly, the before incubation bags received an extra machine wash when compared to the after incubation bags which might increase the loss of particulate matter from described by Playne et al the bags as (1978) and Lindberg and Knutsson (1981). This has been confirmed by the multiple washing experiment (see Table 3.8 and Figure 3.9) where there was significant (P 0.001< difference in the DM loss between bags washed once or twice in the washing machine.

When allowing the factors of rumen dynamics to be considered by calculating the effective degradability of OM disappearance (see Table 3.4), all washing treatments increased the effective degradability of hay in the rumen in a pattern similar to that of OM disappearance (Table 3.3).

137

The precision of the in situ technique was found to be improved by the washing reagents. At 24 hours of all treatments decreased the coefficient of incubation. variation below the cold water wash only (control) (see Table 3.5 and Figure 3.4) with the exception of the IND treatment. The reason that the ND treatments didn't improve the between-bag variability at 72 hours of due experimental incubation is possibly to the and technical difficulties associated with washing the bags in the neutral detergent solution. The percentage reduction in CV at 24 hours of incubation was 41, 67 and 45% for WPI, WPIWP and IWP treatments respectively compared to the control. The detergent washing mav remove contaminating residues both from the bag material and contents which enhance between-bag reproducibility.

of variability between The reduction bags has been observed previously by Playne et al (1978) when acidpepsin was used to wash the bags after rumen incubation. They reported that the CV for the 48 and 72 hours of incubation had been reduced from 19.7 to 11.7% after the Our coefficient was much lower acid-pepsin treatment. than they reported possibly due to the use of the washing machine which gives more uniformly effective washing than the regular tap water wash.

3.4.2 Nitrogen Disappearance from Bags

The results presented in Table 3.6 and Figures 3.5, 3.6 and 3.7 show that nitrogen disappearance was greatly increased by the washing reagents. Neutral detergent and washing powder treatments did not however, alter the form of the degradation curve, but for bags receiving the acid-pepsin treatment, nitrogen disappearance was unaffected by the period of incubation and so no curve was fitted. This is because of the enzymatic nature of pepsin which solubilises protein readily at zero incubation This suggests (see Figure 3.6). that for protein degradation measurements. the use of acid-pepsin as a washing agent is inappropriate. Nitrogen disappearance from bags receiving after incubation washing (IND, IAP and IWP) is increased compared to bags receiving before incubation washing (NDI, API and WPI) and the longer the incubation the greater the effect was (see Table 3.6). This greater nitrogen loss may suggest that the detergent washing may remove adhering bacterial nitrogen which is bound to hay samples after rumen incubation. This conclusion might be supported by the lower nitrogen concentration in bags residues for the bags receiving after incubation washing as compared with bags receiving before incubation washing (see Table 3.7 and Figure 3.8).

٠.

The contamination of bag residues by microbial matter has Mathers and Aitchison been observed in previous reports. reported (1981)that feedstuff residues may be significantly contaminated by micro-organisms and this contamination increased linearly with time in the rumen. This contamination has been supported by electron microscopical studies which showed that rumen bacteria colonise and adhere to plant particles during fermentation (Akin. 1979). These observations could explain our case here where nitrogen loss was greater and nitrogen in residues concentrations bag were lower for bags receiving after incubation washing as compared with bags receiving before incubation washing.

3.5 Conclusion

The use of detergent washing appears to be a useful addition to the in situ technique. For organic matter, post-incubation detergent washing reduces between-bag variability without altering the form of the degradation curve. This suggests that this modification to the in situ technique can be potentially applied with relative confidence to measure forage OM disappearance. For protein degradation measurements, post-incubation detergent washing might remove contaminating bacteria which could otherwise lead to underestimation of protein degradability. However, further work is required to measure the precise effect of detergent washing on removing this contamination.

Washing the bags after rumen incubation with domestic washing powder in the washing machine is both cheap and convenient.

,

.

CHAPTER FOUR

.

.

.

PREDICTION OF ORGANIC MATTER DIGESTIBILITY OF GRASS SILAGES

4.1 Introduction

UK system for ruminants used in the rationing The То requires a knowledge of the ME content of the ration. date there have been insufficient in vivo measurements of ME (see 1.1.3)for the development of satisfactory prediction equations and the UK advisory services have therefore adopted the practice of developing regression equations for the prediction of OMD or DOMD. From these predicted values, it is possible to convert DOMD to ME using a constant factor (see 1.1.4).

To date. UK advisory laboratories have relied on а limited number of populations of silages for developing equations to predict in vivo OMD and DOMD. In this chapter, it is intended to examine the population of silages obtained from different sources around the UK [see Table 2.1] (for which in vivo digestibility data were available) using a range of laboratory methods and then to explore the possibility of establishing an improved predictive technique which could be used by all of the advisory laboratories in the UK.

4.2 Results

The predictive methods used in this study may be categorised as:

- I Chemical Methods
- II Biological Methods
- III A Physical Method (NIR)

4.2.1 Chemical Methods

The results of using MADF and LIGA to predict <u>in vivo</u> OMD are shown in Table 4.1 and figures 4.1 and 4.2 respectively.

Table 4.1

			Predictors							
Pop. No.	n		MADF	LICA	NCOMD	PCOMD	NB48 QMD	IVOMD	NIR 8-term	
2	72	R ² RSD%	0.54 4.5	0.73 3.4	0.62 4.1	0.79 3.0	0.70 3.6	0.75 3.3	0.87 2.4	
3	23	R ² RSD%	0.24 3.5	0.20 3.6	0.28 3.4	0.53 2.8	0.32 3.3	0.48 2.9	0.59 2.6	
4	27	R ² RSD%	0.49 5.0	0.89 2.3	0.57 4.8	0.82 3.0	0.77 3.4	0.83 2.9	0.91 2.2	
TOTAL	122	R ² RSD%	0.34 5.1	0.52 4.4	0.54 ^a 4.3	0.55 4.2	0.68 3.6	0.74 3.2	0.85 2.5	
More than one line ^b		***	***	**	***	NS	NS	NS		
Diffe int Diffe	rences ercept rences	s in ts 3 in slope	*** s NS	*** NS	** NS	***				

Regression Statistics for the Prediction of in vivo OMD of Grass Silages

a = three values were missing, one from population 2 and two from population 4.

b = shows whether the regression lines for individual silage populations are significantly different.

 R^2 adjusted for degrees of freedom. =

**** =

- *** = significant at P < 0.001** = significant at P < 0.01NS = not significant at P = 0.05.



(%) QWO 0714 UI



(%) (WO 0414 UI

The R^2 and RSD% for MADF and LIGA for the overall relationship (N = 122) were 0.34, 5.1 and 0.52, 4.4 respectively. LIGA gave regression for populations 2 and 4 with R^2 = 0.73 and 0.89 respectively, whereas the R^2 for population 3 was rather poor (R^2 = 0.20) [Table 4.1].

For both the MADF and LIGA methods, the regression lines of differ obtained for the three populations silages < 0.001) in intercepts (see Table 4.1 significantly (P and Figures 4.1 and 4.2). Each population regression intercept differed significantly from the other line (see Table 4.2) for MADF relationships. However, for LIGA relationships, it was population 2 and 3 and 3 and 4 for which the intercepts differed significantly.

4.2.2 Biological Methods.

The results of using NCOMD, PCOMD, NB48 OMD and IVOMD to predict in vivo OMD are shown in Table 4.1 and Figures 4.3, 4.4, 4.5 and 4.6 respectively.

The R^2 and RSD% for NCOMD and PCOMD for the overall relationship were 0.54 and 4.3; 0.55 and 4.2 respectively. For PCOMD, the R^2 and RSD% for populations 2, 3 and 4 were 0.79, 3.0; 0.53, 2.8 and 0.82, 3.0 respectively whereas for NCOMD, it was 0.62, 4.1; 0.28, 3.4 and 0.57, 4.8 respectively (Table 4.1).

The regression lines obtained for the three populations of silages differed significantly (P < 0.001) in both intercepts and slopes for PCOMD and only intercepts for NCOMD (P < 0.01) [see Table 4.1 and Figures 4.3 and 4.4].

For PCOMD, each population regression line intercept and slope differed significantly from each other. However, for NCOMD the intercept differences were between population 2 and 4 only (see Table 4.2).

Table 4.2

Regression Coefficients and Significance of Between-Population Differences in MADF, LIGA, NCOVD and PCOVD Regression Equations for the Prediction of in vivo OVD for Silages

MADE

Co- efficients		Regress	ion Coeff	F-Test			
		Pop 2	Pop 3	Pop 4	2 v 3	2 v 4	3 v 4
One Slope Different Intercepts	-0.12	111.1	118.0	114.4	***	***	*

LICA

Co-	Regression Coefficients					F-Test		
efficients		Pop 2	Pop 3	Pop 4	2 v 3	2 v 4	3 v 4	
One Slope Different Intercepts	-14.158	106.8	114.9	107.9	****	NS	***	

NCOMD

Co-		Regression Coefficients					
efficients		Pop 2	Pop 3	Pop 4	2 v 3	2 v 4	3 v 4
One Slope Different Intercepts	0.58	31.9	33,2	35.2	NS	**	NS

TOTIN

Co-	Regressio	on Coeff	icients	F-Test		
efficients	Pop 2	Pop 3	Pop 4	2 v 3	2 v 4	3 v 4
Different Slopes	0.68	0.43	0.99	*	** *	****
Different Intercepts	27.4	47.0	17.6	***	***	***
*** = Significant at	P < 0.001		** = S	Significant	at P < 0	.01
* = Significant at	P < 0.05		NS = N	bt Signific	cant at P	= 0.05











Tഋ

(*) GMO 0VJV UI

For both the NB48 OMD and IVOMD methods, the R^2 and RSD% for the overall relationship were 0.68, 3.6 and 0.74 and 3.2 respectively (Table 4.1).

In contrast with the methods mentioned previously, both the NB48 OMD and IVOMD methods can describe the three populations of silages by a single regression line (Table 4.1 and Figures 4.5 and 4.6). Table 4.3 shows the regression coefficients of the regression equations in predicting the <u>in vivo</u> OMD by the NB48 OMD and IVOMD methods as indicated by a single regression line for all populations.

Analysis of variance of the regression equations for the 3 populations used in the calibration set for chemical and biological methods is shown in Appendix 9. Analysis of variance of the individual population regression equations to indicate which population line is different from the other for the MADF, LIGA, NCOMD and PCOMD methods is shown in Appendix 11. The individual values for <u>in vivo</u> OMD, chemical and biological parameters for the 122 calibration silages used in this study is shown in Appendix 12.

4.2.3 The Use of Bivariate Relationships to Predict in vivo OMD.

The chemical and biological predictors studied in this thesis have been combined in a bivariate relationship to study whether there was an improvement in predicting in vivo OMD over the univariate relationship for the 122 calibration silages. Table 4.4 shows that the combination of predictors in bivariate relationships can improve the prediction of in vivo OMD over the univariate relationship shown in Table 4.1. All possible combinations can improve the prediction over the univariate relationship significantly, with the exception of the bivariate relationships fMADF with IVOMD, PCOMD, NCOMD

Table 4.3

Regression Coefficients for the Prediction of <u>in vivo</u> OMD by NB48 OMD and IVOMD

Method	Regression C	befficients
	Intercept	Slope
NB48 QMD	, 1.44	0.82
IVAMD	9.96	0.92

¥

.

Table 4.4

Bivariate	Regres	ssion	Stati	stic.	cs f	\mathbf{or}	the	Prediction	of
	in	vivo	OMD	\mathbf{of}	122	Si	lages	3	

Predictors	R ²	RSD%	Significance of the Improvement
IVOMD + NB48 OMD IVOMD + PCOMD IVOMD + NCOMD IVOMD + MADF IVOMD + LIGA	0.76 0.75 0.75 0.74 0.77	3.1 3.2 3.2 3.2 3.0	*** * NS ***
NB48 OMD + PCOMD NB48 OMD + NCOMD NB48 OMD + MADF NB48 OMD + LIGA	0.71 0.71 0.69 0.72	3.4 3.4 3.5 3.4	*** *** * *
PCOMD + NCOMD PCOMD + MADF PCOMD + LIGA	0.61 0.55 0.61	3.9 4.2 3.9	*** NS ***
NCOMD + MADF NCOMD + LICA	0.55 0.65	4.2 3.8	NS xxx
LIGA + MADF	0.52	4.4	NS

a = To test whether there was a significant improvement over the univariate relationship.

- R^2 = Adjusted for degrees of freedom.
- NS = Not Significant at P = 0.05.
- * = Significant at P < 0.05.
- ** = Significant at P < 0.01.
- *** = Significant at P < 0.001.

.

154

٠.

and LIGA.

The analysis of variance test is shown in Appendix 13.

4.2.4 A Physical Method (NIR)

Figure 4.7 shows the mean NIR spectra of all silages investigated in this study. The calibration regression NIR spectroscopy and the statistics obtained by between-population differences in NIR significance of regression equations for the prediction of in vivo OMD are shown in Table 4.5. As the number of terms included in the regression increased, the R^2 value increased and the The R² and SEC % for the NIR 8-term SEC decreased. equation were 0.85 and 2.5 respectively.

Analysis of variance showed that at least six terms must be included in the calibration equation to avoid significant between-population differences in regression equations (see Table 4.5). Figure 4.8 shows the relationship between the OMD predicted by an 8-term multivariate regression equation and <u>in vivo</u> OMD for the calibration set.

The wavelengths selected by stepwise multiple linear regression to produce calibration regression equations for the prediction of in vivo OMD are shown in Table 4.6.

Figure 4.9 shows a graph of the correlation coefficients between NIR optical densities and <u>in vivo</u> OMD of the 122 calibration silages at each wavelength. The minimum and maximum spectral peaks indicate wavelengths where the correlation is strongest.

Analysis of variance of the regression equations for the 3 populations used in the calibration set for the NIR regression equations is shown in Appendix 10. An example of 700 Log $\frac{1}{R}$ values (data points) for one silage sample only is shown in Appendix 14. The NIR regression

FIGURE 4.7

Average NIR Spectrum of All Silages Studied in this Work.



Overall Population Statistics and Significance of Between-Population Differences in NIR Regression Equations for the Prediction of Silage OMD for Silages Included in the Calibration Set

Table 4.5

No of Terms	R ²	SEC %	More Than One Line	Differences in Intercept	Differences In Slope
. 1	0 52	հհ	***	***	NC
1	0.52	4.4			GN
2	0.64	3.8	****	***	NS
3	0.71	3.4	****	** **	*
4	0.76	3.1	***	***	NS
5	0.79	2.9	***	***	NS
6	0.82	2.7	NS	-	-
7	0.83	2,6	NS	***	-
8	0.85	2,5	NS	-	-
9	0.85	2,4	NS	-	-

 R^2 adjusted for degrees of freedom.

**** = Significant at P < 0.001 = Significant at P < 0.05*

NS = Not Significant at P = 0.05


Table 4.6

*1738" 9th 1150" 1842' 8th2354" 1842' 1150" 7th Wavelength Selection by SMLR (λ 1150" 2426" 2426" 2426" 6th 1646" 1646" 1646" 1274 1 1274 5th 1262' 1230" 1274' 1646" 1646" 22541 1230" 4th 1230" 1274" 1230" 1230" 1230" 1910" 1850" 1274' 3rd 1274' 2266" 2266" 2266" 2266" 1662" 1662" 1262' 1262' 2nd 2290" 1662" 1662" 1662" 1662" 1662" 1662" 2266" 2266" 1st No of Terms

Wavelength Selection by Stepwise Multiple Linear Regression for the Prediction of in vivo OMD of Silages by NIR.

4 math treatments. , t t t 1st derivative using 1, 16, 2nd derivative using 2, 24, 11 11 E

4 math treatments.

Wavelengths are ordered according to its contribution to each added term.

< 10. Wavelength which has variance ratio *

FIGURE 4.9





CORRELATION COEFFICIENT

equations produced by stepwise multiple linear regression for the prediction of <u>in vivo</u> OMD of 122 calibration silages are shown in Appendix 15.

4.2.5 Validation of the Calibration Equations Produced by the Traditional Methods and NIR

Validation procedure is an important step to test the predictive power of any prediction equation. In this study, 48 silages were reserved for this purpose (see Table 4.7 shows the results of the validation Table 2.9). procedure for all methods tested in this study except the NB48 OMD method where data was not available (see 2.3). For the NIR calibration equations, this table shows that the SEP decreases and the R^2 increases progressively as up to six terms are included in the calibration equation. mean bias and However, for the the slope of the regression between measured and predicted in vivo OMD, there is no clear pattern in relation to the number of Murray (1986) states that the criteria of selecting terms. an NIR calibration equation for predictive purposes should be based on the following validation statistics (in vivo OMD against predicted OMD): SEP (minimum), bias (minimum) and slope (closest to unity). In addition, Westerhaus (1985c) suggests that no explanatory wavelength segment should have an F statistic of less than 10. In our study the calibration equation based on an 8-term equation best meets these criteria. Figure 4.10 shows the relationship between measured in vivo OMD of 48 validation silages and that predicted using the 8-term NIR calibration equation. The results of the NIR 8-term calibration equation produced by the stepwise multiple linear regression are shown in Table 4.8.

Of the traditional methods, the IVOMD method was the best, giving the highest R^2 and lowest SEP (see Table 4.7). The validation of the IVOMD equation with the 48 validation silages is shown graphically in Figure 4.11.

TABLE 4.7

Method	n	R ^{2a}	SEP ^b	Bias ^C	Slope ^d
NIR	1 2 3 4 5 6 7 8 9	0.33 0.46 0.52 0.61 0.68 0.76 0.71 0.76 0.73	4.2 3.8 3.5 3.3 3.1 2.7 3.0 2.6 2.8	+0.22 -0.05 -0.29 -0.81 -0.14 -1.04 -0.03 -0.79 -0.70	0.81 0.79 0.91 0.86 0.78 0.89 0.78 0.93 0.88
MADF		0.20	5.1	-0.59	0.52
LICA		0.14	5.3	1.18	0.48
NCOMD ^e		0.29	5.7	-3.40	0.65
PCOMD		0.40	4.7	-2.33	0.71
IVOMD		0.64	3.6	-1.85	0.89

Validation Statistics for the Prediction of in vivo OMD of 48 Silages not Included in the Calibration Set

 $a = R^2$ adjusted for degrees of freedom.

b = SEP standard error of prediction not corrected for bias.

c = Mean bias between measured and predicted OMD.

d = Slope of regression between measured and predicted OMD.

e = NCOMD contains two missing values.

.

.

n = Number of terms in the NIR calibration equation.



Table 4.8

NIR 8-term Equation Produced by Stepwise Multiple Linear Regression

.

	Coefficient F		Wavelength	Math Treatment			
B(0) = B(1) = B(2) = B(3) = B(4) = B(5) = B(6) = B(7) = B(8) =	63.973 -4311.132 -6062.110 3244.727 1027.880 -1716.427 -4355.047 1921.474 3245.068	15.47 35.83 93.18 104.43 43.90 46.93 23.83 14.57	1842 1274 1662 2266 1646 1230 2426 1150	1 1 2 2 2 2 2 2 2	16 24 24 24 24 24 24 24	4 4 4 4 4 4 4 4	4 4 4 4 4 4 4 4 4

SEC % = 2.5 R-squared (ADJ) = 0.85

.



The individual values for <u>in vivo</u> OMD, chemical and biological parameters for the 48 validation silages used in this study is shown in Appendix 16.

4.2.6 Validation of NIR 8-term Equation with Irish Silages.

Forty grass silages which had been evaluated in vivo at the Department of Agriculture for Northern Ireland were second validation test of the selected obtained as а NIR 8-term calibration equation. This validation will be rigorous exercise to considered as another test the performance of the NIR calibration equation on another set of UK silages obtained from a different source other than the sources studied in present work. The validation test is shown in Figure 4.12. The bias and slope between in vivo OMD measured and the OMD predicted were 1.74 and 0.99 respectively.

4.2.7 Calibration of NIR with IVOMD.

In the present work, it was felt necessary to investigate the possibility of calibrating the NIR with the IVOMD The same 122 calibration silages and the 48 method. validation silages were used in this case. The results of the calibration and validation statistics are shown in Tables 4.9 and 4.10respectively. Based on the validation statistics, the NIR 5-term calibration equation may be the best equation to predict the IVOMD of 122 calibration silages. The R^2 and SEC % for the calibration equation were 0.83 and 2.4 respectively. The R^2 , SEP %. bias and slope for the validation equation were 0.71, 2.8, 0.74 and 0.89 respectively. The relationship between measured IVOMD and that predicted by the 5-term equation for the calibration set is shown in Figure 4.13. Figure 4.14 shows the relationship between the measured IVOMD of the 48 validation silages and that predicted using the 5-term NIR calibration equation.



.

Calibration Statistics Produced by Stepwise Multiple Linear Regression for the Prediction of IVOMD by NIR (n = 122)

No of Terms	R ²	SEC %		
1	0.50	4.2		
2	0.66	3.5		
3	0.77	2,8		
4	0.80	2.6		
5	0.83	2.4		
6	0.84	2.3		
. 7	0.86	2.2		
8	0.87	2.1		
9	0.88	2.1		

 R^2 adjusted for degrees of freedom.

Table 4.10

Valida	tion S	Statis	tics	for	the	Pred	lction	of	IVOMD	of
48	Silage	es no	t Ind	lude	d in	the	Calib	rati	ion Set	

No of Terms	R ^{2a}	SEP % ^b	BIAS ^C	Slope ^d
1	0.14	5.3	2,0	0.57
2	0.54	4.3	2.7	0.99
3	0.65	2.9	0.18	0.97
4	0.55	3.4	0.48	0.86
5	0.71	2.8	0.74	0.89
6	0.65	3.0	0.24	0.89
7	0.64	3.2	-0.84	0.85
8	0.65	3.2	-0.87	0.83
9	0.65	3.4	-1.21	0.80

a = R² adjusted for degrees of freedom.
b = SEP standard error of prediction not corrected for bias.

c = Mean bias between measured and predicted IVOMD.

d = Slope of regression between measured and predicted IVOMD.







The NIR regression equations produced by stepwise multiple linear regression for the prediction of IVOMD of the 122 calibration silages are shown in Appendix 17.

4.3 Discussion

The production of well fermented grass is an important step towards making a good quality silage. pH and volatile N as a percentage of total N are two main indicators of the quality of preservation. The mean pH and volatile N as a percentage of total N for all silages studied in this thesis were 4.2 and 11.5% respectively The achievement of a pH value of 4.2 (see Table 2.3). unwilted silage was found be essential for to for successful preservation (McDonald and Whittenbury, 1973). A recent publication by ADAS (ADAS Paper No 3148, 1988) indicated that a volatile N as a percentage of total N ranging from 8-13% will yield a satisfactory fermentation quality. All silages studied in present work were found to be well preserved as indicated by these parameters.

Murray (1986) stressed that a calibration set with a wide range of analytical data is the most important criterion for successful NIR calibration. The calibration silages studied in this thesis have DOMD values spanning a range of 47.5 to 81.2%, a mean of 64.7% and standard deviation of 5.7%. These figures were found to be compatible with farmers silages analysed in the 1988/89 silage season by the Scottish Agricultural Colleges Advisory Service (see Table 4.11). This suggests that the DOMD values for the calibration silages used in this work were wide enough to cover the range which is likely to be encountered in practice.

4.3.1 Chemical Methods

The use of MADF to predict in vivo OMD of grass silages gave an overall R^2 and RSD% of 0.34 and 5.1 respectively

Table 4.11

The Distribution of DOMD Values for SAC Advisory Service Silages* and the Calibration Silages used in this Study

	SAC Silages ^{**} n = 2125	Calibration Silages n = 122
Mean	64.10	64.7
Min	46.90	47.5
Max	76.80	81.2
SD	4.45	5.7

* The DOMD% values are predicted.

** SAC Silages figures were compiled by Collin Jessiman of North of Scotland College of Agriculture for the 1988/89 season.

(Table 4.1 and Figure 4.1). Wilkins (1981) reported that the correlation between MADF and in vivo digestibility was not as close with silages as with other classes of Alderman et al (1971) quoted an R² of 0.28 for forages. the relationship of in vivo DOMD and MADF for 45 grass better prediction was reported by silages. However, Barber et al (1984) when they quoted an R^2 and RSD% of silages. 3.4 for 80 A further improved 0.48 and prediction was reported by Givens et al (1989a) when they quoted an R^2 and SEP% of 0.57 and 4.1 respectively when predicting in vivo OMD of 124 grass silages.

Analysis of variance of the population regression lines study obtained in the present revealed that the relationship between in vivo OMD and MADF yield significant (P < 0.001) differences in the regression equations and these differences were caused by а significant difference in intercepts (P 0.001) [Table < Moreover, 4.1 and Figure 4.11. each population regression line intercept is significantly different from the other (Table 4.2). These differences were obtained even though the MADF values used were all obtained from one laboratory. Thus, the difference between intercepts is not explained by between-laboratory differences in the measurement of MADF.

These results confirm the observation by Barber <u>et al</u> (1984) that the MADF relationship with <u>in vivo</u> DOMD for population 3 differs substantially from that of population 2. Barber <u>et al</u> (1984) suggested these differences might be caused by geographical differences in the relationship between the measurements made using MADF and <u>in</u> vivo digestibility.

Lignin is considered to be a major factor limiting forage digestibility. For the prediction of 122 in vivo grass silages, LIGA method gave an R^2 and RSD% of 0.52 and

4.4% respectively (Table 4.1 and Figure 4.2). Morrison (1973) found an R^2 of 0.71 for the prediction of <u>in vivo</u> DOMD by the LIGA method for 20 grass silages. A more comprehensive study by Givens <u>et al</u> (1989a) quoted an R^2 and SEP% of 0.63 and 3.7% respectively. While these results may suggest an improvement in prediction over the results reported in this work however, our calibration samples are more heterogenous than the samples used by these workers.

As with MADF, the LIGA method relationship with in vivo OMD gave regression lines which differ significantly in intercepts (P < 0.001) [Table 4.1 and Figure 4.2]. For populations 2 and 4, LIGA predicted in vivo OMD with an R^2 and RSD% of 0.73, 3.4% and 0.89, 2.3% respectively, but the regression line obtained for 3 population differed significantly in intercept from the others (Table 4.2) and its precision was rather poor ($R^2 = 0.20$, RSD₈ = 3.6). These differences might be attributed to the unknown proportion of presence of clover silages in population 3. Minson (1982) commented about the large differences in the regression equations obtained for grasses and legumes when relating in vivo digestibility to the lignin in the food. Sullivan (1959) reported that separate regression equations were needed for grasses and legumes when the acid insoluble lignin method was used to predict in vivo DMD.

4.3.2 Biological Methods

The use of NCOMD to predict in vivo digestibility gave an R^2 and RSD% of 0.54 and 4.3 respectively (Table 4.1 and Figure 4.3). Several workers predicted in vivo DOMD by the NCD method. Downman and Collins (1982) reported an R^2 and RSD% of 0.71 and 1.9% respectively for 16 grass silages. For 68 grass silages, Barber et al (1984) quoted an R^2 and RSD% of 0.63 and 2.9% respectively. For 85 grass silages, Givens et al (1989a) reported an R^2 and

SEP% of 0.65 and 3.9% respectively. As has been said with the LIGA method, the improvement in prediction may be attributed to the homogenity of samples used by those workers compared with the samples used in this study.

The PCOMD method gave an R^2 and RSD% of 0.55 and 4.2% (Table 4.1 and Figure 4.4). This precision of prediction is similar to that found by the NCOMD method. However for PCOMD, the individual regressions for populations 2, 3 and 4 gave an R^2 and RSD% of 0.79, 3.0%; 0.53, 2.8% and 0.82, 3.0% respectively, whereas for NCOMD it was 0.62, 4.1%; 0.28, 3.4% and 0.57, 4.8% respectively (Table 4.1). This suggests that the PCOMD method would be adequate for predictive purposes compared to NCOMD. However, obtained differ significantly the regression equations (P \leq 0.001) in both intercepts and slopes (Table 4.1 and Figure 4.4) and each population line is significantly different from the other (Table 4.2). Terry et al (1978) similar they obtained а result when compared the precision of the rumen liquor and pepsin-cellulase methods in predicting in vivo DMD of 48 grasses and 25 legumes. They showed that one regression equation can be used for both grasses and legumes when using the rumen liquor method, whereas the pepsin-cellulase method failed to satisfy this criteria.

The in situ method is an old technique and has been recommended recently to evaluate feedstuffs (Ørskov et al, However, a major limitation of this technique is 1980). the variations between bag results which make it very difficult to standardise (see Chapter 1, in situ method). In this work, an improved bag washing technique was developed and adopted (Chapter 3) and an internal standard was included with each bag incubated inside the rumen so that variations between rumens (between sheep and with time) could be adjusted (see 2.2.8.2). Table 4.12 shows the coefficient of variation and the regression

.

.

Coefficient of Variation and Regression Statistics for NB48 OMD Uncorrected and Corrected for the Internal Standard

	Regression Statistics		
Parameters	CV%	R ²	RSD%
NB48 OMD Uncorrected	3.6	0.68	3.57
NB48 OMD Corrected	2.9	0.67	3.61

statistics obtained for both NB48 OMD uncorrected and corrected for the internal standard. This table indicates that while the CV% decreased (3.6 for uncorrected; 2.9 for corrected), the overall prediction of in vivo OMD of the 122 calibration silages was slightly lower (R 2 for R^2 0.68: uncorrected ---for corrected = 0.67). Accordingly, the NB48 OMD uncorrected values were used However, it was felt that where the in in this work. situ method may potentially be applied (for example as an standard). the inclusion of intermediate an internal standard would be useful. Mehrez and Ørskov (1977) reported that the greatest source of variation for substrate disappearance from the bags was that between sheep followed by between days and the least between In such cases, the inclusion of an internal standard bags. with bags incubated in the rumen would be of value so that these variations can be corrected for.

The length of incubation period may influence in situ results. However, this depends on the material being incubated. For forages, the incubation period tends to be long so that complete digestion of the samples can be obtained. Van Keuren and Heinmann (1962) reported large variations between bags at short incubation times for various forages. These variations diminished as the incubation times increased up to 72 hours. Ørskov et al (1980) recommended an incubation time of 24-60 hours for medium quality forages and 48-72 hours for poor quality roughages. These times are required to reach the asymptote (potential degradation).

In forage evaluation, longer incubation times were found to be a better predictor of <u>in vivo</u> digestibility. At the start of this study, 72 dried silages were subjected to a polyester bag study using 6 and 48 hours incubation time (Kridis <u>et al</u>, 1987). A more accurate prediction of <u>in</u> vivo OMD was found when 48 hours incubation was used compared to 6 hours. These findings are in agreement with those reported by Chenost <u>et al</u> (1970) when they found that <u>in vivo</u> OMD of various forages was better correlated with 48 hours than with 12 and 24 hours incubation time. The 48 hours incubation time has been subsequently used by Aerts <u>et al</u> (1977) and Cottyn <u>et al</u> (1986). Therefore, that an incubation time of 48 hours was used in this study.

The use of nylon bags to predict in vivo OMD of the 122 calibration silages gave an $\ensuremath{\mathsf{R}}^2$ and $\ensuremath{\mathsf{RSD}}\ensuremath{\$}$ of 0.68 and 3.6% respectively (Table 4.1 and Figure 4.5). Few reports have been cited about the use of this method in silage Aerts et al (1977) quoted an R^2 and RSD% of evaluation. 0.79 and 3.4% respectively for the prediction of in vivo OMD of 56 grass silages. A more comprehensive study reported by Cottyn et al (1986) quoted an R^2 and CV% of 0.85 and 3.2% respectively for the prediction of in vivo OMD of 100 grass silages. However, the bags in these reports are further incubated for 48 hours in acid pepsin after removal from the rumen which may explain their precision.

Of the traditional methods, the IVOMD method gave the best prediction of in vivo OMD (R² and RSD% of 0.74 and 3.2 respectively) [Table 4.1 and Figure 4.6]. This method has been used extensively in forage evaluation and has been cited to be the best predictor of in vivo digestibility over a variety of other techniques. 0'Shea et al (1972) found that the IVOMD method gave the best prediction of in vivo OMD of 31 grass silages over various techniques ($R^2 = 0.66$). For 20 grass silages, Morrison (1973) concluded that IVOMD was a better predictor of in vivo OMD than the LIGA method ($R^2 = 0.74$ compared to $R^2 = 0.71$ respectively). More recently, Givens <u>et al</u> (1989a) compared the ability of 11 laboratory methods to predict in vivo OMD and concluded that the IVOMD method

179

gave the highest R^2 and lowest SEP% of 0.76 and 3.1 respectively for 117 grass silages.

methods In contrast with the chemical and enzymic both NB48 OMD and IVOMD reported previously, the populations by single describe all а methods can regression equation (Table 4.1, Figures 4.5 and 4.6 respectively). These findings are in agreement with those They reported that for reported by Terry et al (1978). 48 grasses and 25 legumes, one regression equation can be used for both grasses and legumes when the in vitro DMD method was used to predict in vivo DMD.

4.3.3 The Use of Bivariate Relationships to Predict <u>in</u> vivo OMD

The results shown in Table 4.4 suggested that the combination of predictors in bivariate relationships can improve the prediction of in vivo OMD over the univariate relationship shown in Table 4.1. This is so for a11 possible bivariate relationships with the exception of the bivariate relationships of MADF with IVOMD, PCOMD, NCOMD and LIGA. The significant improvements in prediction over the univariate relationships suggest that predictor the bivariate measures a each of different However, the inclusion aspect of digestion. of other predictors with MADF failed to improve the prediction with the exception of the NB48 OMD + MADF bivariate relationship. This suggests that the IVOMD, PCOMD, NCOMD and LIGA methods measure feed characteristics which include those measured by MADF. These findings are in agreement with those reported by Kridis et al (1987) and Givens et al (1989a) where they indicated significant improvements in the prediction of in vivo OMD of grass silages when the major predictors of digestibility were combined in a bivariate relationship.

180

Although these results may suggest significant improvements in the prediction of <u>in vivo</u> OMD, this improvement does not provide sufficient justification for the determination of two methods simultaneously. The cost involved is a major setback against such a proposition for routine advisory work.

4.3.4 A Physical Method (NIR)

The use of NIR 8-term equation gave an R^2 and SEC of 0.85 and 2.5 respectively (Table 4.1 and Figure 4.8). Analysis of variance showed that at least six terms must be included in the calibration equation to avoid significant between-population differences in the regression equations (Table 4.5).

The selection of an 8-term calibration equation was based the validation procedure reported by Westerhaus on (1985c) and Murray (1986) [see 4.2.5]. The validation process is a true criteria of the performance of equations selected by stepwise multiple linear regression and found to be essential to prevent overfitting, ie: the inclusion of The SEP will decrease to a point and then random noise. increase when overfitting occurs (Table 4.7). This is in contrast with the SEC which decreases progressively with added each term (Table 4.5) [Westerhaus, 1985b]. Another useful validation statistic is the slope of regression between in vivo OMD measured and in vivo OMD predicted (Table 4.7) [Westerhaus, 1985b; Murray, 1986]. The slope should be as close to 1.0 as possible. A large deviation from 1.0 will imply that high and low values will be over or under-estimated (Westerhaus, 1985b). The SEP% and slope for selected 8-term calibration equations were 2.6% and 0.93 respectively (Table 4.7 and Figure 4.10).

The F-statistic is an additional important regression statistic produced by stepwise multiple linear regression

(see Table 2.7). It indicates how much the given term contributed to the equation. Small F-values signify little contribution to the equation, except to fit random errors. Westerhaus (1985c) suggests that no explanatory wavelength should have an F-statistic less than 10 in the calibration equation. The selected 8-term equation reported in this study meets this criteria (Table 4.8).

The use of several terms in the calibration equation to been reported previously. predict digestibility have Norris et al (1976) used a 9-term calibration equation to predict in vivo DMD of 76 various forages. They reported an R^2 and SEC% of 0.78 and 3.6% respectively. Abrams et al (1987) also used a 9-term equation to predict the IVOMD of 100 forages with SEP% and bias of 2.4% and 0.12 respectively. It can be argued that several terms may be required to predict complex parameters like This stems from the digestibility (Murray et al, 1987). fact that a complete digestion process occurring inside the animal may require more spectral information to unravel its complexity. Additionally, there is no danger of using several terms in the calibration equation as long as the validation procedure is conducted adequately. Therefore, our selected 8-term equation would be perfectly valid for this type of study.

The NIR 8-term calibration equation was used to predict another unknown silage population which has been evaluated in vivo at the Department of Agriculture for Northern Ireland. These silages gave а validation statistic of R², SEP %, bias and slope of 0.76, 3.6, 1.74 and 0.99 respectively (Figure 4.12). This validation exercise will be considered another rigorous test of the validity of the selected 8-term calibration equation in predicting another set of UK silages obtained from a different source.

182

Table 4.6 shows the centred wavelengths which explain the 122 in vivo silages. It is clear that the second derivative segments centred at 1162 and 2266 nm were always the primary explanatory wavelengths followed by The choice of 1662" nm and 2266" 1230 nm (Figure 4.9). well established to correlate favourably to nm are Norris et al (1976) found 1666 nm was the digestibility. first explanatory wavelength and 2266 nm was the third explanatory wavelength to be selected for the prediction of in vivo DMD of 76 various forages by NIR. Shenk et al (1979) found the 1641 nm region was important in the Murray et al (1987) found the 1662" prediction of IVDMD. nm was the primary explanatory wavelength and 2266" nm was the third explanatory wavelength to be selected for the prediction of in vivo OMD of 72 grass silages. In the same study, the selection of 1662" nm as the primary explanatory wavelength was supported by NIR calibration with nylon bag 48 hours digestibility.

Out of the 8 explanatory wavelengths found in our 8-term calibration equation, the 1662" and 2266" nm regions will explain 60% of the variance in <u>in vivo</u> OMD prediction. This suggests the importance of these two regions in the prediction of digestibility.

The 1662 nm wavelength, using a 2, 24, 4, 4 math treatment (see 2.2.9.1), will include the change in Log $\frac{1}{R}$ values from 1634 to 1688 nm. This segment was found to measure the aromatic CH at 1680 nm which belongs to (Murray et al, 1987). The 2266" nm Lignin region may associated to methylene be groups (-CH₂-) and to the methoxyl groups (CH_2 -O-) which are frequently found as ether side groups of Lignin. These two regions were always selected as the primary and secondary wavelengths for the prediction of acetyl bromide Lignin by NIR. Therefore, it is likely that these two regions may arise directly from Lignin (Murray et al, 1987).

183

4.3.5 Calibration of NIR with IVOMD

NIR calibration with in vivo measurements to predict digestibility is the ultimate target so that this technology practical potentially be applied for purposes. can digestibility However, using animals to measure is expensive and requires special facilities, and therefore cannot be used on a large number of samples on places which do not possess feed evaluation units. For these reasons NIR calibration with in vivo measurements will not be possible for many evaluation programs.

In this study, the IVOMD method was found to be the best predictor of in vivo OMD when compared to other traditional methods. Ιt gave the best calibration statistics (Table 4.1), described all populations by a single regression equation (Figure 4.6) and gave the highest R^2 and lowest SEP in a validation test (Table 4.7 and Figure 4.11). Therefore, it was felt that this technique may have a potential value as an intermediate standard to replace the expensive in vivo determinations.

The results shown in Table 4.8 and Figure 4.13 indicated the precision of NIR to predict IVOMD. The 5-term calibration equation gave an R^2 and SEC % of 0.83 and 2.4 respectively. This equation was selected on the basis of the validation procedure shown in Table 4.9, which gave an R^2 , SEP%, bias and slope of 0.71, 2.8%, +0.74 and 0.89 respectively (Figure 4.14).

The use of the IVOMD method to calibrate NIR has been reported previously in the literature. Norris <u>et al</u> (1976) reported an R^2 and SEC% of 0.90 and 2.6% respectively to predict IVDMD of 76 forages using 9-term equation. Brown and Moore (1987) reported an R^2 and SEC% of 0.89 and 2.3% respectively to predict IVOMD of 83 forage samples using 6-term equation. All these workers indicated successful NIR calibration with IVOMD.

According to these results, the IVOMD method is probably a successful candidate for NIR calibration. This suggests that the predicted IVOMD values can be used either to calculate <u>in vivo</u> digestibility using appropriate regression equations or its absolute values can be used to rank forages in many plant breeding programs.

4.4 Comparison of Methods

For many years, the search for a simple, economical and accurate laboratory method to predict the nutritive value of grass silages has been pursued vigorously by advisory services in the UK. Not surprisingly, advisers recognised the difficulty of predicting the nutritive value of grass silages when compared to other classes of forages. Factors like the use of additives, wilting and more importantly, the extent of fermentation which takes place during ensilage, may lead to variation in the nutritive value of the resulting silage. In fact, this variation is likely to be greater when it comes to real farm practices. These factors, coupled with the pre-harvest conditions like species, geographical location and stage of maturity, makes silage evaluation increasingly more difficult.

Many attempts have been made to compare the various laboratory methods in order to determine the most reliable method(s) of predicting the nutritive value of silages. Notably, the comparisons of Aerts <u>et al</u> (1977), Barber <u>et al</u> (1984), Cottyn <u>et al</u> (1986) and Givens <u>et al</u> (1989a) were the most comprehensive. These comparisons were previously mentioned in Chapter 1, Section 2.

For the purpose of this comparison, the methods studied will be compared with regard to two aspects:

- a) Prediction accuracy of in vivo OMD and
- b) The differences in population regression equations as

a) Prediction Accuracy

The comparisons performed in this work indicated clearly the superiority of the rumen liquor methods when compared to methods (NB48 OMD, IVOMD) involving the use of enzymes (NCOMD, PCOMD) or chemical methods (MADF, LIGA) (Table 4.1). This related the prediction superiority may be to the rumen micro-organisms the sensitivity of to and unknown factors which might affect known digestibility. The other methods will not recognise such unknown factors, since they rely heavily on measuring the major parameters which are found to affect digestibility (ie: cellulose, hemicellulose and lignin) [Aerts et al, 1977; Cottyn et al, 1986: McQueen, 1986; Givens et al, 1989a].

The precision of the NB48 OMD method was found to be less than in the case of the IVOMD method (Table 4.1). The observation may be explained by the uncontrolled variation associated with the NB48 OMD method, particularly losses of fines from the bags (Lindberg and Knutson, 1981) [see Chapter 1, Section 2, in situ method]. Although the samples were sieved through a 45 jum sieve, it was not possible to extract all fines present in the samples. I believe that the NB48 OMD method may give a better prediction if fresh minced silages were used, however such material was not available for this study.

Compared to the IVOMD method, the NB48 OMD method profits from a lesser variation due to rumen liquor activity, since bags are incubated in the rumen itself. In addition, with the NB48 OMD method the rumen liquor does not undergo any treatment which might affect the rumen microorganisms activity.

While the NB48 OMD and IVOMD methods improved the prediction precision, these methods, however, were found to be difficult to standardise (Barnes, 1973), slow and expensive, their use demands free fistulated animals and the use access to of surgically prepared animals is strongly objected to bv the animal welfare organisations. These make these methods unfavourable for limitations routine use.

These difficulties have led to a search for a method which could utilise the cellulase enzyme as a means of replacing the rumen liquor. The methods of PCOMD and NCOMD were such a development (Jones and Hayward, 1975; Dowman and Collins, 1982). These techniques were found to predict digestibility of silage with relative success, however they were not as precise as the rumen liquor methods (Givens <u>et al</u>, 1989a). This conclusion is in agreement with findings reported in this work (Table 4.1).

For all 122 silages, the PCOMD regression performed slightly better than the regression of NCOMD. In addition, the individual regressions for the three populations were exceptionally better for PCOMD than the NCOMD (Table 4.1). This suggests that PCOMD method would be adequate for predictive purposes for grasses grown in one geographical location. These findings are in agreement with those reported Minson (1982) for by McLeod and grasses and legumes and by Barber et al (1984) and Givens et al (1989a) for grass silages.

187

Minson (1982) commented about the McLeod and better accuracy of the pepsin-cellulase method when compared to the neutral detergent cellulase method. For the prediction of in vivo DMD of 50 grasses, two methods gave an RSD% of 2.6 and 3.3 the respectively. Those workers further investigated this discrepancy and concluded that the pretreatment with neutral detergent introduced more analytical error than the pretreatment with acid pepsin, which accounts for the lesser prediction precision for the neutral detergent cellulase method.

The cellulase enzymes method benefits from its speed, precision and convenience relative to the rumen liquor method. However, the activity of the enzyme between batches and from different suppliers requires constant monitoring so that the method's accuracy can be maintained (Clark and Beard, 1977).

Plant cell wall components are the major limiting factors which affect digestibility of plants and the methods of MADF and LIGA were used to extract these important fractions. These methods have been used extensively to predict digestibility of grass silages and other forages (Barber et al, 1984; Givens. 1986: 1986), speed, Murray, and reproducibility and cheapness were the most attractive features of these techniques. However, their prediction precision was found to be insufficient. In this work, MADF and LIGA were inferior in prediction precision when compared to other methods (Table 4.1). Although the LIGA method performed considerably better than the MADF method, for population 3 the prediction accuracy was rather poor. Possible explanation of this is the presence in many of population 3 of an unknown proportion of clover which suggests the inaccuracies of various lignin techniques when used to predict the digestibility of a mixture of grasses and legumes (Minson, 1982).

Of the methods studied in this work, NIR was the most precise predictor of in vivo digestibility of grass silages (Table 4.1). The ability of this technique to unravel the compositional information which might relate to digestibility and then the use of this information in a multi-variate way, is primarily behind its successful prediction. be carefully However, this technique has to calibrated and the resultant equations must be adequately validated, otherwise meaningless results The speed, accuracy, precision and unit may occur. cost of analysis (see Chapter 7) make this technique attractive for routine use.

b) The Differences in Population Regression Equations as Assessed by Analysis of Variance The methods of MADF, LIGA, NCOMD and PCOMD gave differences between-population in the regression equations obtained (Table 4.1 and 4.2). These differences cannot be explained by betweenlaboratory differences, since each method was determined in one laboratory.

For the MADF and PCOMD methods, each population regression line is significantly different from the other in intercepts and in intercepts and slopes respectively (Table 4.2, Figures 4.1and 4.4 respectively), whereas for LIGA it was only the population 3 regression line intercept which was significantly different from either population 2 or 4 (Figure 4.2, Table 4.2). Possible explanations of these differences are the combination effects of geographical locations and the heterogeneity of the

Rowett Research Institute silages (population 3), since this population contained unknown proportions of legumes.

Aldrich and Dent (1967) have shown an indication of higher digestibility at northern latitudes when southern latitudes in the UK in compared to cocksfoot. Deinum et al (1968) also found that digestibility of perennial ryegrass was significantly influenced by both light intensity and environmental These findings suggest temperature. that the digestibility - laboratory method relationship may affected by geographical site. These be explanations have been reported by other workers (Osbourn, 1978; Barber et al, 1984; Givens et al, 1989b).

Various workers indicated the need for separate regression equations for grasses and legumes when digestibility was predicted by MADF (Clancy and Wilson. 1966), lignin (Minson, 1982) and pepsincellulase (Terry et al, 1978). All these workers showed that the prediction precision was improved when grass and legume regression equations were considered separately. These findings are in agreement with those reported in this work.

The differences in population regression lines for the NCOMD method were somewhat different from the Population 4 was only significantly other methods. different from population 2, but population 3 was not significantly different from either population 2 or 4 (Figure 4.3, Table 4.2). This suggests that the NCOMD was possibly not influenced by either geographical region or the diversity of the RRI since population 3 was not significantly silages, different from population 2, and yet they were from

different geographical regions. Givens et al (1989b) found a similar conclusion when they reported that the neutral detergent cellulase method was not to by county of harvest when used influenced predict the digestibility of grasses grown throughout England and Wales. However, Givens et al (1989b) showed that the pepsin-cellulase method was also not influenced by county of harvest, an observation which disagrees with the findings reported in our study.

In contrast with the methods mentioned previously, the NB48 OMD, IVOMD and NIR methods can describe three populations of silages by the а single regression line (Table 4.1). This implies that these methods were not affected by either the geographical region, or by the heterogeneity of the RRI silages and so the methods are robust enough to be descriptors of in vivo OMD of grass silages. This agrees well with the findings of Terry et al (1978) for samples of dried forages using the IVOMD method.

4.5 Conclusions

<u>In vivo</u> OMD of grass silages was not accurately predicted by the MADF, LIGA, NCOMD and PCOMD methods. All gave between-population differences in the regression equations obtained. This suggests that their use to evaluate silages routinely may be unjustified. The accuracy of the PCOMD method for individual regressions implies its potential usefulness for predictive purposes.

NB48 OMD, IVOMD and NIR predicted silage digestibility more accurately than any other methods tested in this work. Their application requires one single regression equation to describe all populations. The NIR technique performs substantially better than any other method tested, provided that it is calibrated in the manner described in this work. The speed, precision, accuracy and unit cost of analysis (see Chapter 7) are the most attractive features of this technique. Its application for silage evaluation is therefore strongly recommended.

The results obtained in this work suggest the benefit of the IVOMD method as an intermediate standard either to replace the expensive <u>in vivo</u> determinations or as a calibration method for NIR.

CHAPTER FIVE

.

•

THE EFFECT OF EXTERNAL FACTORS IN THE RELATIONSHIP BETWEEN IN VIVO OMD AND ITS PREDICTORS
5.1 Introduction

Plant growth and environment play a major role in determining the nutritive value of forages and it is these two factors which could alter the composition of the standing crop and hence the resulting silage. In the literature, it has been long recognised that environmental and other external factors could influence in vivo laboratory method relationship. These factors include geographical location (Aldrich and Dent, 1967; Osbourn, 1978; Barber <u>et al</u>, 1984), stage of maturity (Deinum <u>et</u> <u>al</u>, 1968), year of harvest (Morgan and Stakelum, 1987; Givens <u>et al</u> 1989b) and age of ley (Givens <u>et al</u>, 1989b).

to accumulate In this study, possible it was some characteristics regarding the history of the silages These characteristics will be used to assess the studied. effect of environmental and other factors on in vivo OMD laboratory method relationship. The latter aspect forms the theme of this chapter.

5.2 Methods

One hundred grass silages evaluated at ADAS Feed Evaluation Unit were categorised according to the known characteristics for these silages. These characteristics are listed in Table 5.1. The number of silages used in this study are shown in Appendix 18.

The silages were grouped according to their characteristics and the regression equations which resulted from regressing <u>in vivo</u> OMD on the laboratory method for each group were assessed by analysis of variance which was then used to compare the regression equations. This allowed the detection of significant differences in intercept and/or slope according to the statistical models described in 2.4.

194

Table 5.1

	Characteristics	Number of Silages
1	Year of Harvest - 1978 - 1980 - 1982 - 1983 - 1985 - 1986	10 19 17 12 15 13
2	Method of Ensiling - Clamp - Big Bale	72 19
3	Cut Number - First Cut - Second Cut - Third Cut	48 22 7
4	Wilting Time (Hours) - ≤ 12 - > 12-40 - > 40	13 52 13
5	Age of Ley (Years) - \$ 1 - > 1-5 - > 5 - Permanent Pasture	17 32 17 16
6	Additive Application - None - Add-F - Sylade + F100	38 20 21
7	Fertilisation (kgN/ha) - < 100 - > 100-200 - > 200	28 49 9

Characteristics of Silages Studied*

* Not all silages have the known characteristics.

5.3 Results

The effects of different characteristics of silages on the <u>in vivo</u> OMD - laboratory methods relationships are summarised in Table 5.2.

both the NCOMD and IVOMD methods, the between For years regression equations were significantly (P \leq 0.001) The NIR 8-term Log $\frac{1}{R}$ different in intercepts. values significant were less affected bv years, being at However, the regression equations were Р < 0.05. significantly different in slopes. Other predictors were found to be unaffected by years.

For cut number, the regression equations based on the MADF and LIGA methods were significantly (P < 0.01 and P < 0.001 respectively) different in intercepts. Other predictors were unaffected by cut number.

For age of ley, only the IVOMD method gave regression equations which differed significantly (P < 0.05) in intercepts, whereas other methods were unaffected.

The method of ensiling, wilting time, additives application and nitrogen fertilisation gave no effect on the regression equations for all predictors.

Analysis of variance results are shown in Appendix 19.

5.4 Discussion

The regression analyses were rather complex and therefore may need а large number of samples for each characteristic to obtain a valid conclusion. This is mainly because the variable being predicted is fairly complex (ie: digestibility) and thus an appreciable amount of error would be attached to it. Therefore, the require regression equations obtained may a sizeable

The Effect of Different Silage Characteristics on The Regression Equations Between in vivo OMD and the Laboratory Methods Studied in this Work

				Predict	tors ¹			
C	haracter- istics	Statistical Models	MADF	LIGA	NCOMD	PCOMD	IVQMD	NIR 8-term
1	Year of Harvest	More than one Line Differences in Intercepts Differences in Slopes	NS - -	NS 	*** *** NS	NS 	*** *** NS	* NS *
2	Method of Ensiling	More than one Line Differences in Intercepts Differences in Slopes	NS - -	NS - -	NS - -	NS -	NS - -	NS - -
3	Cut Number	More than one Line Differences in Intercepts Differences in Slopes	* ** NS	* *** NS	NS - -	NS -	NS - -	NS -
4	Wilting Time	More than one Line Differences in Intercepts Differences in Slopes	NS - -	NS - -	NS - -	NS - -	NS - -	NS - -
5	Age of Ley	More than one Line Differences in Intercepts Differences in Slopes	NS - -	NS - -	NS - -	NS - -	* * NS	NS - -
6	Additives Application	More than one Line Differences in Intercepts Differences in Slopes	NS - -	NS - -	NS - -	NS - -	NS 	NS - -
7	Fertilis- ation	More than one Line Differences in Intercepts Differences in Slopes	NS - -	NS - -	NS - -	NS - -	NS - -	NS - -

*** = Significant at P < 0.001

NS = Not Significant at P = 0.05

** = Significant at P < 0.01= Significant at P < 0.05

1 = NB48 OMD method not reported here

because of missing data.

number of samples to control that error. However, in many situations this requirement may sometimes seem difficult to achieve and thus the researcher is obliged to accept the available information at his disposal and conclusions based on the analysis can be only provisional.

In this experiment therefore, we were not able to increase the number of samples for each silage characteristic so that our conclusions can be fundamentally sound. This may be attributed to the various groups of silages studied with each group having different characteristics (Table 5.1) and more importantly, the variable was an <u>in vivo</u> parameter which did not permit us to acquire more samples easily.

The results reported in this chapter indicated that the regression equations based on the IVOMD and NCOMD methods have been strongly affected by the year of harvest (significant at P < 0.001, Table 5.2). These findings are in agreement with those reported by Barber et al (1984), where they reported a parallel lines relationship when various laboratory methods including the in vitro and neutral detergent cellulase methods were used to predict in vivo DOMD of hays. For fresh grasses, Morgan and Stakelum (1987) also found significant intercept differences when the IVOMD method was used to predict in vivo OMD of 35 grasses. However in the same study, the NCOMD method was found to be unaffected by years. These workers showed better predictions when each year was considered alone for the IVOMD method.

The effect of years has also been reported by Givens <u>et</u> <u>al</u> (1989b). Those workers showed significant differences not only in intercepts but also in slopes of the regressions obtained for the IVDOMD method when used to predict <u>in vivo</u> OMD of various grasses. For the neutral detergent cellulase method, Givens et al (1989b) reported significant differences only in intercepts, a finding which has been shown in this work (Table 5.2). However, for the regression equations based on the PCDOMD and MADF significant methods Givens et al (1989b) showed а intercepts slopes and difference in intercepts and respectively, an observation which disagrees with the findings reported in this study (Table 5.2). However, our results must be viewed from the fact that all of the analytical work was done in the same year in contrast to other studies so that a significant difference between years suggests a real difference in the in vivo OMD laboratory method relationship rather than a variation between years in analytical techniques.

The years effect on the IVOMD method raises queries about the stability of this regression over varying years. This instability may also be applied to the NCOMD method. However, the fact that the PCOMD method, another enzyme method, was not affected by years did not validate this conclusion. The disagreement between the two enzymic methods may be related to the technical differences in the actual determinations for each method. McLeod and Minson (1982) indicated that the pretreatment of grass samples with neutral detergent introduced more analytical error than the pretreatment with acid pepsin. This error accounted for the lesser accuracy of predicting in vivo of 50 grasses by the neutral detergent cellulase DMD method when compared to the pepsin-cellulase method.

Slope differences were also noted when the NIR 8-term equation was used as a predictor. However, this method was noticed to be less affected by years when compared to the previous methods (significant at P < 0.05, Table 5.2).

The second notable factor which was found to affect in vivo OMD - laboratory method relationship is cut number

(Table 5.2). Significant differences in intercepts were the regression equations based on the found for MADF (P < 0.01) and LIGA (P< 0.001) methods. This is in agreement with work reported by Deinum et al (1968), where they found a strong influence of stage of maturity on regression equations based on ADF and Lignin This has been confirmed lately by Morgan for 12 grasses. and Stakelum (1987) when ADF was used to predict in vivo The effect of cut number on the MADF OMD of 35 grasses. and LIGA relationships is not surprising since we know that cell wall components increase as plant advances in maturity.

For age of ley, intercept differences were found when the IVOMD method was used as a predictor. However, this effect was slightly less when compared to the years effect (significant at P < 0.05). This is in agreement with a study reported by Givens et al (1989b). However, those workers found that the regression equations based on the MADF method were found to be significantly different in intercepts and slopes, findings which disagree with results reported in this work (Table 5.2). Givens et al reported that (1989b) also the PCDOMD and NCDOMD methods were not affected by age of ley. This is in agreement with results shown in this study.

For nitrogen fertilisation, all predictors studied in this work were unaffected by this parameter (Table 5.2). This is in close agreement with results reported by Deinum <u>et al</u> (1968) with regards to the ADF and Van Soest's Lignin methods.

The method of ensiling, wilting time and additive application were found not to affect the regression equations of any predictor reported in this work (Table 5.2). This may not be surprising since these characteristics were expected to effect fermentation quality

5.5 Conclusions

The IVOMD and NCOMD methods were strongly affected by year of harvest which also showed a slight effect on NIR. Only the MADF and LIGA methods were affected by cut number. The IVOMD method was slightly affected by age of ley.

The method of ensiling, wilting time, additive application and nitrogen fertilisation were found not to affect the regression equations of any predictors studied in this work.

The fact that the number of silages for each characteristic was small and variable in number may warrant further investigation. A large population of silages with known characteristics is advisable so that the conclusions presented in this work may be tested more fully. CHAPTER SIX

.

.

Prediction of <u>in vivo</u> DOMD and DE of Grass Silages

.

SECTION 1

Prediction of DOMD of Grass Silages

6.1.1 Introduction

The term DOMD (or 'D' value) is the most widely used digestibility measurement among farmers and advisory services in the UK (Barber et al, 1984). Those services adopted the practice of predicting OMD, have but to measure the ash content of silages and then calculate The measurement of the ash content of silages was DOMD. found to be useful so as to alert the farmer about the extent of soil contamination in his silage.

In Chapter 4, it has been reported that the NIR method was the most accurate predictor of <u>in vivo</u> OMD. In this section it is intended to explore further the possibility of using NIR to predict <u>in vivo</u> DOMD and examine the optional routes available to achieve this target.

DOMD can be predicted by using 3 optional routes which all involve the use of NIR:

Route 1

Direct prediction of DOMD by NIR, ie: NIR is calibrated directly against in vivo DOMD.

Route 2

Indirect approach which requires:

- a) OMD prediction from NIR OMD calibration equation and
- b) Ash prediction from NIR ash calibration equation.

From a) and b) DOMD can then be calculated as:

DOMD% = OMD NIR predicted * OM (100 - Ash% NIR predicted)

Route 3

As route 2 but ash is actually measured in muffle furnace.

Figure 6.1 summarises the 3 optional routes available for this study.

6.1.2 Methods

NIR calibration equations were For each route. the produced using the 122 calibration silages and then crossvalidated using the 48 validation silages (see Table 2.9). calibration and the validation statistics for the The prediction of in vivo DOMD, in vivo OMD and ash are shown in Tables 6.1, 6.2 and 6.3 respectively. The NIR regression equations produced by stepwise multiple linear regression for the prediction of in vivo DOMD and ash are shown in Appendices 20 and 21 respectively. According to validation statistics described 4.2.4. the in the calibration equation based on 8, 8 and 3-term equations were the best to describe the prediction of in vivo DOMD, in vivo OMD and ash content of grass silages studied. These selected NIR calibration equations and its subsequent validation procedure formed the basis by which the predictive routes were compared.

6.1.3 Statistical Analysis

The actual and predicted DOMD values of the validation silages were used to assess the prediction performance of each route. For this purpose, 4 statistical terms were calculated:



Figure 6.1

•

The Three Optional Routes Available to Predict DOMD



Calibration	and	Valida	ation :	Stati	lstics	for	The	Prediction	of
	in	vivo	DOME) of	Grass	Sil	ages		

No of	Calibration	Statistics	Va	lidation	Statisti	ics	
Terms	R ² a	SEC%	R ² a	SEP% ^b	Bias ^C	Slope ^d	
			,				
1	0.48	4.2	0.30	4.3	-0.81	0.84	
2	0.60	3.7	0.51	3.8	-1.38	0.84	
3	0.71	3.2	0.62	3.4	-1.24	0.81	
4	0.75	3.0	0.64	3.6	-1.61	0.77	
5	0.76	2.9	0.65	3.4	-1.11	0.76	
6	0.78	2.7	0.56	3.6	-0.87	0.76	
7	0.80	2.6	0.62	3.1	-0.21	0.95	
8	0.83	2.4	0.64	2.9	-0.16	0.93	
9	0.83	2.4	0.64	3.2	-1.13	0.94	

 $a = R^2$ adjusted for degrees of freedom. b = SEP standard error of prediction not corrected for bias.

 \mathbf{c} = Mean bias between measured and predicted DOMD.

d = Slope of regression between measured and predicted DOMD.

No of	Calibration	Statistics	Val	idation	Statisti	.cs
Terms	R ² a	SEC%	R^{2a}	SEP% ^b	Bias ^C	Slope ^d
			,			
1	0.52	4.4	0.33	4.2	0.22	0.81
2	0.64	3.8	0.46	3.8	-0.05	0.79
3	0.71	3.4	0.52	3.5	-0.29	0.91
4	0.76	3.1	0.61	3.3	-0.81	0.86
5	0.79	2.9	0.68	3.1	-0.14	0.78
6	0.82	2.7	0.76	2.7	-1.04	0.89
7	0.83	2.6	0.71	3.0	-0.03	0.78
8	0.85	2.5	0.76	2.6	-0.79	0.93
9	0.85	2.4	0.73	2.8	-0.70	0.88

Calibration and Validation Statistics for The Prediction of in vivo OMD of Grass Silages

.

 $a = R^2$ adjusted for degrees of freedom. b = SEP standard error of prediction not corrected for bias.

c = Mean bias between measured and predicted OMD.

d = Slope of regression between measured and predicted OMD.

No of	Calibration	Statistics	Va	lidation	Statisti	lcs
Terms	R ² a	SEC%	R ² a	SEP% ^b	Bias ^c	Slope ^d
1	0.30	1.3	0.08	2.1	1.08	0.96
2	0.40	1.2	0.33	1.7	0.65	1.40
3	0.46	1.1	0.50	1.5	0.54	1.60
4	0.50	1.1	0.43	1.6	0.59	1.30
5	0.60	1.0	0.27	1.8	0.66	0.69
6	0.63	0.95	0.34	1.8	0.57	0.62
7	0.65	0.93	0.39	1.7	0.68	0.67
8	0.69	0.88	0.35	1.9	1.10	0.97
9	0.72	0.83	0.22	2.2	1.40	0.66

Calibration and Validation Statistics for The Prediction of Ash Content of Grass Silages

.

 $a = R^2$ adjusted for degrees of freedom. b = SEP standard error of prediction not corrected for bias.

c = Mean bias between measured and predicted ash.

d = Slope of regression between measured and predicted ash.

3 SEP corrected for bias = (Actual DOMD - Predicted DOMD - Bias)² N - 1

Here the intercept and slope were obtained from the regression equation of actual DOMD vs Predicted DOMD for each route.

The calculated SEPs obtained from steps 2, 3 and 4 were compared by analysis of variance to examine the significance of bias and/or intercept and slope correction.

6.1.4 Results

The results of the <u>in vivo</u> DOMD prediction using the 3 routes are summarised in Table 6.4.

The mean bias between actual and predicted in vivo DOMD were the lowest for route 1 (-0.16) followed by route 3 (-0.65) then route 2 (-1.019). In terms of prediction precision, the $\ensuremath{\mathsf{R}}^2$ and SEP% not corrected for bias were the highest for route 3 (0.78, 2.43) followed by route 2 (0.71, 2.87) then route 1 (0.64, 2.97). Analysis of variance suggests that routes 1 and 3 produced nonsignificant bias correction, whereas in route 2 the bias correction was required (significant at P < 0.05). A11 routes gave а non-significant intercept and slope correction. Analysis of variance test is shown in Appendix 22.

6.1.5 Discussion

The use of NIR to predict directly in vivo DOMD (route 1) gave an R^2 and SEP% in a validation test of 0.64 and 2.97

Predictions of in vivo DOMD Using Various Routes by NIR

Significance of Intercept and Slope Correction	SN SN	SN
Significance of Bias Correction	SN *	SN
SEP Corrected for Intercept and Slope	3.00 2.73	2.37
SEP Corrected for Bias	3.00 2.71	2.37
SEP not Corrected for Bias	2.97 2.87	2.43
R ^{2b}	0.64 0.71	0.78
Bias a	-0.16 -1.019	-0.65
Routes	Route 1 Route 2	Route 3

a = Mean Bias between actual and predicted in vivo DOVD

 $b = R^2$ adjusted for degrees of freedom.

* = Significant at P < 0.05.

NS = Not Significant at P = 0.05.

respectively. This is inferior to the NIR prediction of <u>in</u> <u>vivo</u> OMD ($R^2 = 0.76$ and SEP% = 2.6). However for DOMD, this is not surprising since NIR will attempt to predict both OMD and ash content of silages and hence more spectral information is required to predict both parameters. The precision of <u>in vivo</u> OMD prediction was better for route 2 ($R^2 = 0.71$ and SEP% = 2.87) and route 3 ($R^2 = 0.78$ and SEP% = 2.43) than for route 1 (R^2 = 0.64 and SEP% = 2.97).

The lesser prediction precision for route 2 compared to route 3 was entirely due to the rather weak prediction of ash content by NIR (Table 6.3). However, this inferior prediction of ash can be viewed from the fact that have reflectance any minerals do not spectra and correlation existing between NIR and minerals is related variation of these minerals the in accordance to with the variations of some organic constituents present in the samples (Shenk et al, 1979) The selected 3-term ash calibration equation gave an $R^2 = 0.46$ and SEC% = 1.1. and the validation test gave an $R^2 = 0.50$ and SEP% = 1.5 (Table 6.3). This inferior prediction of ash gave a significant (P < 0.05) bias correction for route 2 when compared to either routes 1 or 3 (Table 6.4). This type 2 unfavourable of observation makes route for the prediction of in vivo DOMD.

The choice of NIR DOMD prediction either directly (route 1) or indirectly (route 3) is a matter of debate. Route 1 will offer a simple and economical approach since no ash measurement is required. However, precision has to be sacrificed. On the other hand, it can be argued that route 3 is more desirable from an advisory point of view. This stems from the fact that this route is more precise than route 1 and, more importantly, the measurement of ash content of a farmer's silage in itself is an essential requirement for advisory services to give the farmer an

211

indication about the extent of soil contamination in his silage.

6.1.6 Conclusions

NIR can predict directly <u>in vivo</u> DOMD, however with lesser precision than predicting <u>in vivo</u> OMD. This approach is simpler and more economical since ash determination is not required. However, the calculation of <u>in vivo</u> DOMD by NIR prediction of <u>in vivo</u> OMD and then measuring ash content, will provide a more precise prediction and also the ash measurement in itself is a useful indicator of the extent of soil contamination in silage samples. For this set of silages, ash content was not predicted with sufficient precision by NIR.

Prediction of the DE of Grass Silages

6.2.1 Introduction

The literature contains few reports about the prediction of the DE of grass silages from laboratory measurements. Those reports include that of Alderman et al (1971) who predicted the DE of 45 grass silages from various laboratory methods. In addition, O'Shea et al (1972) reported DE prediction of 31 grass silages from the IVOMD and other fibre methods. In a study by Morrison (1973), the use of IVOMD, MADF and LIGA methods to predict the DE of 20 grass silages was reported. More recently, NIR been used to predict the DE of 60 various has also forages including silages (Eckman et al, 1983).

While these reports attempted to predict DE with various degrees of accuracy, the number of silages used were limited and it may be that the herbages used to make the silages were grown in a confined area.

The objective of this section is to report the ability of the various laboratory methods studied in this work to predict <u>in vivo</u> DE using a large number of silages obtained from different sources around the UK.

6.2.2 Methods

One hundred and forty silages which have been evaluated in vivo for DE, were collected from different sources around the UK. Table 6.5 shows the sources of silage populations used for this study.

The individual values for <u>in vivo</u> DE, chemical and biological parameters used in this study are shown in Appendix 23.

Sources	and	in	viv	οI	ΟE	Dis	stribution	of
S	ilages	U	sed	in	Tł	nis	Study	

Source	DE	MJ kg ⁻¹ oven	Number	
	Mean	Range	SD	
ADAS Feed Evaluation Unit - Pre-1980	14.6	10.8-17.4	1.67	27
ADAS Feed Evaluation Unit - Post-1980	15.0	9.5-20.6	2.03	71
Rowett Research Institute	15.4	13.1-17.5	1.02	42
TOTAL	15.0	9.5-20.6	1.73	140

6.2.3 Statistical Analysis

The significant differences between population regression lines were assessed by analysis of variance as described in 2.4.

6.2.4 Results

The results of using the laboratory methods to predict in <u>vivo</u> DE are shown in Table 6.6. Of the conventional methods, the best predictors of in vivo DE were PCOMD ($R^2 = 0.32$, RSD = 1.43) and IVOMD ($R^2 = 0.31$, RSD = 1.44) followed by LIGA ($R^2 = 0.22$, RSD = 1.53), NCOMD ($R^2 = 0.16$, RSD = 1.6) and MADF ($R^2 = 0.09$, RSD = 1.65).

For all these methods, the regression lines obtained for the three populations of silages differed significantly in intercepts, being significant at P < 0.001 for MADF, LIGA and NCOMD, significant at P < 0.01 for PCOMD and significant at P < 0.05 for IVOMD.

The best predictor of in vivo DE found in this study was the use of NIR (Table 6.7). The 9-term equation gave an R^2 of 0.72 and SEC of 0.91. In contrast to the conventional methods, one regression line can be used to describe all data. This is true for all NIR terms with the exception of 1 which term significant gave (P < 0.001) regression line differences in intercepts.

Analysis of variance of the regression equations for the 3 populations used in the calibration set for the prediction of <u>in vivo</u> DE by conventional predictors is shown in Appendix 24. Analysis of variance for the NIR regression equation is shown in Appendix 25. The NIR regression equations produced by stepwise multiple linear regression are shown in Appendix 26.

215

Overall Regression Statistics and Significance of Between-Population Differences in Laboratories Methods Regression Equations for the Prediction of <u>in vivo</u> DE (MJ kg oven DM) (n = 140)

Predictors ¹	R ²	RSD	More than One Line	Differences in Intercepts	Differences in Slopes
MADF	0.09	1.65	**	***	NS
LIGA	0.22	1.53	***	***	NS
NCOMD	0.16	1.60	**	****	NS
PCOMD	0.32	1.43	**	**	NS
IVOMD	0.31	1.44	*	*	NS
NIR 8-Term	0.71	0.93	NS	-	-

- R^2 adjusted for degrees of freedom
- *** = Significant at P < 0.001
- ** = Significant at P < 0.01
- * = Significant at P < 0.05
- NS = Not Significant at P = 0.05
- 1 = NB method was not reported here because of missing data

Overall Population Statistics and Significance of Between-Population Differences in NIR Regression Equations for the Prediction of Silage DE (MJ kg oven DM) (n = 140).

No of NIR Terms	R ²	SEC	More than One Line	Differences in Intercepts	Differences in Slopes
1	0.38	1.36	**	****	NS
2	0.49	1.24	NS '		-
3	0.54	1.17	NS		-
4	0.59	1.11	NS	-	-
5	0.63	1.05	NS	-	-
6	0.66	1.01	NS	-	-
7	0.70	0.95	NS	-	-
8	0.71	0.93	NS	-	_
9	0.72	0.91	NS	-	-
					·

- R^2 adjusted for degrees of freedom
- *** = Significant at P < 0.001
- ****** = Significant at P < 0.01
- NS = Not Significant at P = 0.05

6.2.5 Discussion

The use of the MADF and LIGA methods to predict in vivo DE gave R^2 and RSD of 0.09, 1.65 and 0.22, 1.53 respectively (Table 6.6). These figures are in close agreement with those reported by Alderman et al (1971) and Morrison (1973). Alderman et al (1971) found an R^2 of 0.04 and 0.23 when in vivo DE of 45 grass silages was and Deriaz Lignin respectively. bv MADF predicted Similarly, Morrison (1973) reported an R² of 0.03 and 0.19 for the prediction of in vivo DE of 20 grass silages by the MADF and LIGA methods respectively. For the NCOMD and PCOMD methods, a striking difference between the two methods in their predictive ability to in vivo DE was observed (R^2 of 0.16 for NCOMD to 0.32 for PCOMD).

The use of the IVOMD method to predict in vivo DE gave an R^2 and RSD or 0.31 and 1.44 respectively. While this prediction is relatively poor, the study conducted by O'Shea <u>et al</u> (1972) gave a better prediction ($R^2 = 0.64$) for 31 grass silages. However, this better prediction may be related to the limited number of silages used by O'Shea <u>et al</u> (1972).

Brown and Radcliffe (1971) commented about the poor predictability of silage DE from digestibility measurements and concluded that the energy values of the various organic compounds present in silages varied independently of digestibilities.

In this study, NIR proved to be the best predictor of in <u>vivo</u> DE. The 9-term equation gave an R^2 of 0.72 and RSD of 0.9, a precision which is substantially better than the conventional methods used in this work. Eckman <u>et al</u> (1983) obtained a similar result when they predicted the DE of 60 various forages with an R^2 of 0.67 using the 2-term NIR equation.

218

The precision of NIR to predict DE may be related to its ability to predict both silage digestibility and GE, even though the samples were dried. This suggests that the GE of grass silages is largely variable and could be for when it is used with a digestibility accounted 1989a). То measurement to predict DE (Givens et al, demonstrate this aspect, the GE of silages used in this study was combined with the IVOMD method in a bivariate relationship to predict DE. The prediction power was increased from an R^2 of 0.32 for IVOMD alone to 0.80 when IVOMD and GE were in combination, suggesting the importance of GE in energy evaluation of silages (see 1.1.3, GE section).

In addition to the low predictability of DE by the conventional methods, the regression lines obtained from the three populations of silages differed significantly in intercept (Table 6.6). Only NIR described all populations by a single regression line (Table 6.7).

The above results clearly highlight the difficulty of DE prediction for silages. Methods other than NIR are unsatisfactory predictors of DE. In fact, attempting to predict ME will be more difficult since the error in ME prediction is greater than DE prediction error. Large proportions of this error can be accounted for by the GE silages, suggesting the importance of GΕ of in the evaluation process.

While the use of NIR makes a substantial improvement in DE prediction ($R^2 = 0.72$), however this is inferior to the NIR prediction of <u>in vivo</u> OMD ($R^2 = 0.85$).

6.2.6 Conclusions

The DE of grass silages was poorly predicted by the laboratory methods studied in this work. This suggests that the DE prediction (and similarly ME) is more complex and cannot be explained by a simple digestibility measurement. Only the multi-term NIR equation makes a significant improvement in DE prediction. The GE of silages plays an important role in energy evaluation of silages.

,

,

CHAPTER SEVEN

.

GENERAL DISCUSSION

indicators of the useful ME and digestibility are the most nutritive value of silages and the former is the basis for allocating energy allowances for ruminants in the UK. Ideally, the of animals to measure these parameters is the ultimate use measurement however, this approach is labour intensive, expensive and requires special facilities and consequently, it is not likely to be a practical proposition. Therefore, a rapid, simple, accurate and comparatively inexpensive laboratory method(s) to predict these parameters is an absolute requirement.

ADAS Feed Evaluation Unit proposed several regression equations to predict ME from simple laboratory methods (Barber <u>et al</u>, 1984; Givens, 1986). However, advisers have long recognised that those equations were unreliable and gave an ME value which resulted in an over-prediction of animal performance (Barber <u>et al</u>, 1983). Four plausible reasons were postulated to explain this overprediction, namely:

- a) The GE of silage is widely variable;
- b) The fate of the energy dense silage volatiles is unknown;
- c) The accuracy of predicting methane energy by Blaxter and Clapperton's (1965) equation is questionable; and
- d) The unreliability of the current efficiency values of ME utilisation for silages. (see 1.1.4)

This situation created some disagreement among advisers about which direction the prediction of the nutritive value of grass silages should proceed. The routes illustrated in Figure 1.4 indicated the two paths that may be used by the advisory organisations in the UK in calculating ME.

The choice of any path shown in Figure 1.4, is a matter of debate. However, the uncertainties attached to the direct

prediction of ME, which have been stated previously, may prevent the immediate adoption of route II. Until these uncertainties are fully understood and then removed, the adoption of route I will provide an alternative simple approach which has been found to be more understandable and can deliver sound practical advice (Barber <u>et al</u>, 1983). It is important therefore, for the advisory services to find a reliable predictive method which can predict <u>in</u> vivo digestibility of grass silages accurately and inexpensively.

The search for such a laboratory method(s) has been a major concern of the different advisory organisations in the UK. Distinguished attempts to develop prediction equations for OMD of grass silages were reported by Barber <u>et al</u> (1984) and Givens <u>et al</u> (1989a).

Until recently, the different advisory laboratories in the UK have used different laboratory methods to predict digestibility of grass silages with various degrees of accuracy. Table 7.1 illustrates the methods used by those laboratories in the UK.

Commercial NIR instruments became an accepted methodology in forage evaluation and recently ADAS, SAC and DANI have adopted this technique in their silage evaluation programs. However, instead of a direct calibration of NIR on in vivo digestibility, those laboratories used NIR to predict fibre parameters (MADF, LIGNIN and MADF respectively). Digestibility and ME were then calculated using the appropriate regression equation. The overall error of silage evaluation is increased by this indirect approach because the error of the NIR prediction must be added to the prediction equation relating fibre error of the content to digestibility.

The work described in this thesis was intended to examine this problem by testing the ability of various laboratory methods, including those used routinely by advisory services in the UK, to predict in vivo OMD of a large population of silages gathered from

223

Table 7.1

•

Methods Used by Advisory Laboratories in the UK

Laboratory	Methods Used
ADAS	MADF NIR (MADF)
SAC	LIGA NIR (LIGA)
DANI	MADF NIR (MADF)

different sources around the UK (Table 2.1). A longer term aim of this work was to explore the possibility of establishing an improved predictive technique which could be used by all advisory laboratories in the UK.

For all 122 calibration silages studied in this work (Chapter 4), the rumen liquor methods (NB48 OMD, IVOMD) gave more precise predictions than either the enzymic methods (NCOMD, PCOMD) or the chemical methods (MADF, LIGA). This better prediction may be related to the sensitivity of the rumen micro-organisms to the known and unknown factors which might affect digestibility. Other predictors will not be expected to recognise these unknown factors since they rely heavily on measuring the major components which are found to affect digestibility (ie: cellulose, hemicellulose and Lignin).

For individual population regressions, the PCOMD method was found to give superior prediction of <u>in vivo</u> OMD (Table 4.1). This suggests that this technique might be adequate for predictive purposes for grasses grown in one geographical region.

The LIGA method gave a rather poor prediction of <u>in vivo</u> OMD of silages obtained from the Rowett Research Institute ($R^2 = 0.20$). This inferior prediction may be explained by the presence in many of these silages of unknown proportions of clover. This suggests the imprecision of the LIGA method for the prediction of the digestibility of a mixture of grasses and legumes (Minson, 1982).

The NB48 OMD method gave less precise predictions when compared to the IVOMD method (Table 4.1). This inferior prediction of the NB48 OMD method may be related to the inherent uncontrolled variations associated with this technique. Two important sources of variation may be important to improve the predictive power of this technique. Firstly, the washing procedure after bags are removed from the rumen. This aspect has been investigated in detail in Chapter 3 where polyester bags were subjected to a range of different washing reagents. The results indicated that the combination of machine washing with non-biological washing powder after bags were removed from the rumen reduced considerably the between-bag variability without altering the form of the degradation curve. This washing procedure was found to be both cheap and convenient.

The second source of variation is the differential losses of fines from bags during rumen incubation. It was felt that the use of freshly minced silage may improve the prediction. However, such material was not available for this study.

Of the methods tested in this study, the NIR method gave the most precise prediction of <u>in vivo</u> OMD of grass silages (Table 4.1). The ability of this technique to unravel compositional information contained in the light reflected off samples and then analyse this information by multiple regression analysis is primarily behind its accurate prediction.

Analysis of variance of the regression equations obtained for the populations of silages used in the calibration set indicated that significant differences in intercepts were found for the regression equations based on the MADF, LIGA and NCOMD methods, and in both intercepts and slopes for the PCOMD method (Table 4.1). One regression line can be used to describe all data when the NB48 OMD, IVOMD and NIR techniques were used as predictors. This implies the robustness of these methods as predictors of digestibility. The fact that the MADF, LIGA, NCOMD and PCOMD methods gave between-population differences in the regressions obtained suggests the inadequacy of these techniques as predictors of digestibility, and their use to evaluate silages routinely may not be justified.

On the basis of this investigation, only the NIR method can provide a reliable alternative to the conventional methods studied in this work. Provided that this technique is calibrated and validated in the manner described in this work, the speed, precision and cost effectiveness are the most attractive features of

226

this method. These qualities may give strong justification for NIR to be applied on a national basis.

With increasing economic pressures on advisory laboratories, the selected method needs to provide an economic and fast service for Tables 7.2 and 7.3 illustrate the unit cost of analysis farmers. and laboratory turn-around time at its best for some of the methods tested in present work. These tables show the relatively low unit cost of analysis for NIR when compared to other methods. Even though the MADF method has been found to be cheap and fast to operate, the NIR method can provide a competitive price, and greater speed and accuracy than the MADF method. In terms of labour input, only NIR requires the minimal labour when compared to other methods (Dixon, Personal Communication). To scan a silage sample, which takes less than a minute, simple drying and grinding are the only required procedures to report ME and other determinations.

<u>Practical Evaluation of MADF and NIR Regression Equations</u> The error involved in using the MADF and NIR 8-term regression equations for the prediction of <u>in vivo</u> OMD of the 122 calibration silages was assessed to determine the effect of such error on two aspects:

- 1 The Effect on ME Calculation and
- 2 The Effect on Ration Formulation.

1 The Effect on ME Calculation

The use of MADF and NIR 8-term regression equations to predict <u>in vivo</u> OMD gave an RSD% and SEC% of 5.1 and 2.5 respectively (Table 4.1).

From the normal distribution curve (Figure 7.1), the percentage of the actual values which will be more or less than a given range from a predicted ME value of 10.4

Unit Cost of Analysis and Laboratory Turn-Around Time for Some of the Methods Tested in This Work (Dixon, Personal Communication)

Method	Unit Cost of Laboratory Analysis (£)*	Turn-Around Time (Working Days)
MADF	2.80	2-3
NCD	7.40	3-4
IVOMD	3.90**	12-15
NIR	See Below	2

* Excluding cost of sample preparation and laboratory overheads.

** Excludes the cost of maintaining surgically modified animals.

TABLE 7.3

Unit Cost of NIR Analyis*

Annual Costs	£/Annum	
 Capital Cost of Equipment £87,500 depreciated over 5 years Interest on Capital at 20% APR Labour Requirement (1 full-time 	17,500 17,500	
technician) 4 Maintenance & Insurance	10,000 5,000	
200 Samples can be scanned/day for 250 days		
= 50,000 samples/annum. Unit Cost at Full Capacity £1/sample at Half Capacity £2/sample		

* It is assumed that silage analysis is only a small proportion of the total use of the instrument.

The Normal Distribution Curve for the Prediction of Silage ME (MJ $\rm kg^{-1}$ DM).



Shaded areas are percentages of actual values which will be out by \pm 0.96 of ME from the value predicted.
MJ kg⁻¹ DM were calculated. For ME prediction, Figure 7.2 shows the percentage of the actual values which will be greater than the predicted ME value by the stated number of ME units. For example, whereas 12% of the actual values will be more than 0.96 MJ of ME from the predicted value when MADF is used as a predictor, the corresponding figure for NIR is only 1%.

These calculations assume a conversion factor of 0.16 to Therefore, the true error of calculate ME from DOMD%. these prediction will more than that since ME be calculations assume that urine and methane losses can be predicted without error.

2 The Effect on Ration Formulation

The effect of prediction error on ration formulation (and so on the advice which could be given to farmers) has been shown using the SAC dairy cow rationing software.

The amount of concentrate required for a 100 cow dairy herd during a 182 day winter feeding was calculated for a silage (ME = 10.4 MJ Kg^{-1} DM) and concentrate (ME = 13 MJ kg⁻¹ DM).

As with ME, similar calculations can be performed here. Figure 7.3 shows the percentage of cases in which overpurchase of concentrate for winter feeding will result. For example, the advice given using the MADF regression equation will result in over-purchase of concentrate of 29.5 tonnes in 12% of cases, whereas for NIR it is only 1% of cases which will result in over-purchase.

The above comparisons clearly highlights the magnitude of the prediction error when translated to real farm The use of either the MADF or NIR methods as practice. a predictor of the energy value of grass silages can have serious financial consequences to the farmer.



ME Prediction Error





Ration Formulation Error Tonnes of Concentrate for Winter Feeding (100 cows)



Assuming that the current price of one tonne of concentrate is equal to £170, this means that 12% of the farmers will needlessly pay an extra £5,015 for buying extra concentrate for winter feeding for a herd of 100 cows if the MADF method was used as a predictor of ME, whereas with NIR it is only 1% of the farmers who will pay that amount.

The work described in Chapter 5 suggested that the in OMD – laboratory method relationship be vivo can influenced by environmental and other external factors. А factor like year of harvest was found to affect the relationships based on the IVOMD and NCOMD methods. Another noticeable factor was cut number which was found to affect the relationships based on the MADF and LIGA Factors like the method of ensiling, wilting methods. time, additive application and nitrogen fertilisation were not found to affect the relationship between in vivo OMD and its predictors.

In Chapter 4, it was reported that the NIR method was the most precise predictor of in vivo OMD. In Chapter 6, Section 1, it was demonstrated that NIR can be further used to predict in vivo DOMD. However, the in vivo DOMD prediction was found to be inferior to in vivo OMD prediction. The direct prediction of in vivo DOMD using NIR was found to be simpler and more economical since ash determination is not required. However, the NIR prediction of in vivo OMD and then measuring the ash content, was found to give more precise prediction of in vivo DOMD. Additionally, the measurement of the ash content in itself is an important indicator of the extent of soil contamination in silage samples, and a useful guide to the quality of silage making techniques.

The comparison procedure performed between <u>in vivo</u> OMD predictors (Chapter 4) was repeated in Chapter 6, Section

an attempt to demonstrate the ability of 2 in the laboratory methods studied in this work to predict in vivo DE of grass silages by using a large number of The results indicated that the DE of grass silages. silages was poorly predicted by the conventional methods. This poor prediction highlighted the difficulty of the DE prediction (and similarly ME) using a simple digestibility measurement. It has been found that the GE of silages plays an important role in energy evaluation of silages and can account for much of the variation in silage DE. However, this aspect warrants further investigation. Of the methods tested in this Chapter, the NIR method proved to be the best predictor of in vivo DE.

POSSIBILITIES FOR FUTURE RESEARCH

In view of the results obtained from this thesis, the following section describes the avenues of research which are potentially important and may provide valuable and interesting results. These avenues include:

- i) The <u>in situ</u> method is recognised as being of potential value in forage evaluation. This method has been shown in this work to be as precise as the IVOMD method. However, it is felt that this method may surpass the precision of the IVOMD method if fresh minced forage was used, provided that it is used in the manner described in Chapter 3.
- ii) A method for accurate and precise in vivo ME prediction for routine silage evaluation is urgently required. Digestibility predictors were found to give poor predictions of in vivo DE (Chapter 6, Section 2) and similarly, in vivo ME (Givens et al, 1989a) of grass was found to give silages. Only NIR а significant improvement in the DE prediction. It is therefore possible that, with the help of NIR, the ME prediction can be improved by:
 - a) Direct Calibration of NIR with large <u>in vivo</u> ME populations in the manner described in Chapter 4;
 - b) Where a) is not possible, the NIR method can be of value to predict the GE of silages since the measurement of this parameter is rather difficult for grass silages. If this prediction is found to be accurate, this will then allow the establishment of a quick and cheap method for the determination of the GE of silages. A further aim of this requirement is to study the exact role of GE in the energy evaluation of silages. This stems from the fact that

GE was found to play an important role in ME prediction and can account for almost 50% of the variability in ME (Givens and Brunnen, 1987).

However, the prediction of 'correct' ME values for silage must be combined with 'correct' values for efficiency of utilisation of ME at the rationing stage and further information on this aspect is also needed.

Between-laboratory differences are an important aspect of iii) the choice of any predictive method(s). It is therefore important to investigate this problem before any predictive method(s) can be applied between laboratories. This aspect is particularly important if the NIR method is to be used on a national basis. The matching of NIR instruments and then the need for continuous monitoring of their performance is required. NIR instrument matching has already been made by other workers (Shenk et al, 1985).

REFERENCES

- ARC, 1965. The Nutrient Requirements of Farm Livestock, No. 2, Ruminants, Agriculture Research Council, London.
- ARC, 1980. The Nutrient Requirements of Ruminant Livestock. Commonwealth Agricultural Bureau, Slough.
- ADAS Advisory Paper No. 3148, (1988). Grass Silage Quality and Feeding. HMSO, UK.
- ABRAMS, S.M. (1985). Populations. In: Near Infra-red Reflectance Spectroscopy (NIRS): Analysis of Forage Quality, pp 37. Ed. G.C. Margin, J.S. Shenk and F.E. Barton II. US Department of Agriculture Handbook No. 643.

N

- ABRAMS, S.M., HARTADI, H., CHAVES, C.M., MOORE, J.E. and OCUMPAUGH, W.R. (1981). Relationship of Forage – Evaluation Techniques to the Intake and Digestibility of Tropical Grasses. Proc. XIV Int. Grassland Congr., Lexington, Kentucky, USA, pp 568-511.
- ABRAMS, S.M., SHENK, J.S., WESTERHAUS, M.O. and BARTON II, F.E. (1987). Determination of forage quality by near infrared reflectance spectroscopy: Efficiency of broadbased calibration equations. J. Dairy Sci., 70: 806-813.
- ADAMSON, A.H. and TERRY, G.R. (1980). The Relationship between the <u>in vivo</u> Digestibility of Hay and its Solubility in Pepsin-Hydrochloric Acid and Fungal Cellulase Solutions. J. Sci. Fd Agric., 31: 854-856.

- AERTS, J.V.; DeBOEVER, J.L.; COTTYN, B.G.; DeBRABANDER, D.L.; BUYSSE, F.X. (1984). Comparative Digestibility of Feedstuffs by Sheep and Cows. Anim. Feed Sci. Technol. 12: 47-56.
- AERTS. J.V., DeBRABANDER, D.L. and COTTYN, B.G. (1986). Use of Sheep as Pilot Animals for Predicting Digestibility of Feeds to be Fed to Dairy Cows. In: Recent Advances in Feed Evaluation and Rationing Systems for Dairy Cattle in Extensive and Intensive Production Systems. pp 22-25. Ed. C.C. Balch and A.J.H. van Es. International Dairy Federation Bulletin No. 196/1986.
- AERTS. J.V., DeBRABANDER, D.L., COTTYN, B.G. and BUYSSE, F.X. (1977). Comparison of Laboratory Methods for Predicting the Organic Matter Digestibility of Forages. Anim. Feed Sci. Technol., 2: 337-349.
- AERTS. J.V., DeBRABANDER, D.L., COTTYN, B.G., BUYSSE, F.X. and MOERMANS, R.J. (1974). Comparisons of Methods for Dry Matter Determination of High Moisture Roughages and Faeces. J. Sci. Fd Agric., 25: 619-.
- AKIN, D.E. (1979). Microscopic Evaluation of Forage Digestion by Rumen Micro-organisms - A Review. J. Anim. Sci., 48: 701-710.
- ALDERMAN, G., COLLINS, F.C. and DOUGALL, H.W. (1971). Laboratory Methods of Predicting Feeding Value of Silage. J. Br. Grassland Soc., 26: 109-111.
- ALDRICH, D.T.A. and DENT, J.W. (1967). The Relationship between Yield and Digestibility in the Primary Growth of Nine Grass Varieties. J. Natl. Inst. Agr. Botany. 11: 104-113.

- ALEXANDER, R.H. and MCGOWAN, M. (1966). The Routine Determination of <u>in vitro</u> Digestibility of Organic Matter in Forages - An Investigation of the Problems Associated with Continuous Large Scale Operation. J. Br. Grassland Soc., 21: 140-147.
- ALEXANDER, R.H. and MCGOWAN, M. (1969). The Assessment of the Nutritive Value of Silage by Determination of <u>in vitro</u> Digestibility on Homogenates Prepared from Fresh Undried Silage. J. Br. Grassland Soc., 195–198.
- ASPLUND, J.M., BERG, R.T., MCELROY, L.W. and PIDGEN, W.J. (1958). Dry Matter Loss and Volatile Fatty Acid Production in the Artificial Rumen as Indices of Forage Quality. Can. J. Anim. Sci., 38: 171-180.
- BALCH, C.C. and JOHNSON, V.W. (1950). Factors Affecting the Utilisation of Breakdown of Cellulose (Cotton Thread). Br. J. Nutr., 4: 389-395.
- BARBER, G.D., OFFER, N.W. and CASTLE, M.E. (1983). An Assessment of Regression Equations for the Prediction of Silage ME. Proc. ADAS Nutrition Chemists Training Conf. Cambridge.
- BARBER, W.P., ADAMSON, A.H. and ALTMAN, J.F.B. (1984). New Methods of Forage Evaluation. In: Recent Advances in Animal Nutrition, pp 161-176. Ed. W. Haresign and D.J.A. Cole, Butterworths, London.
- BARBER, W.P., BARBER, G.D. and OFFER, N.W. (1980). Predicting the Energy Potential of Grass Silage. Proc. COSAC 4th Study Conf., Perth.
- BARNES, R.F. (1973). Laboratory Methods of Evaluating Feeding Value of Herbage. In: Chemistry and Biochemistry of Herbage, Vol. 3, pp 179-214. Ed. G.W. Butler and R.W. Bailey. Academic Press, London.

- BARTON II, F.E. and BURDICK, D. (1981). Prediction of Forage Quality with NIR Reflectance Spectroscopy. Proc. XIV Int. Grassland Congr., Lexington, Kentucky, USA, pp 532-534.
- BLAXTER, K.L. (1962). The Energy Metabolism of Ruminants. Hutchinson, London.
- BLAXTER, K.L. (1964). Utilisation of the Metabolisable Energy of Grass. Proc. Nutr. Soci., 23: 62-71.
- BLAXTER, K.L. and CLAPPERTON, J.L. (1965). Prediction of the Amount of Methane Produced by Ruminants. Br. J. Nutr., 19: 511-522.
- BLAXTER, K.L., CLAPPERTON, J.L. and WAINMAN, F.W. (1966). The Extent of Differences between Six British Breeds of Sheep in their Metabolism, Feed Intake and Utilisation, and Resistance to Climatic Stress. Br. J. Nutr., 20: 283-294.
- BLAXTER, K.L., GRAHAM, N. McC. and ROOK, J.A.F. (1954). Apparatus for the Determination of the Energy Exchange of Calves and of Sheep. J. Agric. Sci., 45: 10-18.
- BLAXTER, K.L., GRAHAM, N. M. and WAINMAN, F.W. (1956). Some Observations on the Digestibility of Food by Sheep and Related Problems. Br. J. Nutr., 10: 69-91.
- BRATZLER, J.W. and FORBES, E.B. (1940). The Estimation of Methane Production by Cattle. J. Nutr., 19: 611-613.
- BROWN, D.C. and RADCLIFF, J.C. (1971). Prediction of <u>in vivo</u> dry matter, organic matter and energy digestibilities of silage by <u>in vitro</u> digestion techniques. Aust. J. Agric. Res., 27: 787-796.

- BROWN, L.D. (1966). Influence of Intake on Feed Utilisation. J. Dairy Sci., 49: 223-230.
- BROWN, W.F. and MOORE, J.E. (1987). Analysis of forage research samples utilising a combination of wet chemistry and near infrared reflectance spectroscopy. J. Anim. Sci., 64: 271-282.
- BYERS, F.M., JOHNSON, D.E. and MATSUSHIMA, J.K. (1976). In: Energy Metabolism of Farm Animals. pp 253-256. Ed. M. Vermorel, EAAP Publ. No. 19.
- CASTLE, M.E. (1972). Acid Additives for Silage Making. Scot. Agric., 51: 302-304.
- CASTLE, M.E. (1982). Feeding High Quality Silage. In: Silage for Milk Production, pp 127-150. Ed. J.A.F. Rook and P.C. Thomas, Technical Bulletin No. 2. National Institute for Research in Dairying, Reading.
- CHENOST, M., GRENET, E., DERMARQUILLY, C. and JARRIGE, R. (1970). The Use of Nylon Bag Technique for the Study of Forage Digestion in the Rumen and for Predicting Feed Value. Proc XI Int. Grassland Congr., Queensland, pp 697-701.
- CLANCY, M.J. and WILSON, R.K. (1966). Development and Application of a New Chemical Method for Predicting the Digestibility and Intake of Herbage Samples. Proc. Xth Inter. Grassland Congr., Helsinki, pp 445-453.
- CLARK, J. and BEARD, J. (1977). Prediction of the Digestibility of Ruminant Feeds from Their Solubility in Enzyme Solutions. Anim. Feed Sci., Technol., 2: 153-159.

COELHO, M., HEMBRY, F.G., BARTON, F.E. and SAXTON, A.M. (1988). A Comparison of Microbial, Enzymatic, Chemical and Near Infrared Reflectance Spectroscopy Methods in Forage Evaluation. Anim. Feed Sci. Technol., 20: 219-231.

CONNOLLY, J.G. (1984). MSc Thesis, University of Aberdeen.

- COTTYN, B.G., AERTS, J.V. and VANACKER, J.M. (1986). Use of New Chemical and Biochemical Determinations in Predicting Digestibility and Energy Value. In: Recent Advances in Feed Evaluation and Rationing Systems for Dairy Cattle in Extensive and Intensive Production Systems. pp 28-32. Ed. C.C. Balch and A.J.H. van Es. International Dairy Federation Bulletin No. 196/1986.
- DEBOEVER, J.L., COTTYN, B.G., ANDRIES, J.I., BUYSSE, F.X. and VANACKER, J.M. (1988). The Use of a Cellulase Technique to Predict Digestibility, Metabolisable and Net Energy of Forages. Anim. Feed Sci. Technol., 19: 247-260.
- DEINUM, B., van Es. A.J.H. and VAN SOEST, P.J. (1968). Climate, Nitrogen and Grass, II. The Influence of Light Intensity, Temperature and Nitrogen on <u>in vivo</u> Digestibility of Grass and the Prediction of these Effects from some Chemical Procedures. Neth. J. Agric. Sci., 16: 217-223.
- DEWAR, W.A. and McDONALD, P. (1961). Determination of Dry Matter in Silage by Distillation with Toluene. J. Sci. Fd Agric., 12: 790-795.
- DONALDSON, E. and EDWARDS, R.A. (1976). Feeding Value of Silage: Silages Made from Freshly Cut Grass, Wilted Grass and Formic Acid Treated Wilted Grass. J. Sci. Food Agric., 27: 536-544.

- DONALDSON, E. and EDWARDS, R.A. (1977). Feeding Value of Wilted Silages Made Using Formic Acid, Formaldehyde and Propionic Acid. Anim. Prod., 25: 71-81.
- DONALDSON, E. and EDWARDS, R.A. (1979). Metabolism Within the Rumen of Silages Made with Formaldehyde. In: Forage Conservation in the 80's, pp 350-355. Ed. C. Thomas. Occasional Symposium No. 11. British Grassland Society, Maidenhead.
- DONALDSON, E., OFFER, N.W. and THOMSON, J.K. (1980). The Degradability of Feed Proteins in the Rumen. 4th Study Conf. of the Scottish Agricultural Colleges, Perth.
- DONEFER, E., NIEMANN, P.J., CRAMPTON, E.W. and LLOYD, L.E. (1963). Dry matter disappearance by enzyme and aqueous solutions to predict the nutritive value of forages. J. Dairy Sci., 46: 965-970.
- DOWMAN, M.G. and COLLINS, E.C. (1982). The Use of Enzymes to Predict the Digestibility of Animal Feeds. J. Sci. Fd Agric., 33: 689-696.
- DOWNEY, G., ROBERT, P., BERTRAND, D. and FRANCOISE DEVANY, M. (1987). Near Infrared Analysis of Grass Silage by Principal Component Analysis of Transformed Reflectance Data. J. Sci. Fd Agric., 41: 219-229.
- ECKMAN, D.D., SHENK, J.S., WANGSNESS, P.J. and WESTERHAUS, M.O. (1983). Prediction of Sheep Responses by Near Infrared Reflectance Spectroscopy. J. Dairy Sci., 66: 1983-1987.
- EDWARDS, R.A. (1986). The Energy Value of Grass Silage. In: Ruminant Feed Evaluation, pp 35-58. Ed. B.A. Stark, D.I. Givens and M.J. Wilkinson. Chalcombe Publications, Marlow.

- EGAN, H., KIRK, R. and SAWYER, R. (1981). In: Chemical Analysis of Food. Churchill Livingstone, Edinburgh.
- EKERN, A. and SUNDSTØL, F. (1974). Energy Utilisation of Hay and Silages by Sheep. In: Energy Metabolism of Farm Animals, pp 221-224. Ed. K.H. Menke, H.J. Lantzsch and J.R. Reichl, EAAP Publication No. 14.
- EL KHIDIR, O.A. and VESTERGAARD THOMSEN, K. (1983). Effect of Intake on Digestibility of Plant Cell Walls and Cell Contents of Complete Diets for Ruminants. Anim. Feed Sci. Technol., 9: 197-204.
- ERWIN, E.S. and ELLISTON, N.G. (1959). Rapid Method of Determining Digestibility of Concentrates and Roughages in Cattle. J. Anim. Sci., 18: 1518 (Abst).
- EVANS, P.J. and COTTRIL, B.R. (1984). Collaborative Ring Test to Estimate Protein Degradability. Anim. Prod., 38: 547.
- FLANNERY, K.V. (1965). The Ecology of Early Food Production. Science, 147: 1247-1256.
- FRY, G. (1885). Sweet Ensilage, Agriculture Press, London.
- GENSTAT (1983). A General Statistical Program. Rothamstead Experimental Station, Harpenden, Hertfordshire, UK.
- GIVENS, D.I. (1986). New Methods for Predicting the Nutritive Value of Silage. In: Developments in Silage. pp 66-75. Ed. B. Stark and M. Wilkinson. Chalcome Publications, Marlow.
- GIVENS, D.I. and BRUNNEN, J.M. (1987). Factors Influencing the Gross Energy Concentration of Grass Silage. Anim. Prod., 44: 495 (Abs).

- GIVENS, D.I., EVERINGTON, JEANNIE M. and ADAMSON, A.H. (1989a). The Digestibility and ME Content of Grass Silages and Their Prediction from Laboratory Measurements. Anim. Feed Sci. Technol., in press.
- GIVENS, D.I., EVERINGTON, JEANNIE M. and ADAMSON, A.H. (1989b). The Nutritive Value of Spring Grown Herbage Produced on Farms through England and Wales over Four Years. II. The Prediction of Apparent Digestibility <u>in</u> <u>vivo</u> from Various Laboratory Measurements. Anim. Feed Sci. Technol., in press.
- GOFFART, A. (1877). Manual of the Culture and Ensilage of Maize and Other Fodder Crops. Masson, Paris.
- GOTO, I. and MINSON, D.J. (1977). Prediction of the Dry Matter Digestibility of Tropical Grasses Using a Pepsin-Cellulase Assay. Anim. Feed Sci. Technol., 2: 247-253.
- HAIGH, P.M. and HOPKINS, J.R. (1977). Relationship between Oven and Toluene Dry Matter in Grass Silages. J. Sci. Fd Agric., 28: 477-480.
- HARRIS, C.E. and RAYMOND, W.F. (1963). The Effect of Ensiling on Crop Digestibility. J. Br. Grassld Soc., 18: 204-212.
- HARRIS, C.E., RAYMOND, W.F. and WILSON, R.F. (1966). The Voluntary Intake of Silage. Proc. 10th Inter. Grassland Congr., Helsinki, pp 564-567.
- HARTLEY, R.D., JONES, E.C. and FENLON, J.S. (1974). Prediction of the Digestibility of Forages by Treatment of Their Cell Walls with Cellulotic Enzymes. J. Sci. Fd Agric., 25: 947-954.

HELLIWELL, S.H. (1986). MSc Thesis, University of Aberdeen.

- HENDERSON, D.R. (1987). Silage Making: Biotechnology on the Farm. Outl. Agric., 16: 89-94.
- HOPSON, J.D., JOHNSON, R.R. and DEHORITY, B.A. (1963). Evaluation of the Dacron Bag Technique as a Method for Measuring Cellulose Digestibility and Rate of Forage Digestion. J. Anim. Sci., 22: 448-453.
- HYMOWITZ, T., DUDLEY, J.W., COLLINS, F.I. and BROWN, C.M. (1974). Estimations of Protein and Oil Concentration in Corn, Soyabean and Oat Seed by Near-infrared Light Reflectance. Crop Sci., 14: 713-715.
- С. JARRIGE. R., THIVEND, P. and DERMARQUILLY, $\{1970\}.$ Development of а Cellulolytic Enzyme Digestion for Predicting the Nutritive Value of Forages. Proc. XIth Intr. Grassland Congr., Queensland, pp 762-766.
- JONES, D.I.H. (1986). Cellulase Based Methods for Evaluating Forage Quality. In: Recent Advances in Feed Evaluation and Rationing Systems for Dairy Cattle in Extensive and Intensive Production Systems. pp 33-35. Ed. C.C. Balch and A.J.H. Van Es. International Dairy Federation Bulletin No. 196/1986.
- JONES, D.I.H. and HAYWARD, M.V. (1973). A Cellulase Digestion Technique for Predicting the Dry Matter Digestibility of Grasses. J. Sci. Fd. Agric., 24: 1419-1426.
- JONES, D.I.H. and HAYWARD, M.V. (1975). The Effect of Pepsin Pretreatment of Herbage on the Prediction of Dry Matter Digestibility from Solubility in Fungal Cellulase Solutions. J. Sci. Fd Agric., 26: 711-718.
- KAYE, W. (1954). Near Infrared Spectroscopy. 1. Spectral Identification and Analytical Applications (A Review). Spectrochim Acta. 6: 257.

- KELLNER, O. and KOHLER, A. (1900). Untersuchungen uber den Stott und Energieumsatz der erwachsenen Rind bei Erhaltungs und Produktions futter. Landw. Versuchs. Stationen, 53, 1.
- KELLY, N.C. and THOMAS, P.C. (1978). The Nutritive Value of Silages: Energy Metabolism in Sheep Receiving Diets of Grass Silage or Grass Silage and Barley. Br. J. Nutr., 40: 205-219.
- KIVIMAE, A. (1960). Estimation of the Digestibility of Grassland Crops from their Chemical Composition. Proceedings of the 8th Intr. Congr., Reading, England, UK, pp 466-470.
- KRIDIS, M.S., BARBER, G.D., OFFER, N.W., MURRAY, I., GIVENS, D.I. and DIXON, J. (1987). Prediction of Organic Matter Digestibility of Grass Silages, Recent Developments. Proc. 8th Silage Conf., Hurley, pp 53-54.
- KRIDIS, M.S., OFFER, N.W. and BARBER, G.D. (1989). The Effect of Different Washing Procedures on the Losses of OM and N from Samples of Hay Incubated in Nylon Bags Within the Rumen of Sheep. New Techniques in Cattle Production. Ed. C Phillips, Butterworths. In Press.
- KRUGGLE, W.G., FIELD, R.A., RILEY, M.L., RADOLF, H.D. and HORTON, K.M. (1981). Near-infrared Reflectance Determination of Fat, Protein and Moisture in Fresh Meat. J. Assoc. of Anal. Chem., 64: 692-696.
- LINDBERG, J.E. (1981a). The Effect of Sample Size and Sample Structure on the Degradation of Dry Matter, Nitrogen and Cell Walls in Nylon Bags. Swedish J. Agric. Res., 11: 71-76.

.

- LINDBERG, J.E. (1981b). The Effect of Basal Diet on the Ruminal Degradation of Dry Matter, Nitrogenous Compounds and Cell Walls in Nylon Bags. Swedish J. Agric. Res., 11: 159-169.
- LINDBERG, J.E. and KNUTSSON, P.G. (1981). Effect of Bag Pore Size on the Loss of Particulate Matter and on the Degradation of Cell Wall Fibre. Agric. Envirom., 6: 171-182.
- LINDBERG, J.E. and VARVIKKA, T. (1982). The Effect of Bag Pore Size on the Ruminal Degradation of Dry Matter, Nitrogenous Compounds and Cell Walls in Nylon Bags. Swedish J. Agric. Res., 12: 163-171.
- LUSK, J.W., BROWNING, C.B. and MILLES, J.T. (1962). Small Sample <u>in vivo</u> Cellulose Digestion Procedure for Forage Evaluation. J. Dairy Sci., 45: 69-73.
- MAFF (1975). Energy Allowances and Feeding Systems for Ruminants. Technical Bulletin 33. HMSO, London.
- MARSH, R. (1979). The Effect of Wilting on Fermentation in the Silo and on the Nutritive Value of Silage. Grass and Forage Science, 34: 1-9.
- MATHERS, J.C. and AITCHISON, E.M. (1981). Direct Estimation of the Extent of Contamination of Food Residues by Microbial Matter after Incubation within Synthetic Fibre Bags in the Rumen. J. Agric. Sci., Camb., 96: 691-693.
- MCNALLY, R.A. (1942). Digestion of Straw by the Ruminant. Biochem. J., 36: 392-399.
- MCDONALD, P. (1983). The Control of Silage Fermentation. Hannah Research Institute Report 1983, pp 59-67.

- MCDONALD, P. and EDWARDS, R.A. (1976). The Influence of Conservation Methods on Digestion and Utilisation of Forages by Ruminants. Proc. Nutr. Soc., 35: 201-211.
- MCDONALD, P. and WHITTENBURY, R. (1973). The Ensilage Process, Vol. 3. In: Chemistry and Biochemistry of Herbage. Ed. G.W. Butler and R.W. Bailey, pp 33-60. Academic Press, London.
- MCLAREN, G.A., ASPLUND, R.O., CROW, D.G., ISAI, L.I. and PORTERFIELD, I.D. (1964). Influence of Dried Grass Silage and Silage Fractions on the Basal Metabolic Rate of Rats. N. Nutr., 83: 218-224.
- MCLEOD, M.N. and MINSON, D.J. (1971). The Error in Predicting Pasture Dry Matter Digestibility from Four Different Methods of Analysis for Lignin. J. Br. Grassland Soc., 26: 251-256.
- MCLEOD, M.N. and MINSON, D.J. (1978). The Accuracy of the Pepsin-Cellulase Technique for Estimating the Dry Matter Digestibility <u>in vivo</u> of Grasses and Legumes. Anim. Feed Sci. Technol., 3: 277-287.
- MCLEOD, M.N. and MINSON, D.J. (1982). The Accuracy of Predicting Digestibility by the Cellulase Technique; the Effect of Pre-treatment of Forage Samples with Neutral Detergent or Acid Pepsin. Anim. Feed Sci. Technol., 7: 83-92.
- MCMANUS, W.R., MANTA, L., MCFARLANE, J.D. and GRAY, A.C. (1972). The Effects of Diet Supplements and Gamma Irradiation on Dissimilation of Low Quality Roughage by Ruminant. I. Studies on Terylene-Bag Technique and Effects of Supplementation of Basal Ration. J. Agric. Sci., Camb., 79: 27-40.

MCQUEEN, R.E. (1986). Laboratory Evaluation of Forage Quality. In: Recent Advances in Feed Evaluation and Rationing Systems for Dairy Cattle in Extensive and Intensive Production Systems, pp 42-46. Ed. C.C. Balch and A.J.H. Van Es. International Dairy Federation Bulletin No. 196/1986.

۰.

- MCQUEEN, R. and VAN SOEST, P.J. (1975). Fungal Cellulase and Hemicellulase Prediction of Forage Digestibility. J. Dairy Sci., 58: 1482-1491.
- MEHREZ, A.Z. and ØRSKOV, E.R. (1977). A study of the artificial fibre bag technique for determining the digestibility of feeds in the rumen. J. Agric. Sci., Camb., 88: 645-650.
- MINSON, D.J. (1982). Effect of Chemical Composition on Feed Digestibility and Metabolisable Energy. Nutr. Abs. Rev., 52: 591-615.
- MINSON, D.J., BUTLER, K.L., GRUMMITT, N. and LAW, D.P. (1983). Bias when predicting crude protein, dry matter digestibility and voluntary intake of tropical grasses by near infrared reflectance. Anim. Feed Sci. Technol., 9: 221-237.
- MINSON, D.J. and LANCASTER, R.J. (1963). The Effect of Oven Temperature on the Error in Estimating the Dry Matter Content of Silage. N.Z.J. Agric. Res., 6: 140-146.
- MORGAN, D.J. and STAKELUM, G. (1987). The Prediction of the Digestibility of Herbage for Dairy Cows. Ir. J. Agric. Res., 26: 23-34.
- MORRISON, I.M. (1972). A semi-micro Method for the Determination of Lignin and its Use in Predicting the Digestibility of Forage Crops. J. Sci. Fd Agric., 23: 455-463.

- MORRISON, I.M. (1973). A Note on the Evaluation of the Nutritive Value of Forage Crops by the Acetyl Bromide Technique. J. Br. Grassland Soc., 28: 153-154.
- MURRAY, I. (1986). Near infrared Reflectance Analysis of Forages. In: Recent Advances in Animal Nutrition, pp 141-156. Ed. W. Haresign and D.J.A. Cole. Butterworths, London.
- MURRAY, I., KRIDIS, M.S., BARBER, G.D., GIVENS, D.I. and OFFER, N.W. (1987). Grass Silage 'In Vivo' Digestibility Prediction by NIR. CEC Monograph, Brussels, 3-4 December 1987.
- NORRIS, K.H. (1985). NIRS Instrumentation. In: Near Infra-red Reflectance Spectroscopy (NIRS): Analysis of Forage Quality, pp 12-17. Ed. G.C. Martin, J.S. Shenk and F.E. Barton II. US Department of Agriculture Handbook No. 643.
- NORRIS, K.H., BARNES, R.F., MOORE, J.G. and SHENK, J.S. (1976). Predicting forage quality by infrared reflectance spectroscopy. J. Anim. Sci., 43: 889-897.
- OH, H.K., BAUMGARDT, B.R. and SCHOLL, J.M. (1966). Evaluation of Forages in the Laboratory. V. Comparison of Chemical Analysis, Solubility Tests, and <u>in vitro</u> Fermentation. J. Dairy Sci., 49: 850-855.
- O'KEEFE, M., DOWNEY, G. and BORGAN, J.C. (1987). The Use of Near Infrared Reflectance Spectroscopy for Predicting the Quality of Grass Silage. J. Sci. Food Agric., 38: 209-216.
- ØRSKOV, E.R., HINE, R.S. and GRUBB, D.A. (1978). The Effect of Urea on Digestion and Voluntary Intake by Sheep of Diets Supplemented with Fat. Anim. Prod., 27: 241-245.

- ØRSKOV, E.R., HOVELL, F.D.DEB. and MOULD, F. (1980). The Use of Nylon Bag Technique for the Evaluation of Feedstuffs. Trop. Anim. Prod. 5: 195-213.
- (1979).The Estimation of and MCDONALD, Ι. ØRSKOV, E.R. Rumen Protein Degradability in the from Incubation Measurements Weighted According to Rate of Passage. J. Agric. Sci., Camb., 92: 499-503.
- ØRSKOV, E.R., REID, G.W. and KAY, M. (1988). Prediction of Intake by Cattle from Degradation Characteristics of Roughages. Anim. Prod., 46: 29-34.
- OSBOURN, D.F. (1978). Principles Governing the Use of Chemical Methods for Assessing the Nutritive Value of Forages. A Review. Anim. Feed Sci. Technol., 3: 265-275.
- OSBOURN, D.F. (1980). In Grass, its Production and Utilisation, p 70. Ed. W. Holmes, Blackwell, London.
- O'SHEA, J., WILSON, R.K. and SHEELAN, W. (1972). Prediction of Silage Digestibility by <u>in vitro</u> and Chemical Methods. Ir. J. Agri. Res., 11: 175-179.
- PLAYNE, M.J., KHUMNUALTHANG, W. and ECHEVARRIA, M.G. (1978). Factors Affecting the Digestion of Oesophageal Fistula Samples and Hay Samples in Nylon Bags in the Rumen of Cattle. J. Agric. Sci., Camb., 90: 193-204.
- QASIM, S.A. and STELZING, D.A. (1973). Stimulation of Basal Metabolic Rate of Rats Fed Dried Grass Silage and Silage Flavoniods. J. Nutri., 103: 1658-1664.
- QUINN, J.I., VAN DER WATH, J.G. and MYBURGH, S. (1938). Studies on the Alimentary Tract of Merino Sheep in South Africa. IV. Description of Experimental Technique. Onderstepoort J. Vet. Sci. Animal Ind., 11: 341.

RAYMOND, W.F. (1969). The Nutritive Value of Forage Crops. Advan. Agron., 21: 1-108.

REW, R.H. (1888). Stack Ensilage, Walter Scott, London.

- ROBERT, P., BERTRAND, D. and DERMARQUILLY, C. (1986). Prediction of Forage Digestibility by Principal Component Analysis of Near Infrared Reflectance Spectra. Anim. Feed Sci. Technol., 16: 215-224.
- RODRIGUEZ, H. (1968). <u>In vivo</u> Bag Digestibility: the Relative Position within the Rumen. Rev. Cubana Cienc. Agric., 2: 285-287.
- ROOK, J.A., AKINSOYINU, A.O. and ARMSTRONG, D.G. (1983). The Release of Mineral Elements from Grass Silages Incubated <u>in sacco</u> in the Rumen of Jersey Cattle. Grass and Forage Science, 38: 311-316.
- ROUGHAM, P.G. and HOLLAND, R. (1977). Predicting <u>in vivo</u> Digestibilities of Herbages by Exhaustive Enzymic Hydrolysis of Cell Walls. J. Sci. Fd Agric., 28: 1057-1064.
- ROWETT RESEARCH INSTITUTE, (1975). Feedingstuffs Evaluation Unit, First Report. Department of Agriculture and Fisheries for Scotland, Edinburgh, pp 29-33.
- ROWETT RESEARCH INSTITUTE, (1983). Chemistry and Biochemistry of Plant Tissues. In: Annual Report of Studies in Animal Nutrition and Allied Sciences, pp 33-36, Vol. 39.
- RYAN, B.F., JANIER, B.L and RYAN, T.A.J. (1985). Minitab Handbook, Second Edition. Duxbury Press, Boston.

- SAVITSKY, A. and GOLAY, M.J.E. (1964). Smoothing and Differentiation of Data by Simplified Least Squares Procedures. Anal. Chem., 36: 1627-1639.
- SCHNEIDER, B.H. and FLATT, W.P. (1975). The Evaluation of Feeds through Digestibility Experiments. Athens: University of Georgia Press.
- SCHOEMAN, E.A., DE WET, P.J. and BURGER, W.J. (1972). The Evaluation of the Digestibility of Treated Proteins. Agroanimalia, 4: 35-46.
- SCHUKKING, S. (1976). The History of Silage Making. Stikstof, 19: 2-11.
- SHENK, J.S. (1986). Near Infrared Analysis of Forage and Feed for Dairy Cows. In: Recent Advances in Feed Evaluation and Rationing Systems for Dairy Cattle in Extensive and Intensive Production Systems, pp 36-41. Ed. C.C. Balch and A.J.H. Van Es. International Dairy Federation Bulletin No. 196/1986.
- SHENK, J.S., LANDA, I., HOOVER, M.R. and WESTERHAUS, M.O. (1981). Description and Evaluation of a Near Infrared Reflectance Spectro-Computer for Forage and Grain Analysis. Crop Sci., 21: 355-358.
- SHENK, J.S., HOOVER, M.R. and WESTERHAUS, M.O. (1979). Analysis of Forages by Infrared Reflectance. J. Dairy Sci., 62: 807-812.
- SHENK, J.S., WESTERHAUS, M.O. and TEMPLETON, W.C. (1985). Calibration Transfer Between Near Infrared Reflectance Spectrophotometers. Crop Sci., 25: 159-161.
- SIDDIGI, M. (1985). MSc Thesis, University of Aberdeen.

- SILVA, A.T. and ØRSKOV, E.R. (1988). The Effect of Five Different Supplements on the Degradation of Straw in Sheep Given Untreated Barley Straw. Anim. Feed Sci. Technol., 19: 289-298.
- SULLIVAN, J.T. (1959). A Rapid Method for the Determination of Acid-Insoluble Lignin in Forages and its Relation to Digestibility. J. Anim. Sci., 18: 1292-1298.
- SUNDSTØL, F. and EKERN, A. (1976). The Nutritive Value of Frozen, Dried and Ensiled Grass Cut at Three Different Stages of Growth. In: Energy Metabolism of Farm Animals, pp 241-244. Ed. M. Vermorel. EAAP Publication No. 19.
- SWIFT, R.W. (1957). The Caloric Value of TDN. J. Anim. Sci., 16: 753-756.
- SWIFT, R.W., BRATZLER, T.W., JAMES, W.H., TILLMAN, A.D. and MEEK (1948). The Effect of Dietary Fat on Utilisation of the Energy and Protein of Rations by Sheep. J. Anim. Sci., 7: 475-485.
- TERRY, R.A., MUNDELL, D.C. and OSBOURN, D.F. (1978). Comparison of two <u>in vitro</u> procedures using rumen liquorpepsin or pepsin-cellulase for prediction of forage digestibility. J. Br. Grassld Soc., 33: 13-18.
- THOMAS, C. and THOMAS, P.C. (1985). Factors Affecting the Nutritive Value of Grass Silage. In: Recent Advances in Animal Nutrition, pp 223-256. Ed. W. Haresign and D.J.A. Cole. Butterworths, London.
- THOMAS, P.C. and CHAMBERLAIN, D.G. (1982). Silage as a Foodstuff. In: Silage for Milk Production, pp 63-102. Ed. J.A.F. Rook and P.C. Thomas, Technical Bulletin No. 2. National Institute for Research and Dairying, Reading.

- TILLEY, J.M.A., DERIAZ, R.F. and TERRY, R.A. (1960). The in vitro Measurement of Herbage Digestibility and Assessment of Nutritive Value. Proc. of the 8th Intr. Grassland Congr., Reading, England, pp 533-537.
- TILLEY, J.M.A. and TERRY, R.A. (1963). A Two-Stage Technique for the <u>in vitro</u> Digestion of Forage Crops. J. Br. Grassland Soc., 18: 104-111.
- UDEN, P., PARRA, R. and VAN SOEST, P.J. (1974). Factors Influencing Reliability of the Nylong Bag Technique. J. Dairy Sci., 57: 622 (Abst).
- ULYATT, M.J. (1973). The Feeding Value of Herbage, Vol. 3. In: Chemistry and Biochemistry of Herbage. Ed. G.W. Butler and R.W. Bailey, pp 131-174. Academic Press, London.
- UNSWORTH, E.F. and GORDON, F.J. (1985). The Energy Utilisation of Wilted and Unwilted Grass Silages by Lactating Dairy Cows. 58th Annual Report of the Agricultural Research Institute of Northern Ireland, pp 13-20.
- VAN DYNE, G.M. (1962). Methods for Nutritive Evaluation of Range Forages. J. Range. Mgmt., 15: 303-314.
- VAN KEUREN, R.W. and HEINEMANN, W.W. (1962). Study of a Nylon Bag Technique for in vivo Estimation of Forage Digestibility. J. Anim. Sci., 21: 340-345.
- VAN SOEST, P.J. (1963). Use of Detergents in the Analysis of Fibrous Feeds. II. A Rapid Method for the Determination of Fibre and Lignin. J. Assoc. of Agric. Chem., 46: 829-835.

- VAN SOEST, P.J. (1967). Development of a Comprehensive System of Feed Analysis and Its Application to Forages. J. Anim. Sci., 26: 119-128.
- VAN SOEST, P.J. (1982). Nutritional Ecology of the Ruminant. OCB Books, Inc., 1215 NW Klive Place, Corvallis, Oregon, 97330, USA.
- VAN SOEST, P.J. and WINE, R.H. (1967). Use of Detergents on the Analysis of Fibrous Feeds. IV. Determination of Plant Cell Wall Constituents. J. Assoc. of Anal. Chem., 50: 50-55.
- VAN SOEST, P.J. and WINE, R.H. (1968). Determination of Lignin and Cellulose in Acid Detergent Fibre with Permanganate. J. Assoc. of Anal. Chem., 51: 780-785.
- VAN SOEST, P.J., WINE, R.H. and MOORE, L.A. (1966). Estimation of the True Digestibility of Forages by the <u>in</u> <u>vitro</u> Digestion of Cell Walls. Proc. Xth Int. Grassland Congr., Helsinki, pp 438-441.
- WAINMAN, F.W. (1977). Digestibility and Balance in Ruminants. Proc. Nutr. Soc., 36: 195-202.
- WAINMAN, F.W., DEWEY, P.J.S. and BOYNE, A.W. (1978). Rowett Research Institute, Feedingstuffs Evaluation Unit, Second Report, Department of Agriculture and Fisheries for Scotland, Edinburgh, pp 12-15.
- WAINMAN, F.W., DEWEY, P.J.S. and BREWER, A.C. (1984). Rowett Research Institute, Feedingstuffs Evaluation Unit, Fourth Report, Department of Agriculture and Fisheries for Scotland, Edinburgh, pp 3-5.

- WAINMAN, F.W., DEWEY, P.J.S. and MCDONALD, M.S. (1976). The Metabolisable Energy of Cereals Fed with Dry Fodder or Silages. In: Energy Metabolism of Farm Animals, pp 229-232. Ed. M. Vermorel. EAAP Publ. No. 19.
- WAINMAN, F.W., SMITH, J.S. and DEWEY, P.J.S. (1979). The Predicted and Observed Metabolisable Energy Values of Mixtures of Maize Silage and Barley Fed to Cattle, pp 55-58. In: Energy Metabolism. Ed. L.E. Mount. EAAP Publ. No. 26.
- WALDO, D.R. (1978). In: Fermentation of Silage. A Review. Ed. M.E. McCullough. National Feed Ingredients Association, Iowa, pp 120–179.
- WESTERHAUS, M.O. (1985a). Equation Development. In: Near Infrared Reflectance Spectroscopy (NIRS): Analysis of Forage Quality, p 38-39. Ed. G.C. Martin, J.S. Shenk and F.E. Barton II. US Department of Agriculture Handbook No. 643.
- WESTERHAUS, M.O. (1985b). Validation. In: Near Infrared Reflectance Spectroscopy (NIRS): Analysis of Forage Quality, p 40. Ed. G.C. Martin, J.S. Shenk and F.E. Barton II. US Department of Agriculture Handbook No. 643.
- WESTERHAUS, м.о. (1985c). Interpretation of Regression Near Infrared Reflectance Spectroscopy Statistics. In: (NIRS): Analysis of Forage Quality, pp 39-40. Ed. G.C. Martin, J.S. Shenk and F.E. Barton II. US Department of Agriculture Handbook No. 643.
- WILKINS, R.J. (1974). The Nutritive Value of Silages. University of Nottingham Nutrition Conference for Feed Manufacturers:
 8. Ed. H. Swan and D. Lewis, pp 167–189. Butterworths, London.

- WILKINS, R.J. (1981). The nutritive value of silages. In: Recent developments in ruminant nutrition, pp 268-281. Ed. W. Harresign and D.J.A. Cole.
- WILKINS, R.J., WILSON, R.F. and COOK, J.E. (1974). Restriction of Fermentation During Ensilage: The Nutritive Value of Silages Made with the Addition of Formaldehyde. Proc. 12th Inter. Grasslnd Congr., Moscow, pp 674-690.
- WILKINSON, J.M. (1980). Production of Hay and Silage in Europe: Estimates for the Countries of the EEC and Scandinavia (1978-1979). Occasional Symposium of the British Grassland Society. No. 11, Brighton, 1979, p 457.
- WILKINSON, J.M. (1985). Beef Production from Silage and other Conserved Forages. Longmans Handbooks in Agriculture, London.
- WILKINSON, J.M., WILSON, R.F. and BARRY, T.N. (1976). Factors Affecting the Nutritive Value of Silage. Outl. Agric., 9: pp 3-8.
- WILSON, R.F. and WILKINS, R.J. (1973). Formic Acid as a Silage Additive for Wet Crops of Cocksfoot and Lucerne. J. Agric. Sci., Camb., 80: 225-231.
- WINDHAM, W.R. and COLEMAN, S.W. (1985). Animal Response Prediction. In: Near Infrared Reflectance Spectroscopy (NIRS): Analysis of Forage Quality, pp 32-35. Ed. G.C. Martin, J.S. Shenk and F.E. Barton II. US Department of Agriculture Handbook, No. 643.
- WOOD, T.B. (1917). Composition and Nutritive Value of Feedingstuffs, 1st Ed., Cambridge University Pitt Press.
- WOOLFORD, M.K. (1984). The Silage Fermentation. Marcel Dekker, New York.



.

-, . .

PUBLICATIONS

- BARBER, G.D., GIVENS, D.I., KRIDIS, M.S., OFFER, N.W. and MURRAY, I. (1988). Prediction of the organic matter digestibility of grass silage. Anim. Feed Sci. Technol. In press.
- KRIDIS, M., BARBER, G.D., OFFER, N.W., GIVENS, D.I. and organic MURRAY. Ι. (1989).Prediction of matter digestibility in vivo of grass silage. 2nd International 'New Symposium on new techniques in Agriculture Techniques in Cattle Production' Bangor, N Wales.
- KRIDIS, M.S., BARBER, G.D., OFFER, N.W., MURRAY, I., GIVENS, D.I. and DIXON, J. (1987). Prediction of organic matter digestibility of grass silages, recent developments. Proc. 8th Silage Conf., Hurley, pp 53-54.
- KRIDIS, M.S., GIVENS, D.I., OFFER, N.W. and BARBER, G.D. (1989). A comparison of NCD and other methods for the prediction of grass silage OMD. 50th Easter School of Agricultural Science Feedstuff Evaluation. University of Nottingham School of Agriculture. 10th-14th July 1989. In press.
- KRIDIS, M.S., OFFER, N.W. and BARBER, G.D. (1989). The effect of different washing procedures on the losses of OM and N from samples of hay incubated in nylon bags within the rumen of sheep. 2nd International Symposium on new techniques in Agriculture 'New Techniques in Cattle Production', Bangor, N Wales.
- MURRAY, I., KRIDIS, M.S., BARBER, G.D., GIVENS, D.I. and OFFER, N.W. (1987). Grass Silage 'In Vivo' Digestibility Prediction by NIR. CEC monograph, Brussels, 3-4 Dec. 1987.

APPENDICES

. 1

.

.

.

.

APPENDIX 1

Program Used to Calculate the Derivalised Log $\frac{1}{R}$ Segments (W values) for NIR Regression Equations

·f\$=" #8##.####":f1\$="####"

def fna(x)=sgn(x)#int(abs(x)#10000+.5)/10000 dim r(701),a1%(10,10),b1%(10,10),a2%(10,10),b2%(10,10),c2%(10,10),w15%(701) dim an1%(10,10), bn1%(10,10), an2%(10,10), bn2%(10,10), cn2%(10,10), w(10), dr%(10) print:print input "enter WAYELENGTH DATA source file name";file2\$ print:print input "enter RAW DATA source file name";file\$ print:print input "enter OUIPUT file name";file1\$ create file1\$ as 2 open file\$ as 1 OFEN FILE2\$ AS 3 READ #3; TERHX w1%=1100 for t%=1 to 700 wls%(t%)=w1% w1X=w1X+2 next t% for b%=1 to TERMX READ #3; DR%(b%) IF DRX(67,)<>1 THEN 1.1 FOR TX=1 TO 4:READ #3;A1X(67,TX):NEXT TX FOR 1X=1 10 4:READ #3;B1X(b%,1X):NEXT 1% goto 1.2 1.1 REM 6%>1 FOR TX=1 TO 4:READ #3;A2X(bX,TX):NEXT TX FOR 1%=1 TO 4:READ #3;B2%(b%,T%):NEXT T% FOR TX=1 TO 4:READ #3;c2%(b%,T%):NEXT TX 1.2 rem skip for t%=1 to 4 for tt%=1 to 700 if dr%(b%)<>1 then 1.3 rem first if a1%(b%,t%)=w1s%(tt%) then an1%(b%,t%)=tt% if biX(bX,tX)=wisX(ttX) then bniX(bX,tX)=ttX goto 1.4 rem second 1.3 if a2%(b%,t%)=w1s%(tt%) then an2%(b%,t%)=tt% if b2X(bX,tX)=wlsX(ttX) then bn2X(bX,tX)=ttXif c2%(b%,t%)=w1s%(tt%) then cn2%(b%,t%)=tt% 1.4 next tt% next t%

next b%

```
if end #1 then 2
for ttt%=1 to 200
read #1:n%
print "sample ";n%
for tt%=1 to 700
read #1;r(tt%)
next tt%
FOR bX=1 TO TERMX
sega1=0:sega2=0:segb1=0:segb2=0:seqc2=0
for tt%=1 to 4
IF DRX(bX)(>1 THEN 1.6
sega1=sega1+(r(an1%(b%,tt%)))
seqbi=segbi+(r(bni%(b%,tt%)))
goto 1.7
1.6 sega2=sega2+(r(an2X(bX,ttX)))
segb2=segb2+(r(bn2%(b%,tt%)))
segc2=segc2+(r(cn2%(b%,tt%)))
1.7 next tt%
IF DRX(bX)=1 then w(bX)=(100001(seqa1-seqb1))/4
IF DRX(bX)=2 then w(bX)=(10000#(sega2-2#seqb2+segc2))/4
next b%
on term% goto 11,12,13,14,15,16,17,18,19
ii print using fi$+f$;#2;n%,w(1):goto 20
12 print using f1$+f$+f$;#2;n%,w(1),w(2):goto 20
13 print using f1$+F$+f$+f$;#2;n%,w(1),w(2),w(3):goto 20
14 print using f1$+f$+F$+f$+f$;#2;n%,w(1),w(2),w(3),w(4):goto 20
15 print using f1$+f$+f$+f$+f$;$2;n%,w(1),w(2),w(3),w(4),w(5):goto 20
16 print using f1$+f$+f$+f$+f$+f$+f$;#2;nX,w(1),w(2),w(3),w(4),w(5),w(6):goto 20
17 print using f1$+f$+f$+f$+f$+f$+f$+f$;#2;n%,w(1),w(2),w(3),w(4),w(5),w(6),w(7):gata 20
18 print using f1$+f$+f$+f$+f$+f$+f$+f$+f$;#Z;n%,w(1),w(2),w(3),w(4),w(5),w(6),w(7),w(8):goto 20
19 print using f1$+f$+f$+f$+f$+f$+f$+f$+f$+f$+f$+f$;$2;n%,w(1),w(2),w(3),w(4),w(5),w(6),w(7),w(8),w(9):goto 20
20 next ttt%
2 close 1
close 2
close 3
stop
```

This program was written by N W Offer at the Department of Nutrition and Microbiology.

APPENDIX 2

Detailed Washing Procedures for the Polyester Bags Experiment

Treat Code	Description
А	After Incubation the bags were washed
	with cold water in a washing machine
	which consume approximately 107.5 l of
	water and run for 1^{l_2} hours.

NDI Before incubation, the bags were treated with neutral detergent solution as follows: 8 bags were put in previously boiled 2.5 1 neutral detergent solution (Van Soest and Wine, 1967), kept at 95°C and stirred continuously for 2 hours. Each gram of sample in the bag received 56.8 ml detergent. Then the bags were washed in the washing machine, as in treatment A, to remove excess detergent (we found it necessary to do this step before the bags went into the sheep, because it was difficult to remove excess detergent with hand washing only). The bags were then incubated in the rumen and after incubation they were washed in the washing machine as in A.

NDIND Before incubation the bags were treated with neutral detergent only, as in NDI, then washed in the washing machine, as in A, then incubated in the rumen, then treated again with neutral detergent, as in NDI, then washed in the washing
IND Bags were incubated, then treated with neutral detergent, as in NDI, then washed in the washing machine, as in A.

API Before incubation the bags were treated with acid pepsin solution as follows: 8 bags were put in 2.5 1 of acid pepsin solution (Tilley and Terry, 1963) in a water bath (40°C) for 96 hours with occasional stirring. Each gram of sample in the bags received 56.8 ml acid pepsin. Then the bags were washed in the washing machine, as in A, then incubated in the rumen. After incubation they were washed in the washing machine, as in A.

APIAP Before incubation the bags were treated with acid pepsin, exactly as in API, then washed in the washing machine, as in A, then incubated in the rumen, then treated again with acid pepsin, as in API, then washed in the washing machine, as in A.

IAP Bags were incubated in the rumen, then treated with acid pepsin, as in API, then washed in the washing machine, as in A. Before incubation the bags were treated with 60g of commercial washing powder (non-biological Persil brand) in the hot cycle (95°C, 3 hour run) of the washing machine. The bags were then incubated, and after incubation they were washed in the washing machine, as in A.

WPI

WPIWP Before incubation the bags were treated with washing powder, as in WPI, then incubated in the rumen, then treated again with washing powder, as in WPI, then washed in the washing machine, as in A.

IWP Bags were incubated in the rumen, then treated with washing powder, as in WPI, then washed in the washing machine, as in A.

ł

Analysis of Variance of % Organic Matter Disappearance of Hay from Polyester Bags Incubated in the Rumen of Sheep and Treated with Different Washing Techniques

		Variance H	Rati o (F))	
Source of Variation	D.F.	3 hrs	8 hrs	24 hrs	72 hrs
Sheep	3	2.685	0.104	7.422	10.260
Period	2	35.278	1.843	2.923	54.800
Sheep & Period	6	2.725 1.2		0.101	7.325
Treatments in Period 1	3	65.725	16.123	3.469	24.013
Treatments in Period 2	3	135.352	34.478	9.624	27.241
Treatments in Period 3	3	213.144	36.027	12.809	25.158
				· # - • · · · · · · · · · · · · · · · · · ·	
Residual MS	27	1.403	5.746	7,469	1,520

r

Analysis of Variance of Effective Degradability (%) of Hay Organic Matter Disappearance from Polyester Bags Incubated in the Rumen of Sheep and Treated with Different Washing Techniques at Selected Outflow Rate

	Va	ariance Rat	io (F)		
Source of Variation	D.F.	ED 2%	ED 4%	ED 8%	
Sheep	3	9.139	3.641	2.416	
Period	2	24.203	8.352	10.761	
Sheep & Period	6	1.255	0.556	0.409	
Treatments in Period 1	3	27.893	23.276	26.807	
Treatments in Period 2	3	60.040	50.314	61.300	
Treatments in Period 3	3	61.884	62.895	85.697	
Residual MS	27	1.400	2.187	2.377	

1

Organic Matter Disappearances (%) and Effective Degradabilities (ED) (%) for All Washing Treatments.

.

ಿಡ	0	ents Code ^b						29	%	85
Period	Sheep 1	Treatm	0 hrs	3 hrs	8 hrs	24 hrs	72 hrs	ED at 2	ED at 4	臣 at 8
11111111111111111111222222222222222222	1234123412341234123412341234123412341234	01 01 01 02 02 02 03 03 03 03 04 04 04 04 01 01 01 01 05 05 05 06 06 06 07 07 07 07 07 07 01 01 01 01 02 02 03 03 03 03 04 04 04 04 04 04 04 04 04 04	$\begin{array}{c} 11.68\\ 11.68\\ 11.68\\ 11.68\\ 11.68\\ 11.68\\ 11.68\\ 11.68\\ 11.97\\ 21.97\\ 21.97\\ 23.69\\ 23$	$\begin{array}{c} 18.78\\ 17.72\\ 18.76\\ 19.37\\ 29.80\\ 30.31\\ 28.13\\ 25.16\\ 27.88\\ 29.78\\ 29.78\\ 29.78\\ 29.78\\ 29.11\\ 29.01\\ 26.66\\ 29.78\\ 29.78\\ 16.26\\ 16.88\\ 15.73\\ 18.79\\ 16.26\\ 29.18\\ 30.62\\ 34.91\\ 27.75\\ 31.95\\ 30.61\\ 33.37\\ 33.16\\ 26.21\\ 27.75\\ 31.95\\ 30.61\\ 33.37\\ 33.16\\ 26.21\\ 27.75\\ 31.95\\ 30.61\\ 33.37\\ 33.16\\ 26.21\\ 27.75\\ 31.95\\ 30.61\\ 33.37\\ 33.16\\ 26.21\\ 27.75\\ 31.95\\ 30.61\\ 33.37\\ 33.16\\ 26.21\\ 27.75\\ 31.95\\ 30.61\\ 33.37\\ 33.41\\ 36.31\\ 36.31\\ 35.23\\ 32.34\\ 33.41\\ 36.31\\ 36.31\\ 37.50\\ 29.10\\ 29.53\\ 30.70\\ 30.79\\ 30.79\\ 30.79\\ 30.79\\ 30.70\\ 30.79\\ 30.70\\ 30.79\\ 30.70\\ 30.79\\ 30.70\\ 30$	30.97 29.61 30.03 37.46 39.96 40.12 41.84 40.08 41.99 42.19 44.95 43.62 38.60 39.07 41.09 32.02 30.49 43.73 46.15 42.42 44.09 47.90 47.65 46.91 46.91 46.58 41.27 40.60 47.65 46.91 40.42 35.66 28.18 23.48 41.65 28.48 30.44 31.50 42.32 39.004 41.50 35.66 28.18 23.48 30.44 31.65 42.32 39.004 41.65 42.32 39.004 41.65 42.32 39.004 42.34 42.85 39.36 48.04 40.76 38.72 48.04 40.76 38.72 40.76 39.36 42.85 39.36 48.04 40.76 38.72 40.76 40.76 39.36 40.76 39.36 40.76 39.36 40.76 39.36 40.76 39.36 40.76 39.36 40.76 39.36 40.76 39.36 40.76 39.36 40.76 39.36 40.76 39.36 40.76 39.36 40.76 39.36 40.76 39.36 40.76 39.36 40.76 39.36 40.76 39.72 40.76 39.36 40.76 39.72 40.76 39.72 40.76 39.72 40.76 39.72 40.76 39.72 40.76 30.72 40.76 30.72 40.76 30.72 30.72 40.76 30.72 30.72 40.76 30.72 30.72 40.76 40.76 30.72 40.76 40.76 30.72 40.76 40.76 30.72 40.76 40.76 30.72 40.76 30.72 40.76 40.76 30.72 40.76	59.27 55.66 61.33 63.83 65.97 59.90 61.31 62.09 64.78 63.39 64.13 59.64 63.39 64.13 59.64 63.86 62.92 49.58 61.16 67.00 64.22 67.58 67.17 68.65 58.58 67.17 68.65 57.98 64.22 67.58 57.98 62.92 67.58 57.98 62.92 67.58 57.98 62.92 57.98 62.31 62.72 57.98 62.31 62.72 57.16 62.31 62.57 58.58 57.98 62.31 62.72 57.16 62.52 65.73 62.99 53.90 49.77 56.89 57.98 62.31 62.52 65.74 62.85 65.74 65.97 6	67.96 71.53 72.14 72.39 71.79 79.46 75.69 81.08 76.19 80.11 69.63 74.20 76.99 74.85 73.34 73.99 81.64 79.87 73.34 77.26 80.01 82.05 79.02 82.15 81.54 77.21 77.21 77.21 77.37 75.43 77.16 73.58 74.70 74.67 75.43 77.21 77.21 77.21 77.21 77.22 81.54 77.22 81.54 77.21 77.21 77.21 77.21 77.21 77.22 81.54 77.22 81.54 77.21 77.21 77.21 77.21 77.22 81.58 74.70 74.67 78.48 81.52 81.52 81.52 81.29 82.11 79.29 82.211 79.29 82.211 79.29 82.20 77.62 77.62 77.62 77.62 77.62 77.62 77.64 77.62 77.62 77.64 77.62 77.62 77.64 77.62 77.62 77.64 77.62 77.62 77.64 77.62 77.64 77.62 77.64 77.62 77.64 77.62 77.64 77.62 77.64 77.62 77.64 77.62 77.64 77.62 77.64 77.62 77.64 77.62 77.64 77.62 77.64 77.62 77.64 77.62 77.64 77.62 77.64 77.62 77.64 77.62 77.62 77.64 77.62 77.64 77.62 77.64 77.62 77.64 77.62 77.64 77.62 77.64 77.62 77.64 77.62 77.64 77.62 77.64 77.64 77.62 77.64	556691402266458293985976566878894017775956668666666666666666666666666666666	444555555555555555444488777799889151165475552600084745 55555556666555555555555555555555555	$\begin{array}{c} 36\\ 34\\ 35\\ 39\\ 44\\ 43\\ 42\\ 46\\ 45\\ 1\\ 45\\ 40\\ 36\\ 36\\ 36\\ 78\\ 48\\ 50\\ 35\\ 46\\ 35\\ 52\\ 29\\ 44\\ 46\\ 45\\ 12\\ 29\\ 48\\ 46\\ 46\\ 47\\ 15\\ 22\\ 9\\ 44\\ 46\\ 46\\ 46\\ 46\\ 46\\ 46\\ 46\\ 46\\ 46$
	Perio Perio Perio	a 1 d 2 d 3	= Neu = Aci = Wasi	tral De d Pepsi hing Po	erergen In Wash owder	ing	rug			
Ъ	Treat	ment	s Code	= 01 0 05 4 09 V	Control API; 06 WPIWP;	; 02 N APIAP 10 IWP	DI; 03 ; 07 I <i>4</i>	NDI AP;	ND; 08 W	04 IND PI;

Analysis of Variance of % Nitrogen Disappearance of Hay from Polyester Bags Incubated in the Rumen of Sheep and Treated with Different Washing Techniques

			Variance 1	Ratio (F)
Source of Variation	D.F.	3 hrs	8 hrs	24 hrs	72 hrs
	ſ	0 700	1 204	2, 190	r 200
Sheep	3	0.702	1.294	3.180	1.300
Period	2	243.144 44.86		33.531	155.584
Sheep & Period	6	0.883 0.640		2.538	10.694
Treatments in Period 1	3	84.337 53.788		47.544	161.775
Treatments in Period 2	3	417.427	143.276	96.701	119.890
Treatments in Period 3	3	214.922	105.829	113.594 138.332	
Residual MS	27	7.343	13.110	5.306	0.773

•

Analysis of Variance of % Nitrogen Concentration (g Kg⁻¹ DM) in Bag Residues

		,	Variance R	atio (F))
Source of Variation	D.F.	3 hrs	8 hrs	24 hrs	72 hrs
Sheep	3	1.083	1.257	0.925	2.772
Sheep & Period	6	0.813	40.554 0.666	49.751 2.756	6.161
Treatments in Period 1 Treatments in Period 2	3 3	30.352 243.153	29.496 96.616	42.485 97.889	94.736 79.960
Treatments in Period 3	3	86.998	53.034	73.076	129.083
Residual MS	27	0.170	0.3826	0.3331	0.1266

•

	Niti	rogen	Di	sapp	earances	(%)	an	d N	itro	gen	Conce	ntration	
(g	kg 1	DM)	in	Bag	Residues	(NF	()	for	A11	Was	shing	Treatments	•

Periods ^a	Sheep No	freatments Code ^b) hrs	hrs	3 hrs	14 hrs	2 hrs·	R 0 hrs	R 3 hrs	R 8 hrs	R 24 hrs	R 72 hrs
1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 2 2 2 2	1234123412341234123412341234123412341234	$\begin{array}{c} 01\\ 01\\ 01\\ 02\\ 02\\ 02\\ 02\\ 03\\ 03\\ 03\\ 04\\ 04\\ 04\\ 04\\ 01\\ 01\\ 01\\ 05\\ 05\\ 05\\ 06\\ 06\\ 06\\ 06\\ 06\\ 06\\ 06\\ 07\\ 07\\ 01\\ 01\\ 01\\ 01\\ 01\\ 01\\ 01\\ 01\\ 01\\ 01$	16.35 16.35 16.35 37.12 3	32.47 24.19 24.22 46.24 48.67 49.61 42.08 57.39 57.38 54.56 57.39 57.38 42.03 57.39 57.38 42.03 20.32 24.62 24.62 24.62 24.62 24.62 24.62 24.62 27.57 81.36 77.57 81.36 80.515 79.02 77.74 21.08 23.307 26.98 53.357 79.02 77.74 21.08 23.307 25.53 56.472 67.50 57.360 53.357 53.357 53.357 53.357 24.62 24.62 25.533 74.366 80.515 57.57 81.399 84.366 80.555 79.02 77.74 21.08 23.307 25.53 56.472 57.50 57.60 73.60 52.88 56.01 55.02 58.96 57.60 52.88 56.01 55.02 56.01 55.02 56.01 55.02 56.01 55.02 56.01 55.02 56.01 55.02 56.01 57.50 57.60 57.60 57.60 57.60 57.60 57.80 57.60 57.80 5	$\begin{array}{c} 45.26\\ 40.43\\ 41.52\\ 40.25\\ 58.83\\ 55.64\\ 58.28\\ 60.25\\ 79.41\\ 65.05\\ 73.59\\ 73.79\\ 64.24\\ 70.55\\ 63.83\\ 39.69\\ 38.54\\ 39.42\\ 41.89\\ 76.38\\ 39.42\\ 41.89\\ 76.38\\ 39.42\\ 41.89\\ 76.38\\ 39.42\\ 41.89\\ 76.38\\ 39.42\\ 41.89\\ 76.38\\ 39.42\\ 41.89\\ 76.38\\ 39.42\\ 41.89\\ 76.38\\ 39.42\\ 41.89\\ 76.38\\ 39.42\\ 41.89\\ 39.42\\ 41.89\\ 39.42\\ 41.89\\ 39.42\\ 41.89\\ 39.42\\ 41.89\\ 80.73\\ 74.71\\ 75.54\\ 80.73\\ 79.13\\ 41.80\\ 33.03\\ 40.26\\ 80.73\\ 79.13\\ 41.80\\ 33.03\\ 40.26\\ 80.73\\ 79.13\\ 41.80\\ 33.03\\ 40.26\\ 80.73\\ 79.13\\ 41.80\\ 30.36\\ 80.73\\ 79.13\\ 41.80\\ 30.36\\ 80.73\\ 79.13\\ 41.80\\ 30.26\\ 80.73\\ 79.13\\ 41.80\\ 30.26\\ 80.73\\ 79.13\\ 41.80\\ 30.26\\ 80.73\\ 79.13\\ 41.80\\ 30.26\\ 80.73\\ 79.13\\ 41.80\\ 30.6\\ 80.73\\ 79.13\\ 41.80\\ 30.6\\ 80.73\\ 79.13\\ 41.80\\ 30.6\\ 80.73\\ 79.13\\ 41.80\\ 30.6\\ 80.73\\ 79.13\\ 41.80\\ 30.6\\ 80.73\\ 79.13\\ 41.80\\ 30.6\\ 80.73\\ 79.13\\ 41.80\\ 30.6\\ 80.73\\ 79.13\\ 41.80\\ 30.6\\ 80.73\\ 79.13\\ 41.80\\ 30.6\\ 80.73\\ 79.13\\ 41.80\\ 30.6\\ 80.73\\ 75.41\\ 81.01\\ 75.78\\ 69.47\\ 75.41\\ 81.01\\ 75.78\\ 69.47\\ 75.41\\ 81.01\\ 75.41\\ 81.01\\ 75.78\\ 69.47\\ 75.41\\ 81.01\\ 75.64\\ 18.01\\ 75.64\\ 81.01\\ 75.64\\ 18.01\\ 75.64\\ 18.01\\ 75.64\\ 81.01\\ 75.64\\ 18.01\\ 75.64\\ 81.01\\ 75.64\\ 81.01\\ 75.64\\ 18.01\\ 75.64\\ 18.01\\ 75.64\\ 18.01\\ 75.64\\ 18.01\\ 75.64\\ 18.01\\ 75.64\\ 18.01\\ 75.64\\ 18.01\\ 75.64\\ 18.01\\ 75.64\\ 18.01\\ 75.64\\ 18.01\\ 75.64\\ 18.01\\ 75.64\\ 18.01\\ 75.64\\ 18.01\\ 75.64\\ 18.01\\ 75.64\\ 18.01\\ 75.64\\ 18.01\\ 75.64\\ 18.01\\ 75.66\\ 18.01\\ 18.01\\ 75.78\\ 18.01\\ 75.66\\ 18.01$	$\begin{array}{c} 63.43\\ 66.41\\ 66.40\\ 62.92\\ 70.19\\ 65.57\\ 71.07\\ 67.74\\ 82.44\\ 60.64\\ 78.31\\ 79.64\\ 78.31\\ 79.64\\ 78.31\\ 79.64\\ 78.31\\ 79.64\\ 78.33\\ 79.64\\ 78.33\\ 79.64\\ 78.33\\ 96.42\\ 79.64\\ 78.33\\ 96.42\\ 83.40\\ 89.26\\ 83.40\\ 89.26\\ 85.92\\ 87.51\\ 85.02\\ 86.93\\ 63.63\\ 27.30\\ 89.36\\ 85.02\\ 87.51\\ 85.02\\ 85.92\\ 76.95\\ 85.02\\ 87.51\\ 85.92\\ 76.95\\ 85.02\\ 87.51\\ 85.92\\ 76.95\\ 87.52\\ 85.92\\ 76.95\\ 85.91\\ 90.36\\ 90.56\\ 90$	78.08 78.83 79.91 80.34 81.46 86.20 82.66 89.79 93.34 91.24 93.27 85.32 91.17 89.48 91.23 79.45 80.01 78.39 78.27 83.54 85.72 82.22 82.01 90.08 89.94 89.44 89.40 89.44 89.20 89.94 89.44 89.22 82.73 85.52 82.73 81.94 85.52 82.73 81.94 85.52 82.73 81.94 85.52 82.73 81.94 85.52 82.73 81.94 85.52 82.73 81.94 85.52 87.20 89.15 87.20 89.43 87.20 89.43 87.20 89.43 87.20 89.43 87.20 89.43 87.20 89.43 87.20 89.43 87.20 89.43 87.20 89.44 89.45 87.20 89.44 89.45 87.20 89.45 87.20 89.45 87.20 89.45 87.20 89.45 87.20 89.43 87.20 89.43 87.20 89.43 87.20 89.43 87.20 87.95 87.20 89.43 87.20 89.43 87.20 89.43 87.20 89.43 87.20 89.43 87.20 87.20 87.40 87.20 87.40 87.20 87.40 87.20 87.40	9.8 99.8 99.8 8.4 4.4 8.4 4.4 8 8.4 4.4 8 8.4 4.4 8 8 9.9 9.9 9.3 3.1 1.1 1.3 3.3 3.3 1.1 1.1 8 8 8 8 9.9 9.8 8 9.0 8 8 8 9.0 8 8 9.0 8 8 9.0 8 8 9.0 8 9.0 8 9.0 8 9.0 8 9.0 8 9.0 8 9.0 8 9.0 9.0 8 9.0 9.0 8 9.0 8 9.0 9.0 9.0 9.0 9.0 9.0 9.0 9.0 9.0 9.0	$\begin{array}{c} 8.7\\ 9.61\\ 9.8\\ 2.8\\ 5.21\\ 4.4\\ 1.22\\ 4\\ 0.3\\ 4.63\\ 1.29\\ 4.63\\ 1.29\\ 2.3\\ 1.89\\ 1.29\\ 7.5\\ 4.9\\ 5.4\\ 9.5\\ 4.5\\ 6.7\\ 6.\\ 7\\ 7\\ 7\\ 7\\ 7\\ 8.\\ 7\\ 7\\ 7\\ 8.\\ 7\\ 7\\ 7\\ 8.\\ 7\\ 7\\ 7\\ 8.\\ 7\\ 7\\ 7\\ 7\\ 7\\ 8.\\ 7\\ 7\\ 7\\ 7\\ 7\\ 7\\ 7\\ 7\\ 7\\ 7\\ 7\\ 7\\ 7\\$	88897777736546566989844541222223338898115565444455555 888977777365465698984454122222333889811977567054 197755670545555 197755670545555 197755670545555 197755670545555 1977556705455555 1977556705455555 1977556705455555 1977556705455555 1977556705455555 1977556705455555 1977556705455555 1977556705455555 1977556705455555 1977556705455555 1977556705455555 1977556705455555 1977556705455555 19775567054555555 1977556705455555 19775567054555555 19775567054555555 19775567054555555 19775567054555555 197755670545555555 197755670545555555 19775567054555555 19775567054555555 19775567054555555 197755670545555555 19775567054555555 19775567054555555 197755670545555555 19775567054555555 19775567054555555 19775567054555555 19775567055455555555 197755670554555555555555555555555555555555	9779881785655588558899887893353333788886655533323333 	77776776433353534487889789546655555777756666223333333
Pe	eriod	2 = Ac	id Pepsi	n Washin	g	03	NDIND; C	4 IND:	: 05 AI	PI; 06	APIAP	, ,

Period 3 = Washing Powder

.

.

.

07 IAP; 08 WPI; 09 WPIWP; 10 IWP.

.

Analysis of Variance of the Regression Equations for the 3 Populations Used in the Calibration Set for the Conventional Predictors of in vivo OMD.

Models	MADF	LIGA	NCOMD	PCOMD	NB48 OMD	IVOMD
Model DF	120	120	117	120	120	120
1 MS	26.19	19.07	18.39	17.82	12.73	10.44
Model DF	118	118	115	118	118	118
2 MS	20.27	10.91	17.12	10.21	12.93	10.57
Model DF	116	116	113	116	116	116
3 MS	19.81	10.63	16.95	8.911	12.27	9.942

Analysis of Variance of the Regression Equations for the 3 Populations Used in the Calibration Set for the NIR Regression Equation**s**

dels 1-term 2-term 3-term 4-term 5-term del DF 120 119 118 117 116 1 MS 18.97 14.13 11.38 9.496 8.387 del DF 18.97 14.13 11.38 9.496 8.387 del DF 118 117 116 115 114 del DF 16.39 9.097 9.051 7.563 7.293 del DF 116 113 110 7.563 7.293 del DF 16.42 8.903 8.529 7.434 7.011										
al DF 120 119 118 117 116 MS 18.97 14.13 11.38 9.496 8.387 al DF 118 117 116 115 114 MS 16.39 9.097 9.051 7.563 7.293 al DF 116 113 110 107 104 MS 16.42 8.903 8.529 7.434 7.011	sls	1-term	2-term	3-term	4-term	5-term	6-term	7-term	8-term	9-term
el DF 118 117 116 115 114 NS 16.39 9.097 9.051 7.563 7.293 el DF 116 113 110 107 104 16.42 8.903 8.529 7.434 7.011	lel DF MS	120 18.97	119 14.13	118 11.38	117 9.496	116 8.387	115 7.303	114 6.769	113 6.13	112 5.88
iel DF 116 113 110 107 10 ⁴ 3 NS 16.42 8.903 8.529 7.434 7.011	lei DF 2 MS	118 16.39	117 9.097	116 9.051	115 7.563	114 7.293	113 6.982	112 6.331	111 5.833	110 5.632
	jel DF 3 MS	116 16.42	113 8.903	110 8.529	107 7.434	104 7.011	101 6.747	98 6.44	95 5.574	92 5.489

.

•

Analysis of Variance of the Individual Regression Equations for Each Population Used in the Calibration Set

			DUNN			T TCA			THUR I				
Models		Pop 2 & 4	Pop 2 & 3	Pop 4 & 3	Pop 2 & 4	Pop 2 & 3	Pop 4 & 3	Pop 2 & 4	Pop 2 & 3	Pop 4 & 3	Pop 2 & 4	Pop 2 & 3	Pop 4 & 3
Model	DF	97	93	48	97	93	48	94	92	46	97	93	48
1	MS	23.78	25.21	23.49	1 0. 32	21.59	19.87	20.02	15.59	18.96	18.91	12.92	13.10
Model	NG DF	96	92	47	96	92	47	93	91	45	96	92	47
2		21.54	18.58	21.11	10.16	12.27	9.793	18.21	15.46	18.33	10.01	9.343	11.91
Model	fa A	95	91	46	95	91	46	92	90	44	95	91	46
3		21.42	18.28	19.53	10.09	12.08	8.878	18.11	15.43	17.64	9.193	8.838	8.476

The Individual Values for <u>in vivo</u> OMD, <u>in vivo</u> DOMD, Chemical and Biological Parameters for the 122 Calibration Silages Used in this Study

,

.

.

	0	Ð	MO							
*0	e N	8	8					Ð		
a a	lag	F	5	皆	GA	Ð	Ð	6 8 1 8	Ð	۲ ۲
ፈ	Si	띱	티	MA	ГI	Ř	R	é	1 M	Asl
2	79047	68.50	60.3485	330	2.745	71.7174	64.68	88.27	66.31	11.90
2	B0022 B0031	/1.73 25 EA	63.4363 F0.103/	362	2.790	68.7019	63.89	81.40	70.80	9.00
2	80170	0J.J4 68 56	37,1020 41 6227	410	2.803	54.8235	51,38	82.57	62.78	9.70
2	80179	48.69	62,1645	282	2.195	01.2220 76 0360	37.72 70.00	82.10 03.07	64.40 /0 /0	9.70
2	B1003	52.80	47.4672	424	3.829	43.9377	76.03	61 47	67.00	7.30
2	81005	72.00	63.8640	326	2.603	72.3843	69.10	91.00	70.01	11.30
2	81006	75.93	70.6149	283	2.511	77.7419	67.71	87.63	69.27	7.00
2	81007	58.56	51.7085	353	3.105	61.1657	58.31	65.03	63.45	11.70
2	81012	82.79	73.6831	261	1,817	84,7389	81.23	96.83	75.06	11.00
2	81014	62.70	58.2483	385	3.081	56.3630	55.34	80.30	60.61	7.10
2	81020	/2.31	66.3806	218	2.761	63.0074	62.01	87.30	64.77	8.20
2	82004 82005	63.89	3/,11//	279	2,977	55.8652	54.38	76.77	60.72	10.60
2	82008	63.73	37.7438 KÚ 1435	383 710	3.1/5	4 50 4140	49.35	11.35	56.71	8,80
2	82013	65.92	60.4486	348	2.913	57,0340 85,1036	54 25	70.7V D1 50	60.43	7.10
2	82014	66.57	61,1778	392	3.013	62.8944	54.66	72.17	61.00 Aŭ 44	8.30 8.10
2	82015	72.85	66.6578	353	2.647	75.1913	66.47	88.07	68.83	8.50
2	82016	74.40	68.3736	342	2.544	68.5704	67.73	90.43	69.21	8.10
2	82022	66.58	60.2549	351	2.644	60.1360	61.57	81.50	66.21	9.50
2	82023	67.41	61.7476	373	2.658	64.8920	57.08	79.50	69.27	8.40
2	82025	12.30	65./662	335	2.493	72,2772	67.23	89.27	68.39	9.10
2	82031	71.86	61.0237	334 110	21927	00./995 70 4015	63.42	65.20	67.51	10.70
2	82033	67.10	61,7991	374	2.825	70,400J 62,1064	57.33	87.37 82.57	68.48	7.30
2	82057	61.00	57.7060	409	3.212	52.7029	50.53	73.80	56.04	5.40
2	82065	67.46	62.7378	40B	2.612	44.7819	58.35	76.43	63.52	7.00
2	82074	66.87	60.2499	393	2.8:0	62.5567	59.95	83,70	65.91	9.90
2	82079	77.97	70.5B10	257	2.151	77.1125	76.18	93.47	12.97	9.50
2	87008	79.18	65.7234 65.450	3(13	2.448	67.6808	67.36	82.80	69.38	9.20
2	83013	79.59	72.1881	300 279	2,734	32,6040 71,9207	33.78 77 02	/0.6/	59.39	10.00
2	83014	78.04	71.9527	304	1.981	68,0589	74.16	91.30	74.40	7.00
2	83021	72.63	65.1491	334	2.363	65.7921	67.68	90.80	69.81	10.30
2	83026	79.51	74.1033	301	2.229	68.2661	68.94	68.47	70.64	6.80
2	B3027	69.53	62.9246	407	2.385	53.5362	61.20	85.73	60.08	9.50
2	B3031 B3034	52.91	48.4327	397	3.430	50.6896	44.33	67.23	50.70	8.50
2	83049	73.31	0/.101/ 71 1055	313	2.312	71.0478	/1.28	90.90	/1.58	8.50
2	83065	67.67	60.3170	343	2.835	63 0456	67.40	87.43 80.10	63.49	1.80
2	83110	78.49	10.2485	298	2.222	75.8631	25.04	92.80	0).JO 71 RT	10.00
2	84018	66.30	61.9242	377	3.045	54.3314	56.51	77.00	52.55	6.60
2	84v74	68.79	61.3607	333	2.654	69.9694	65.4B	82.30	66.10	10.80
2	84087	61.61	57.0509	377	2.999	57.3031	56.96	76.70	62.92	7.40
2	85016	66.81	60.8639	306	2,418	68.5526	67.02	81.20	66.76	8.90
2	85018	61.07	14.1019 55 0170	230	1.774	89.5758	BU. /9	95.97	77.87	9.00
2	85022	64.72	57.9244	348	2.330	33.0231 57 7654	27.81 51 17	75 31	53.62	8,90
2	B5024	79.72	73.8207	312	2.364	69.7674	64.64	87.67	07.0Z 65.94	10.50
2	85026	80.92	74.5273	285	1.897	74.5797	76.66	94.27	75.38	7.90
2	B5030	66.51	59.9255	367	2.852	58.8235	55,41	80.63	63.82	9.90
2	85033	61.17	55.6035	375	3.148	57.5446	55,83	76.67	60.43	9.10
2	85036	71.15	65.3157	381	2,551	52.5599	61.44	83.03	63.05	8.20
2	83037 85039	60.VI 70.D5	60.9794 71.1740	559	2.582	52.0548	56.40	17.27	57,34	6.20
2	85041	78.56	77,1181	021 287	2,1/1	00.2018 00.2018	07.00 71 5.1	41.IV 52 07	11,14	4.40
2	85043	67.77	60.8575	317	2,283	65,7016	63.54	72.00	57 30	0.20 10.20
2	85048	71.90	63.9191	349	2.465	67,6041	65.24	81.00	65.12	11.10
2	85057	80.08	73.0330	291	1.991	75.8744	73.33	87.70	15.60	8.80
2	86005	73.16	66.5024	332 ·	2.435	73.6073	60.94	85.03	69.92	9.10

.

APPENDIX 12 (cont)

		Ą	Ð							
	No	đ	X					A		
Ŋ	ge	۲Ļ	1 N			Δ	Α	20	A	22
g.	[]a			Ę,	EGA	MQ	ğ	348	Ю	ų
ц Ц	3		비	Ð.	E	M	д	Ē	Ц	As
2	86015 86015	70.35	62.8929 70.3923	360 335	2,710	73.4141	55.40 73 94	81.07 93.73	60.24 20.92	10.60
2	86018	74.99	69.2908	374	2.458	71.1850	64.29	B4-50	67.26	7.60
2	86020	75.92	69.7705	319	2.330	73,9977	69.19	88,70	72.24	8.10
2	86021 86026	75.84 76.12	70.2278	314 338	2.472	73.9618	67.40 71 DD	87.63	69.13	7.40
2	86029	70.92	63.1897	333	2.379	64.1975	60.86	85.77	62.67	10.90
2	86033	68.14	63.5065	383	2.640	64.0412	59.75	82.07	63.56	6.80
2	86036	75.61	66.9149	332 754	2,065	78.6859	71.66	89.50	71.02	11.50
2	86042	78.42	69.4801	334 298	1.935	88.7220 B3.5214	65,18 73,19	92.10	67.96 72.56	10.89
2	86067	69.20	60.2040	32B	2.165	79.0805	71.40	88.77	67.81	13.00
3	. 2	68.70	60.6552	423	3,140	66.8303	54.70	85.50	65.10	11.71
3	4	/5.10 74 90	68.8442	373	2.850	66.7322	58.83	84.00 Do Do	64.10	8.33
3	7	78,70	73.0729	366	2.580	75.6981	64.77	90.80	71.30	10,65
3	8	74.50	65.075B	338	2.570	77.4925	66.82	88.60	70.50	12.65
3	9	68.10	62.8767	424	3.420	53,9727	45.56	76.10	55.00	7.67
3	10	68.70 72.60	61.0056 64 8318	364 405	3.010	69.7237 45.1508	60.02 53.51	83,60 87,30	67.30	11.20
3	12	70,10	64,5972	386	2,850	66.4041	57.70	87.40	69.00	7.85
3	13	63.00	56.3787	410	3,160	62.2761	50.56	81.30	60.50	10.51
3	15	67.70	61.4902	428	3,170	62.3755	48.12	81.00	59.00	9,44
3	10	72.90	64,4363 66,2228	400 390	3.150	/0.0039 A7 A350	57.30	87.30 87.50	67.20 20 20	11.61
3	28	74.20	67.4478	309	2.530	79.7580	71.12	95.10	77.40	8.20 9.10
2	37	76.70	69.4902	382	2.860	69.4088	65.78	92.70	72.60	9.40
3 7	38	77.80	70.4090	358	3.185	71.9930	67.96	82.40	74.50	9.50
3	40	76.10	67.8440	335 414	3,000	68.6999 68.1020	6/.14 50 25	90.70 84.00	13.60	8.70
3	42	72.20	65.4B54	344	2.875	71.9492	67.20	85,10	67.30	9.30
3	44	74.50	66.5285	351	2.765	68.5165	64.17	91.30	68.40	10.70
3 र	45 44	68.60	62.5632	382	3,185	62.6661	52.30	68.50 60 Ju	67.70	8.BV
3	40	77.40	68.6538	370	2.715	67.6315	58.57	90.30 91.30	70,70	6.70 11 30
4	33	78.90	71.0889	331	2.160	69,5894	62.68	93.23	76.20	9.90
4	34	74.10	66.4677	380	2.300	67.3356	63.41	90,53	72.90	10,30
4	35 34	76.80	71.3472	351	2.370	68.7836	61.35	89.47	73.90	7.10
4	37	62.80	58,8436	396	2.840	52.7215	44.01	77.95	57.50	8.00
4	38	62.00	56.7920	384	3,250	53.2751	46.79	76.23	57.50	8.40
4	39	63.50	57.9120	372	3,140	52,5219	47.07	79.80	58.60	8.80
4 4	40	65.10	58.3675 61 6818	574 481	3.240	53.9459 AB AGIL	46.74	80.10 20.40	58.20 59.00	7.50
4	42	72.70	67.2475	357	2.630	56.8649	54.56	65.53	68.30	7.50
4	43	77.60	72.1680	319	2.350	69.4624	59.57	83.57	70,90	7.00
4	44	77.20	71.4100	325	2.310	69.2973	55.15	90.37	68.00	7.50
4 4	40 46	/5.00 70 50	67.2750 63.4500	365 364	2,410	53./34/ 54 0000	58.77	87.77	71.90	10.30
4	47	86.70	81,1512	357	1.800	59.6581	64.32	\$5.37	74.40	6.40
4	48	73.20	67.6368	366	2.610	61.3636	52.53	84.63	67.60	7.60
4	49	74.80	68.9656	345	2.390	66.2670	58.56	90.93	72.60	7,80
4	51	72.20 75.30	63.1966 69.4766	339 317	2.310	65.3878	57.32 57.09	86.97 85 97	70.80 70.50	9.70
4	52	74.30	68.9504	33B	2.610	69.7198	56.02	87.53	69,10	7.20
4	53	72.90	66.7764	362	2.270	71.8341	55.42	90.87	67.80	8.40
4	54	62.40	57.2208	399	3,000	58.6696	44.65	79.27	63.90	8.30
4	55 56	62.60 64.20	06.7106 59.1924	426 380	3.390 2.940	58.1678 61.6052	45.61 47 50	74.80 Rù 17	60,60 60 40	9.40 7 00
4	57	74.00	69.4120	342	2.330	0110VJZ #	57.39	89,70	70.10	6.20
4	58	72.50	66.7000	373	2.490	*	53,83	83.80	68.20	8.00
4	59	78.00	71.5260	353	2.090	73.1734	57.97	20.67	71.20	8.30

.

* Population Numbers = See Table 2.1

.

Analysis of Variance to Test Whether the Bivariate Relationship has a Significant Improvement over the Univariate Relationship for in vivo Predictors Studied in this Work.

							The second se								
	IVOND & NB48 OND	IVOND & FCCND	S CHONI	IVOMD 6. MADF	novd & liga	NB48 CMD & PCCPD	NE48 CMD & NCCMD	NB48 (MD) & MADF	NB48 CPD & LIGA	roced a	PCCMD & MADF	FCCMD & LIGA	NCOYD & MADF	NCOYD & LIGA	LIGA & MADF
Res. DF for Univariate	120	120	117	120	120	120	117	120	120	117	120	120	117	117	120
Res. DF for Blvariate	611	611	116	611	119	611	911	119	119	116	119	119	116	911	119
Res. SS for Univariate	1252.4	1252.4	1252.4	1252.4	1252.4	1527	1527	1527	1527	2138.4	2138.4	2138.4	2151.7	2151.7	2288.9
Res. SS for Bivariate	1141.4	1191.8	1180.3	1223.4	1105.4	1370.4	1376.4	1471.4	1344.4	1841.4	2137.8	1857.9	2092.1	1630.9	2267.6
DF changefrom Univertate		Ч	1	I		1	1	1	1	ч	г	Ч	г	1	1
SS change from Univertate	111	60.6	1.2	29	147	156.6	150.6	55.3	182.6	297	0.6	280.5	59.6	520.8	21.3
SS change MS = DF change	111	60.6	72.1	29	147	156.6	150.6	55.3	182.6	297	0.6	280.5	59.6	520.8	21.3
Res. MS for Bivariate	9,592	10.01	10.17	10.28	9.29	11.52	11.87	12.37	11.30	15.87	17.96	15.61	18.04	14.06	19.06
F =Res. NS for Bivariate	11.57	6.05	7.09	2.82	15.82	13.59	12.69	4.47	16.16	18.71	0.03	17.97	3.30	37.04	1.12
															Ī

700 Log $\frac{1}{R}$ Values (Data Points) taken at 2 nm Intervals Between 1100 and 2500 nm for One Silage Sample Only.

0.1664419025,0.1661406755,0.1659163088,0.1656093001,0.1653319299,0.1650668830 0.1647853851,0.1644063890,0.1640610397,0.1639302820,0.1635922641,0.1633238047 0.1630838513,0.1630213559,0.1629626900,0.1630817503,0.1629874259,0.1629415154 0.1630947292,0.1633338332,0.1635076106,0.1636233032,0.1638325006,0.1640163511 0.1642820388,0.1645977795,0.1650149077,0.1654456854,0.1659781039,0.1663010418 0.1667060703,0.1671374142,0.1675156802,0.1676680480,0.1681915969,0.1684450209 0.1687301099,0.1691008359,0.1694639325,0.1697784662,0.1695914617,0.1700887382 0.1702582985,0.1703150123,0.1703680605,0.1704328656,0.1702793539,0.1701966077 0.1701103449,0.1700074822,0.1698285788,0.1696144789,0.1695683300,0.1694488674 0.1692214906,0.1691991091,0.1683632433,0.1679545790,0.1673880816,0.1668623239 0.1663043499,0.1657318175,0.1651650369,0.1646261513,0.1640713811,0.1635703743 0.1631001830,0.1626051962,0.1623662412,0.1621832401,0.1613762677,0.1610110849 0.1605979949,0.1602031738,0.1578479715,0.1595860571,0.1593152434,0.1590566486 0.1587398052,0.1585166305,0.1583041996,0.1579140276,0.1576710472,0.1577087194 0.1574213058,0.1572260112,0.1570316851,0.1567559540,0.1565779299,0.1563070267 0.1561042368,0.1558322906,0.1555506885,0.1552874744,0.1550554633,0.1548705399 0.1546329111,0.1546607685,0.1542711758,0.1542637646,0.1541135758,0.1541147530 0.1540041417,0.1539461464,0.1540947079,0.1540127248,0.1541282684,0.1542810649 0.1543414444,0.1545412540,0.1548781395,0.1550387826,0.1554214507,0.1558655947 0.1563330591,0.1568033546,0.1574293143,0.1579619944,0.1586765200,0.1594473869 0.1604032815,0.1612188220,0.1621005068,0.1631219834,0.1641786844,0.1659660634 0.1660216835,0.1671226472,0.1680904180,0.1690280885,0.1698619276,0.1707547903 0.1717507518,0.1724800915,0.1730375986,0.1739452779,0.1749657217,0.1757613420 0.1766073257,0.1776651889,0.1786521971,0.1800245643,0.1816809326,0.1836288869 0.1855346200,0.1077747476,0.1901235418,0.1525974190,0.1952980757,0.1582081980 0.2011877596,0.2044590712,0.2078690082,0.2113066316,0.2148710936,0.2184497118 0.2219866216,0.2256174833,0.2268036048,0.2319382131,0.2350013107,0.2382195741 0.2412954271,0.2441992760,0.2469890565,0.2494630218,0.2518257499,0.2538680732 0.2559843063,0.2578205466,0.2593263388,0.2605755329,0.2618782520,0.2631074190 0.2842062306,0.2652140856,0.2661742270,0.2668962777,0.2676506937,0.2683100700 0.2689867616,0.2695714235,0.2701587379,0.2706004381,0.2710570395,0.2714005411 0.2715177834,0.2716079056,0.2717253566,0.2718856633,0.2717204392,0.2716158330 0.2714684308,0.2712344229,0.2709567845,0.2706719935,0.2703635395,0.2699163951 0.2694804668,0.2688513100,0.2682336667,0.2675308287,0.2665483952,0.2659344971 0.2652006447,0.2643675804,0.2637930810,0.2631162405,0.2624122500,0.2617871165 0.2612323761,0.2606284320,0.2600920796,0.2596427798,0.2592709064,0.2589902282 0.2587527633,0.2584812045,0.2582466602,0.2580207586,0.2578342557,0.2577073574 0.2575010359,0.2574255764,0.2572166520,0.2567917440,0.2568757620,0.2567091584 0.2566114068,0.2565608025,0.2565595508,0.2566058636,0.2564971745,0.2565182447 0.2565052807,0.2563743293,0.2564210594,0.2563309073,0.2561783195,0.2560196519 0.2559364736,0.2556432784,0.2553631761,0.2549746037,0.2546684444,0.2540944815 0.2535780966,0.2529406548,0.2523193061,0.2515436577,0.2509938478,0.2503938377 0.2498090863,0.2492330223,0.2485003339,0.2485268776,0.2474023998,0.2466193140 0.2460521013,0.2455114126,0.2448726296,0.2444280086,0.2439302653,0.2434994131 0.2430228293,0.2425473151,0.2420822660,0.2417301834,0.2414832860,0.2410230637 0.2406838983,0.2404534072,0.2402153313,0.2400933057,0.2401308417,0.2401229739 0.2403321266,0.2406950742,0.2411558479,0.2415936589,0.2421769768,0.2429674268 0.2438716292,0.2449292392,0.2460945994,0.2475041747,0.2489637434,0.2498399317 0.2510986924,0.2523252070,0.2536080778,0.2540433542,0.2561614811,0.2574777007 0.2585076968,0.2594169974,0.2602001101,0.2608932555,0.2615994811,0.2622935176 0,2631227970,0.2638834417,0.2646833658,0.2655826807,0.2665478567,0.2674907148 0.2682771385,0.2691568136,0.2700760067,0.2709968090,0.2716031969,0.2722621560 0.2726818919,0.2728420198,0.2727482021,0.2724635601,0.2719557285,0.2714143097 0.2708286345,0.2703500986,0.2699590125,0.2697702646,0.2696323097,0.2696395516 0.2697705626,0.2698569298,0.2700255513,0.2702209353,0.2705434859,0.2703915834 0.2701598108,0.2700112164,0.2697320282,0.2692240775,0.2688496113,0.2662757080 0.2677834994,0.2673152983,0.2668777492,0.2664595842,0.2663003504,0.2657527924 0.2653859654.0.2649788857.0.2645753920.0.2642513216.0.2639431357.0.2636194527 0.2632835269,0.2630139397,0.2620115098,0.2624457777,0.2622210681,0.2619299293 0,2615377307,0.2613935769,0.2612212100,0.2611186802,0.2608555192,0.2605517805 0.2601664662.0.2597897649.0.2594654560.0.2590718567.0.2586893141.0.2582978606 0.2576950192,0.2573883832,0.2570638455,0.2566946978,0.2562040687,0.2559783161 0.2557437122,0.2555425465,0.2555145323,0.2555740178,0.2557483017,0.2560456395 0.2566098869,0.2572382390,0.2579777837,0.2591010630,0.2604070008,0.2619613409 0.2638612390,0.2661888897,0.2689698637,0.2722411156,0.2759392858,0.2802454829 0.2851497319,0.2907494903,0.2969333231,0.3036383092,0.3110366464,0.3189793527 0.3275556862,0.3364481032,0.3458366394,0.3554689884,0.3650854826,0.3744184971 0.3832658529,0.3915078640,0.3992713989,0.4064866689,0.4130870104,0.4189833403 0.4237552319,0.4288577135,0.4328718185,0.4363053143,0.4389313459,0.4408572912 0.4423325956,0.4430640042,0.4433769286,0.4431486130,0.4424123466,0.4413488805 0.4399740398,0.4383350313,0.4365779151,0.4346574247,0.4325914979,0.4306234717 0.4283232689,0.4259450436,0.4238215685,0.4214855731,0.4191545248,0.4169807732 0.4147732854,0.4124507606,0.4101879001,0.4075805613,0.4056442976,0.4033115804 0.4011530876,0.3986839584,0.3566791332,0.3944010735,0.3922523856,0.3902378082 0,3883418441,0.3862911531,0.3843450546,0.3825476170,0.3808482587,0.3794075847 0.3777632713,0.3764037490,0.3753945231,0.3743775785,0.3735508323,0.3728879094 0.3723782599,0.3721068799,0.3720223010,0.3719931245,0.3725113273,0.3729752600 0.3737984896,0.3747599423,0.3758974969,0.3772860467,0.3786665797,0.3802976906 0.3820087208,0.3839582801,0.2857671924,0.3881576923,0.3901884258,0.3922579885 0.3944430053,0.3966223896,0.3987578121,0.4007772777,0.4027979076,0.4045489728 0,4063098133,0.4079932570,0.4095074236,0.4109861553,0.4123948514,0.4137831330 0.4150694907.0.4162746370.0.4173251668.0.4182831049.0.4192278087.0.4201510251 0.4209794972,0.4220056236,0.4228887856,0.4237284660,0.4244770408,0.4253113270 0.4261188507,0.4267950654,0.4274532497,0.4280211031,0.4285456240,0.4289123416 0.4283576777,0.4296621084,0.4299099147,0.4299557507,0.4300192595,0.4299744964 0.4258333824,0.4296799600,0.4294674655,0.4291703701,0.4268789034,0.4285258353 0.4280676544,0.4276603460,0.4272131622,0.4265517572,0.4258557048,0.4252682924 0.4245446920,0.4238066673,0.4230799079,0.4222654700,0.4215750992,0.4208714962 0.4200963974,0.4192675054,0.4183184505,0.4174473682,0.4164970219,0.4155287743 0,4145109653,0.4135829210,0.4125835598,0.4115402102,0.4105018377,0.4034909132 0.4084769759,0.4074446559,0.4054779580,0.4053275287,0.4043481052,0.4033503532 0.4023414254,0.4014060795,0.4005324543,0.3996045589,0.3987701237,0.3978112340 0.3969756961.0.3961873055.0.3957003057.0.3951366842.0.3946704268.0.3943614066 0.3942531645,0.3943619132,0.3945775628,0.3950982988,0.3960566819,0.3973596990 0.3989907205.0.4009269178.0.4032102227.0.4059645236.0.4089819789.0.4120507240 0,4157146811,0,4200131297,0,4240331054,0,4280689657,0,4320653975,0,4360013008 0.4399047494.0.4438870549.0.4475108087.0.4508711696.0.4540425539.0.4565035999 0.4589559436,0.4609471560,0.4624712765,0.4636414647,0.4646218419,0.4653818905 0.4662405550,0.4669789645,0.4679403303,0.4690851867,0.4705319107,0.4725188315 0,4746659372,0.4771664441,0.4797306955,0.4622614193,0.4846513569,0.4862772226 0.4878613055,0.4690013337,0.4698780584,0.4902937412,0.4903680980,0.4900491238 0,4893704355,0.4885947704,0.4878634512,0.4872544110,0.4868578017,0.4866647124 0.4865402877,0.4865468740,0.4866443574,0.4868068099,0.4870768487,0.4873417914 0,4877221584,0.4881027341,0.4684932637,0.4887975454,0.4889072478,0.4887535870 0.4882849455,0.4875869559,0.4866200686,0.4857076705,0.4847748578,0.4839952886 0.4833757281,0.4828316867,0.4824351370,0.4819286466,0.4816359580,0.4813560843 0.4811488092,0.4809777439,0.4808953702,0.4809795320,0.4811027944,0.4811493754 0.4810781181.0.4812864661.0.4814873040.0.4816796753.0.4820566177.0.4826454818 0,4831400216,0,4837927520,0,4846539199,0,4855660498,0,4866005480,0.4876998663 0.4891462028,0.4505466139,0.4920505232,0.4937948287,0.4956251383,0.4975555241 0.4996665120.0.5020325184.0.5042691827.0.5066863258.0.5092632174.0.5118185282 0.5145204663.0.5172706842.0.5159568848.0.5228375793.0.5257425308.0.5286727548 0.5316403508.0.5346855521.0.5378155708.0.5408385396.0.5439858437.0.5469923615 0.5501150489,0.5531948805,0.5561989546,0.5590220094,0.5618109703,0.5644186139 0.5668233037,0.5689949589,0.5713441968,0.5732512474,0.5749572165,0.5764087439 0.5776500702,0.5785942078,0.5752701244,0.5754839595,0.5759936056,0.5798738274 0.5795146227,0.5791345835,0.5784724751,0.5776566863

,

NIR Equations Produced by Stepwise Multiple Linear Regression

.

,

2	Coef	ficient	F	Wavelength	Mat	h Tr	eati	nent
Term	1							
B(0)	m	6 8. 039						
B(1)	=	-995.086	133.12	2290.	2	24.	4.	4.
Term	2							
B(0)	=	57.140						
B(1)	=	-9740.772	77.64	1262.	1	16.	4.	4.
B(2)	=	2453.427	198.04	1662.	2	24.	4.	4.
Term	3							
B(0)	=	62.306						
B(1)	=	210,092	29.74	1910.	2	24.	4.	4.
B(2)	=	-7679.309	52.32	1262.	1	16.	4.	4.
B(3)	-	2824.252	274.05	1662.	2	24.	4.	4.
Term	4							
B(0)	=	63.108						
B(1)	175	-4580.140	37,46	1230.	2	24.	4.	4.
B(2)	=	-6863.108	41.68	1274.	1	16.	4.	4.
B(3)	=	3571.792	202.59	1662.	2	24.	4.	4.
B(4)	=	799.446	54.38	2266.	2	24.	4.	4.
Term	5							
B(0)	12	67,510						
B(1)	=	-5938.944	37.76	1262.	1	16.	4.	4.
B(2)	=	4410.001	183.22	1662.	2	24.	4.	4.
B(3)	=	917.587	68.18	2266.	2	24.	4.	4.
B(4)	=	-1108.577	16.44	1646.	2	24.	4.	4.
B(5)	=	-4583.659	40.06	1230.	2	24.	4.	4.
Term	6							
B(0)	=	74.250						
B(1)	=	3785.838	19.28	1150.	2	24.	4.	4.
B(2)		-7890.381	59.17	1274.	1	16.	4.	4.
B(3)	=	3653.640	106.30	1662.	2	24,	4.	4.
B(4)	=	1008.835	96.94	2266.	2	24.	4.	4.
B(5)	=	-1327.126	25.67	1646.	2	24.	4.	4.
B(6)	=	-4478.543	43.00	1230.	2	24.	4.	4.

-

APPENDIX 15 (cont)

	Coef	ficient	F	Wavelength	Math I	reatment
Term	a 7					
B(0)) =	62.140				
B(1)) =	1113.195	13.01	2354.	2 24	. 4. 4.
B(2)) =	-5481.905	35.40	1274.	1 16	. 4. 4.
B(3)) =	4281,137	187.92	1662.	2 24	. 4. 4.
B(4)) =	1340.853	117.65	2266.	2 24	. 4. 4.
B(5)) =	-1415.554	30.48	1646.	2 24	. 4. 4.
B(6)) ==	-4118.615	38.07	1230.	2 24	. 4. 4.
B(7)	=	1696.556	30.45	2426.	2 24	. 4. 4.
Tern	n 8					
B(0)	=	63.973				
B(1)	=	-4311.132	15.47	1842.	1 16	. 4. 4.
B(2)	=	-6062.110	35.83	1274.	1 16	. 4. 4.
B(3)	=	3244.727	93.18	1662.	2 24	. 4. 4.
B(4)		1027.880	104.43	2266.	2 24	. 4. 4.
B(5)) =	-1716.427	43,90	1646.	2 24	. 4. 4.
B(6)	=	-4355.047	46.93	1230.	2 24	. 4. 4.
B(7)	=	1921.474	23.83	2426.	2 24	. 4. 4.
B(8)	=	3245.068	14.57	1150.	2 24	. 4. 4.
Tern	ı 9					
B(0)	-	63.560				
B(1)	=	882.543	5.81	1738.	2 24	. 4. 4.
B(2)		-4228.508	15.50	1842.	1 16	. 4. 4.
B(3)	=	-6258.256	39.55	1274.	1 16	. 4. 4.
B(4)	=	3648.929	97.52	1662.	2 24	. 4. 4.
B(5)	=	988.268	97,91	2266.	2 24	. 4. 4.
B(6)	=	-2244.549	44.82	1646.	2 24	. 4. 4.
B(7)	=	-4400.404	49.90	1230.	2 24	. 4. 4.
B(8)) =	1888.862	23.98	2426.	2 24	. 4. 4.
B(9)	=	3843.929	19.57	1150.	2 24	. 4. 4.

-

The Individual Values for <u>in vivo</u> OMD, <u>in vivo</u> DOMD, Chemical and Biological Parameters for the 48 Validation Silages Used in this Study.

.

	0	Q	QMOQ						
*0	e N	8	8			•	~	-	
N N N	Lag	5	5	Ъ	Æ	IJ	Į	Į	~ _
Pol	Si	티	티	MA	LI(ğ	ğ	IV	Asl
1	77087	72.6278	63.8398	315.309	2.350	81.6837	62.04	70.1	12,10
1	78001	65.7616	58.6593	325.714	2.750	71.0762	54.18	63.8	10.80
1	78002	15.68/2	65.1345	307.339	2.350	84.1629	60.14	12.2	11.60
1	78003	79.3993	66.2136	271.431	1 000	EV. BYBY DO 7475	66.36 70 70	13.1	11.00
1	70004	75 0605	00.7700 67 50?6	2/0.4/2	2 710	DV./4/J D7 1040	10.10	73.2	0 00
1	78022	70.8724	67 1473	302.103 348 947	7 450	80.6958	60.28	13.7	10.50
	78023	72.9512	66.6045	298.523	2.580	80.764B	59.38	71.8	8.70
i	78074	59.2022	49.2562	436.700	2.900	72.7144	55.37	57.6	16.80
i	78025	73.3333	66.0000	320.808	2.480	77.5556	63.23	71.8	10.00
1	78047	73.3293	64.0898	299.062	2,390	77.2311	62.85	70.2	12.60
i	79035	71.6568	61.4099	258.665	2,156	77.7130	63.12	68.1	14.30
1	79036	76.6455	66.9882	253.969	1.930	84.8970	68.98	71.8	12.60
1	79053	76.2389	68.4625	299.512	2.530	78.3964	63.61	70.1	10.20
1	80021	73.5417	65.6727	352.083	2.590	70.3247	62.92	72.0	10.70
1	B0156	66.2402	58.0927	376.775	2,970	•	51.40	66.0	12.30
i	B0032	63.8563	56.7044	360.656	2,720	71.6216	54.12	63.2	11.20
1	80049	75.9737	67.3887	285.657	2.370	77,4521	64.25	71.7	11.30
i	B0055	73.4447	66.3940	303.592	2.570	75.6637	59.17	69.7	9.60
1	B0056	68.7237	61,7139	382.840	2.950	+	55.15	69.9	10.20
1	80099	63.1933	57,0003	354,444	2.890	68.5144	54.62	64.4	9,80
1	80169	74.1511	66.0686	512.658	2.2.9	80.5428	67.56	64.8	10,90
1	80178	62.1939	54.6962	367.010 744 467	3,120	17 0700	57.15	64.Z	0 66
1	80166	53.4730 71 5700	07.0077 64.0107	344,43/	2.000	73.0300	51.05	60.2	10.50
1	80171 80172	11.3300	64.0173 60 7766	207.109	2.700	71 1691	55 38	60.7	11.50
1	80175	61 3912	54 2698	391.490	3,170	59.8416	51.93	63.3	11.60
i	B0180	65.2946	59.1569	375.070	2.790	59.0508	55.41	65.5	9.40
3	1	64.3000	59.8926	453.000	3.520	53.7707	45.49	56.6	6.87
3	3	65.0000	59.2735	434,000	3.360	55.8668	45.45	60.0	8.81
3	6	60.9000	53,0195	352.000	2.390	68.7844	58.96	58.3	12.94
3	14	73.8000	67.9551	424.000	3.100	63.8727	53.85	62.8	7.92
3	19	66.3000	59.0070	385.000	3,450	63.6916	46.63	64.7	11.00
3	20	70.6000	64.5990	345.000	3.010	74.7372	65.85	13.1	8.50
3	21	70.8000	64.1448	349,000	2,450	71.4358	56.05	75.1	9.40
د ۲	22	78.6000	/1.1330	337.000	2.960	79.4038	10.20	70.2	10 00
3	23	73.3000	67.2703	331,000	2.050	10.1011	47.16	70.3	7 70
3 7	23	12.0000	47 3570	A14 000	7.010	45 4450	54 82	67.1	B. 70
ט ז	20	75 7000	49 9477	191 000	3.200	49 5411	59.93	71.6	8.90
7	20	71 4000	64 5116	422 000	3,195	62.2760	53,11	61.4	9,90
3	30	76.3000	69.5856	314.000	2.595	75.7597	72.31	75.2	8.80
3	32	81.5000	75,7950	336.000	2.875	76.6559	73.39	75.8	7.00
3	33	75.1000	68.7165	327.000	3,225	73.2692	69.13	72.2	8,50
3	39	74.0000	66.0820	367.000	2.950	64.1123	60.53	65.5	10.70
3	48	73.2000	68.2956	410.000	3.060	60.4990	57.13	67.2	6.70
3	50	71.4000	65.1982	381.000	3,165	61.9968	57.56	68.4	8.70
3	51	69.5000	63.6620	389,000	2.965	62.4572	52.24	62.9	8.40

*Population Numbers = See Table 2.1

-

NIR Regression Equations Produced by Stepwise Multiple Linear Regression for the Prediction of IVOMD of the 122 Calibration Silages.

(Coeff	ficient	<u>F</u>	Wavelength	Math	<u>Tre</u>	eatr	nent
Term	1							
B(O) B(1)	8	54.196 -3455.186	123.82	2106.	1	16.	4.	4.
Term	2							
B(O) B(1) B(2)	= = =	53.523 -4777.497 -4947.764	132.07 77.70	1454. 1626.	1 2	16. 24.	4. 4.	4. 4.
Term	3							
B(O) B(1) B(2) B(3)	= = =	56.183 -6458.558 -5378.479 -3489.314	58.40 235.62 63.54	1286. 1454. 1626.	1 1 2	16. 16. 24.	4. 4. 4.	4. 4. 4.
Term	4							
B(O) B(1) B(2) B(3) B(4)	= =	49.890 -2144.556 1719.540 -7156.548 -3717.746	27.80 57.34 71.54 87.27	1602. 2378. 1286. 1454.	1 2 1 1	16. 24. 16. 16.	4. 4. 4. 4.	4. 4. 4. 4.
Term	5							
B(0) B(1) B(2) B(3) B(4) B(5)		60.305 3854.424 -4859.981 -4617.785 -4420.929 -3262.613	123.14 35.38 164.55 32.15 115.83	2386. 1286. 1454. 1258. 2398.	2 1 1 2 2	24. 16. 16. 24. 24.	4. 4. 4. 4. 4.	4. 4. 4. 4. 4.
Term	6							
B(0) B(1) B(2) B(3) B(4) B(5) B(6)		53.539 2725.027 -6197.911 -5340.161 1637.996 -1766.767 -1945.402	91.02 70.36 187.13 38.88 62.50 14.36	2382. 1286. 1454. 1190. 2398. 1750.	2 1 2 2 2	24. 16. 16. 24. 24. 24.	4. 4. 4. 4. 4.	4. 4. 4. 4. 4. 4.

285

APPENDIX 17 (cont)

	Coeff	icient	F	Wavelength	Math	Treatment
Tern	n 7					
B(0) B(1) B(2) B(3) B(4) B(5) B(6) B(7)		60.389 -2220.655 3502.146 -5929.965 -3497.159 -3682.150 -1823.521 -2907.229	26.56 86.64 66.52 55.98 24.92 47.87 34.26	1550. 2378. 1286. 1454. 1782. 2390. 1750.	1 2 1 1 2 2	$\begin{array}{cccccccccccccccccccccccccccccccccccc$
Tern	n 8					
B(0) B(1) B(2) B(3) B(4) B(5) B(6) B(7) B(8)	= = = = = = = =	59.150 -3504.207 -1950.031 -3008.902 -1709.148 -2033.611 3809.010 -5480.321 -3981.551	32.55 50.85 38.36 11.07 17.08 100.52 58.69 67.52	1778. 2390. 1750. 1290. 1554. 2378. 1286. 1454.	1 2 2 1 2 1 1	$\begin{array}{cccccccccccccccccccccccccccccccccccc$
B(0) B(1) B(2) B(3) B(4) B(5) B(6) B(7) B(8) B(9)		63.340 1954.072 -3599.017 -2042.255 -3136.075 -3976.556 -2146.583 3846.112 -6218.856 -3685.737	5.57 35.57 56.82 42.81 13.44 19.61 106.44 65.57 56.28	1294. 1778. 2390. 1750. 1290. 1554. 2378. 1286. 1454.	2 1 2 2 1 2 1 1	$\begin{array}{cccccccccccccccccccccccccccccccccccc$

.

Characteristics of Silages Studied With Their Corresponding Numbers

$\begin{array}{c ccccccccccccccccccccccccccccccccccc$	Silage No	Y	MEn	CN	WT	AL	NF	Add.
78002 4 1	77087 78001	3 4	1	1	2	1	0	3
70003 4 1 1 2 2 2 70004 4 1 2 2 2 2 70004 4 1 2 2 2 2 70024 4 1 1 1 2 2 2 2 70024 4 1 1 2 2 1 2 1 70025 5 1 1 2 2 2 3 1 79035 5 1 1 2 2 1 1 3 2 2 1 1 80021 6 1 1 2 2 1 1 3 3 1 1 3 3 1 1 3 3 1 1 3 3 1 1 3 3 1 1 3 3 1 1 3 3 1 1 3 3 1 1 3 3 1 1 1 3 3 <	78002	4	1	1	2	1	د ح	1
78004 4 1 2 1 <td>78003</td> <td>4</td> <td>ī</td> <td>1</td> <td>3</td> <td>2</td> <td>ა ი</td> <td>1</td>	78003	4	ī	1	3	2	ა ი	1
7B005 4 1 0 2 0 2 0 2 7B022 4 0 <t< td=""><td>78004</td><td>4</td><td>1</td><td>2</td><td>ž</td><td>5</td><td>2</td><td>2 2</td></t<>	78004	4	1	2	ž	5	2	2 2
78022 4 0<	78005	4	1	1	ō	ž	ô	2
$\begin{array}{cccccccccccccccccccccccccccccccccccc$	78022	4	0	0	0	ō	ŏ	ō
78024 4 1 0 2 4 2 1 78047 4 1 2 2 2 1 79035 5 1 1 2 2 2 1 79035 5 1 1 3 2 2 2 1 79035 5 1 1 1 2 2 2 1 800156 6 1 1 2 2 1 1 3 80049 6 1 1 2 2 2 0 0 80056 6 1 2 2 2 1 3 3 80178 6 1 2 0 2 2 2 1 80171 6 1 1 2 2 1 1 3 80171 6 1 1 2 2 1 1 3 1 2 80172 6 1 1 2 <t< td=""><td>78023</td><td>4</td><td>1</td><td>1</td><td>1</td><td>2</td><td>3</td><td>ĭ</td></t<>	78023	4	1	1	1	2	3	ĭ
780247 4 1 1 2 1 2 1 79035 5 1 3 2 2 3 1 79035 5 1 1 3 2 2 3 1 79035 5 1 1 1 0 2 2 1 1 80021 6 1 1 2 2 1 1 3 2 2 1 1 80156 6 1 1 2 2 1 1 3 3 3 1 1 3 3 3 1 3 3 3 1 1 3 3 3 3 3 3 3 3 3 3 1 1 3 3 3 1 1 3 3 1 1 3 3 1 1 3 3 1 1 3 3 1 1 3 3 1 1 3 3 1 1	78024	4	1	0	2	4	2	1
79035 5 1 2 2 2 3 1 79036 5 1 1 3 2 2 2 1 79036 5 1 1 3 2 2 1 1 79036 5 1 1 1 2 2 1 1 80021 6 1 1 2 2 1 1 3 <td>78047</td> <td>4</td> <td>1</td> <td>1</td> <td>2</td> <td>1</td> <td>2</td> <td>1</td>	78047	4	1	1	2	1	2	1
79036 5 1 1 3 2 2 1 79035 5 1 1 1 2 2 1 80021 6 1 1 2 2 1 1 80049 6 1 2 2 1 1 3 80049 6 1 2 2 2 2 2 2 80055 6 1 2	79035	5	1	2 7	2	2	3	1
79083 5 1 1 1 2 2 1 80021 6 1 1 2 2 1 1 80156 6 1 2 2 1 1 3 80049 6 1 1 2 2 2 2 80055 6 1 2 2 2 2 2 80056 6 1 2 2 2 2 0 80179 6 1 2 2 2 2 2 0 80176 6 1 1 2 2 3 2 0 80177 6 1 1 2 2 0	79036	5	1	1	~ ~	2	2	1
B0021 6 1 1 2 2 1 1 $B0155$ 6 1 1 2 1 2 3 $B0049$ 6 1 1 2 2 1 1 3 $B0055$ 6 1 2 2 1 1 3 2 0 $B0056$ 6 1 2 2 2 0<	79053	5	1	i	1	2 0	2 5	1
80156 6 1 1 2 1 2 3 80035 6 1 2 2 1 3 80055 6 1 2 2 2 2 2 80055 6 1 2 2 2 2 2 2 80079 6 1 2 2 2 1 1 3 80176 6 1 2 2 2 1 1 3 1 2 2 2 1 1 3 1 2 2 2 1 1 3 3 1 1 3 3 1 1 3 3 1 1 3 1 1 1 2 2 2 2 2 2 2 1 1 2 2 1 1 2 2 1 1 1 2 2 1	80021	6	1	1	2	ž	1	1
B0032 6 1 2 2 1 1 3 $B0049$ 6 1 1 2	80156	6	1	1	2	1	2	Ĵ
BO0577 G I I 2 3 1 2 B00556 G I 2 2 2 0 B00566 G I 2 2 1 1 3 B01677 G I 2 2 1 1 3 3 2 2 1 1 B0177 G I 2 2 2 1 1 3 3 2 2 1 1 3 B0177 G I I 2 2 2 1 1 3 1 2 0 1 1 3 1 1 3 1 1 3 1 1 1 1 2 1 1 3 1 <td>80032</td> <td>6</td> <td>1</td> <td>2</td> <td>2</td> <td>1</td> <td>1</td> <td>3</td>	80032	6	1	2	2	1	1	3
B = 0 = 0 = 0 = 0 = 0 = 0 = 0 = 0 = 0 =	80055	6	1	1	2	3	1	2
$B00^{-0}$ 6 1 2 2 1 1 3 $B016^{-0}$ 6 1 3 2 2 1 1 $B0176$ 6 1 2 2 2 1 1 $B0176$ 6 1 2 2 2 3 2 $B01776$ 6 1 1 2 2 3 2 $B0175$ 6 1 1 3 1 2 0 $B0175$ 6 1 1 2 2 0 0 $B01022$ 6 1 1 2 2 0 0 $B00170$ 6 1 2 2 3 1 1 $B1005$ 7 1 1 2 2 2 1 1 $B1006$ 7 1 1 2 2 2 1 1 $B1005$ 7 <	80056	6	1	2	2	2	2	0
B0169 6 1 2 2 1 1 3 3 B0176 6 1 2 0 2 2 1 1 3 B0176 6 1 2 2 2 1 1 2 2 1 1 3 B0175 6 1 1 2 2 2 1 1 B0180 6 0 0 0 0 0 0 0 B0175 6 1 1 2 2 2 1 1 B0180 6 0 0 0 0 0 0 0 B0170 6 1 1 2 1 1 1 2 3 1 1 B10031 6 2 1 3 3 1 1 2 1 1 2 2 2 1 1 2 2 2 1 1 1 2 1 1 1 2 2	80099	6	1	2	2	2	2	0
80178 6 1 2 2 2 1<	80169	6	1	3	2	1	1	د
B0146B 6 1 2 2 1 1 B0171 6 1 0 2 2 3 1 B0172 6 1 1 2 2 3 1 2 0 B0175 6 1 1 3 1 2 0 0 B0175 6 1 1 2 2 1 1 2 1 1 2 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 2 1	80178	6	1	2	ō	2	2	1
B01/1 6 1 0 2 2 3 2 B0175 6 1 1 2 2 2 1 B0175 6 1 1 3 1 2 0 0 B0180 6 0 0 0 0 0 0 0 0 B0175 6 1 1 2 2 3 1 1 B0022 6 1 1 2 2 3 1 1 B003170 6 1 3 2 3 1 1 B1003 6 2 1 3 3 0 1 B1003 6 2 1 3 3 0 1 B1003 6 2 1 3 3 0 1 B1004 7 1 1 2 2 1 1 2 B1012 7 1 1 2 1 1 2 2 1 <td>80168</td> <td>6</td> <td>1</td> <td>2</td> <td>2</td> <td>ź</td> <td>ĩ</td> <td>ĭ</td>	80168	6	1	2	2	ź	ĩ	ĭ
60172 6 1 1 2 2 2 1 80180 6 0 0 0 0 0 0 80180 6 1 1 3 1 1 2 0 0 80022 6 1 1 2 2 0 0 80170 6 1 2 2 3 1 1 80170 6 1 2 2 3 1 1 80177 6 1 2 2 3 1 1 80177 6 1 2 2 3 1 1 80177 6 1 2 2 3 1 1 81005 7 1 1 2 2 1 1 1 2 2 1 1 1 2 2 2 2 2 2 2 <t< td=""><td>80171</td><td>6</td><td>1</td><td>0</td><td>2</td><td>2</td><td>3</td><td>2</td></t<>	80171	6	1	0	2	2	3	2
B01130 B 1 1 3 1 2 0 B01430 6 1 1 3 1 2 0 0 B0170 6 1 1 2 2 0 0 0 B0170 6 1 2 2 3 1 1 2 3 1 1 B1005 7 1 2 2 3 1	80172	6	1	1	2	2	2	1.
79047 5 1 1 3 1 1 80022 6 1 1 2 2 0 80170 6 1 2 3 2 3 80170 6 1 3 3 1 1 81003 6 2 1 3 3 0 1 81003 6 2 1 3 3 0 1 81003 6 2 1 3 3 0 1 81003 6 2 1 3 3 0 1 81004 7 1 1 2 2 2 3 81007 7 1 1 2 2 2 3 3 1 1 81020 7 1 1 2 2 2 3 2 1 82004 8 1 2	80180	6	1	1	3	1	2	0
B0022 6 1 1 2 2 0 0 $B0031$ 6 1 1 2 1 2 3 1 1 $B0170$ 6 1 2 2 3 1 </td <td>79047</td> <td>5</td> <td>1</td> <td>1</td> <td>0 7</td> <td>Q 7</td> <td>0</td> <td>0</td>	79047	5	1	1	0 7	Q 7	0	0
60031 6 1 1 2 2 3 2 3 80170 6 1 2 2 3 2 0 81003 6 2 1 2 2 3 1 1 81003 6 2 1 2 2 2 1 1 81006 7 1 1 2 2 2 1 1 81006 7 1 1 2 2 2 1 1 81007 7 1 1 2 2 2 2 3 3 1 1 81007 7 1 0 2 2 3 3 1 1 2 2 3 3 1 1 2 2 3 3 1 1 2 2 3 3 1 1 1 1 1 1 1 <td>80022</td> <td>6</td> <td>1</td> <td>1</td> <td>2</td> <td>2</td> <td>1</td> <td>1</td>	80022	6	1	1	2	2	1	1
B0170 6 1 2 2 3 2 0 B0177 6 1 3 2 3 1 1 B1005 7 1 2 2 2 1 1 B1005 7 1 2 2 2 1 1 B1006 7 1 1 2 2 2 1 1 B1007 7 1 1 2 2 2 3 3 1 1 2 2 3 3 3 1 1 2 2 2 3 3 1 1 2 2 3 3 1 1 3 3 1 1 1 2 3 3 1 1 1 3 3 1 1 1 3 3 1	80031	6	1	1	2	~ 1	2	र
80179 6 1 3 2 3 1 1 810003 6 2 1 3 3 0 1 810005 7 1 2 2 1 1 81006 7 1 1 2 2 2 1 81007 7 1 1 2 2 2 3 81012 7 1 1 2 1 1 2 81014 7 1 1 2 1 1 2 81012 7 1 0 2 0 1 1 82004 8 1 2 2 2 3 1 1 82005 8 2 1 3 4 2 1 1 82005 8 2 1 3 4 2 1 82005 8 1 2 2 3 2 2 82014 8 1 0	80170	6	1	2	2	3	ĩ	ŏ
B1003 6 2 1 3 3 0 1 $B1005$ 7 1 2 2 2 1 1 $B1007$ 7 1 1 2 2 2 1 $B1012$ 7 1 1 2 2 2 3 $B1014$ 7 1 1 2 2 1 1 2 $B1014$ 7 1 1 2 1 1 2 $B1020$ 7 1 0 2 0 1 1 2 $B2004$ B 1 2 2 1 3 3 1 1 $B2005$ B 2 1 3 4 2 1 $B2014$ B 2 1 2 2 3 2 2 3 2 2 3 2 2 2 2 2 2 <t< td=""><td>80179</td><td>6</td><td>1</td><td>3</td><td>2</td><td>3</td><td>1</td><td>1</td></t<>	80179	6	1	3	2	3	1	1
31006 7 1 1 2 2 2 1 1 81007 7 1 1 2 2 2 1 1 81012 7 1 1 2 2 2 3 81014 7 1 1 2 1 1 2 81020 7 1 0 2 0 1 1 82004 6 1 2 2 2 2 3 82005 8 2 0 2 4 1 1 82005 8 2 1 3 4 2 1 82013 8 2 1 3 4 2 1 82014 8 1 2 2 3 2 2 82016 8 1 2 2 3 2 2 82022 8 1 0 3 2 2 2 82023 8 2	81005 81005	6 7	2	1	3	3	0	1
1 1 1 2 2 2 1 1 1 1 1 1 1 1 2 1 1 1 1 1 1 2 2 2 3 1 1 2 1 1 2 2 2 2 3 1 1 2 2 2 2 2 2 3 1 2 2 2 2 2 2 3 1 2 2 2 2 2 2 3 2 2 2 2 2 2 3 2 1 1 2 2 3 4 2 1 1 1 2	81006	7	1	2	2	2	1	1
B1012 7 1 1 2 1 1 2 3 B1014 7 1 1 2 1 1 2 1 1 2 B1014 7 1 1 2 1 1 2 1 1 2 B1020 7 1 0 2 0 1 1 2 B2004 8 1 2 2 2 2 3 3 1 1 B2005 B 2 0 2 4 1 1 B2005 B 2 1 3 3 1 1 B2013 B 2 1 3 4 2 1 B2014 B 2 1 3 4 2 1 B2015 B 1 2 2 3 2 2 B2022 B 1 0 0 3 2 2 2 B2031 B 2 1	81007	7	1	1	2	2	2	1
B10147112112 $B1020$ 7102011 $B2004$ 8122223 $B2005$ 8202411 $B2008$ 8213311 $B2013$ 8213421 $B2014$ 8213421 $B2015$ 8122322 $B2016$ 8122322 $B2022$ 8100022 $B2025$ 8211431 $B2033$ 8213421 $B2033$ 8213421 $B2033$ 8213222 $B2045$ 8101222 $B2074$ 8000000 $B2079$ 8110333 $B3013$ 9112210 $B3024$ 9110402	81012	7	1	1	2	2	2	3
B10207102011 $B2004$ B 122223 $B2005$ B 202411 $B2008$ B 213311 $B2013$ B 213421 $B2014$ B 213421 $B2015$ B 122322 $B2016$ B 122322 $B2022$ B 100022 $B2025$ B 211431 $B2027$ B 102122 $B2033$ B 213421 $B2033$ B 213321 $B2045$ B 101222 $B2074$ B 000000 $B2079$ B 110333 $B3013$ 9 112213 $B3013$ 9 110210 $B3024$ 9 110402	81014	7	1	1	ź	1	1	2 2
82004 8 1 2 2 2 2 2 3 82005 8 2 0 2 4 1 1 82008 8 2 1 3 3 1 1 82013 8 2 1 3 4 2 1 82014 8 2 1 3 4 2 1 82015 8 1 2 2 3 2 2 82016 8 1 2 2 3 2 2 82016 8 1 0 0 3 2 2 82025 8 2 1 0 1 2 2 2 2 2 1 82027 8 1 0 1 2 2 2 2 2 2 2 2 2 2 2 2 2 <t< td=""><td>81020</td><td>7</td><td>1</td><td>0</td><td>2</td><td>ō</td><td>1</td><td>1</td></t<>	81020	7	1	0	2	ō	1	1
82005 8 2 0 2 4 1 1 82008 8 2 1 3 3 1 1 82013 8 2 1 3 4 2 1 82014 8 2 1 3 4 2 1 82015 8 1 2 2 3 2 2 82016 8 1 2 2 3 2 2 82022 8 1 0 0 0 2 2 82025 8 2 1 1 4 3 1 82027 8 1 0 2 1 2 2 82033 8 2 1 3 4 2 1 82033 8 2 1 3 3 2 1 82045 8 1 0 1 2 2 2 82074 8 0 0 0 3 3	82004	8	1	2	2	2	2	3
B = 200B $B = 2$ 1 3 3 1 1 $B = 2013$ $B = 2$ 1 3 4 2 1 $B = 2015$ $B = 1$ 2 2 3 2 2 $B = 2015$ $B = 1$ 2 2 3 2 2 $B = 2016$ $B = 1$ 2 2 3 2 2 $B = 2022$ $B = 1$ 0 0 0 2 2 $B = 2025$ $B = 2$ 1 1 4 3 1 $B = 2025$ B 2 1 3 4 2 1 $B = 20237$ B 1 0 2 1 2 2 $B = 2033$ B 2 1 3 3 2 1 $B = 2057$ B 1 0 1 2 2 2 $B = 2045$ B 1 0 1 2 2 2 $B = 2074$ B 0 0 0 0 0 0 $B = 2074$ B 0 0 0 0 0 0 $B = 2074$ B 0 0 0 3 3 $B = 30013$ 9 1 1 0 3 3 $B = 3013$ 9 1 1 0 4 0 2 $B = 3014$ 9 1 1 0 4 0 2 $B = 3024$ 9 1 1 0 4 0	82005	8	2	0	2	4	1	1
B = 2 1 3 4 2 1 $B = 2014$ B 2 1 3 4 2 1 $B = 2015$ B 1 2 2 3 2 2 $B = 2016$ B 1 2 2 3 2 2 $B = 2022$ B 1 0 0 0 2 2 $B = 2025$ B 2 1 1 4 3 1 $B = 2027$ B 1 0 2 1 2 2 $B = 2027$ B 1 0 2 1 2 2 $B = 2033$ B 2 1 3 4 2 1 $B = 2057$ B 1 0 1 2 2 2 2 $B = 2045$ B 1 0 1 2 2 2 2 $B = 2074$ B 0 0 0 0 3 3 3 $B = 30013$ 9 <td>82008</td> <td>8</td> <td>2</td> <td>1</td> <td>3</td> <td>3</td> <td>1</td> <td>1</td>	82008	8	2	1	3	3	1	1
B2015 B 1 2 1 3 4 2 1 $B2015$ B 1 2 2 3 2 2 $B2016$ B 1 2 2 3 2 2 $B2022$ B 1 0 0 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 1 4 3 1 2 2 2 1 2 2 1 2 2 1 2 2 2 1 2 2 2 1 2 2 1 2	82014	8	2	1	3	4	2	1
B2016 B 1 2 3 2 2 $B2022$ B 1 0 0 0 2 2 2 $B2022$ B 1 0 0 0 2 1 2 2 2 1 2 2 1 2 2 1 2 2 1 2 2 1 2 2 1 2 2 1 2 2 1 2 <td>82015</td> <td>8</td> <td>1</td> <td>2</td> <td>ు ా</td> <td>4 →</td> <td>2</td> <td>1</td>	82015	8	1	2	ు ా	4 →	2	1
B2022 B 1 0 0 2 2 $B2023$ B 0 1 0 3 2 2 $B2025$ B 2 1 0 3 2 2 $B2025$ B 2 1 0 2 1 2 2 $B2037$ B 2 1 3 4 2 1 $B2033$ B 2 1 3 4 2 1 $B2033$ B 2 1 3 3 2 1 $B2057$ B 1 0 1 2 <td>82016</td> <td>8</td> <td>ĩ</td> <td>2</td> <td>2</td> <td>्र र</td> <td>2</td> <td>2</td>	82016	8	ĩ	2	2	्र र	2	2
$\begin{array}{c ccccccccccccccccccccccccccccccccccc$	82022	8	1	õ	õ	ŏ	2	2
B2025 B 2 1 1 4 3 1 B2027 B 1 0 2 1 2 2 2 B2031 B 2 1 3 4 2 1 B2033 B 2 1 3 4 2 1 B2057 B 1 0 1 2 2 2 B2045 B 1 0 1 2 2 2 2 B2074 B 0 0 0 0 3<	82023	8	0	1	ò	š	2	2
$\begin{array}{c ccccccccccccccccccccccccccccccccccc$	82025	8	2	1	1	4	3	ĩ
B2031 B 2 1 3 4 2 1 B2033 B 2 1 3 3 2 1 B2033 B 2 1 3 3 2 1 B2057 B 1 0 1 2 2 2 B2045 B 1 0 1 2 2 2 B2074 B 0 0 0 0 0 0 B2079 B 1 1 0 0 3 3 B3007 7 1 1 2 2 2 3 B30013 9 1 1 2 2 1 0 B3014 9 1 1 0 2 1 0 B3021 9 1 1 1 3 2 0 B3026 7 1 1 0 4 0 2	82027	8	1	0	2	1	2	2
B2057 B 1 0 1 2 2 2 B2057 B 1 0 1 2 2 2 B2045 B 1 0 1 2 2 2 B2074 B 0 0 0 0 0 0 B2079 B 1 1 0 0 3 3 B3007 7 1 1 2 2 2 3 B3008 9 1 2 2 1 3 B3013 9 1 1 2 2 1 0 B3014 9 1 1 0 2 1 0 B3021 9 1 1 3 2 0 B3026 7 1 1 0 4 0 2	82031	8	2	1	3	4	2	1
B2045 B 1 0 1 2 2 2 B2074 B 0 0 1 2 2 2 B2074 B 0 0 0 0 0 0 B2079 B 1 1 0 0 3 3 B3007 7 1 1 2 2 2 3 B300B 9 1 2 2 2 1 3 B3013 9 1 1 2 2 1 0 B3014 9 1 1 0 2 1 0 B3021 9 1 1 1 3 2 0 B3026 9 1 1 0 4 0 2	82057	8	2	1	د .	3	2	1
B2074 B 0 3 3 3 0 0 0 0 3 3 0 0 0 0 3 3 3 0 0 0 0 1 1 1 2 2 2 3 3 0 0 1 1 3 1 1 1 1 0 1 1 0 1 0 1 0 1 0 1 0 1 0 1 0 2 0 0 1 0 1 0 1 0 1 0 1 0 1 0 1 0 1 0 1 0 1 0 1 0 1 <th1< th=""></th1<>	82065	8	1	ŏ	1 1	2	2	2
82079 8 1 1 0 0 3 3 83007 7 1 1 2 2 2 3 83008 9 1 2 2 2 1 3 83013 9 1 1 2 2 1 3 83014 9 1 1 0 2 1 0 83021 9 1 1 3 2 0 83026 9 1 1 0 4 0 2	82074	8	ō	ō	ô	ô	0	2 0
B3007 7 1 1 2 2 2 3 B300B 9 1 2 2 2 1 3 B3013 9 1 1 2 2 1 3 B3013 9 1 1 2 2 1 0 B3014 9 1 1 0 2 1 0 B3021 9 1 1 1 3 2 0 B3026 9 1 1 0 4 0 2	82079	8	1	1	0	ō	ž	ž
H300B Y 1 2 2 2 1 3 B3013 Y 1 1 2 2 1 3 B3013 Y 1 1 2 2 1 0 B3014 Y 1 1 0 2 1 0 B3021 Y 1 1 3 2 0 B3026 Y 1 1 0 4 0	83007	9	1	1	2	2	2	3
B3013 7 1 1 2 2 1 0 B3014 9 1 1 0 2 1 0 B3021 9 1 1 3 2 0 B3026 9 1 1 0 4 0	83008	9	1	2	2	2	1	3
B3021 9 1 1 0 2 1 0 B3021 9 1 1 1 3 2 0 B3026 9 1 1 0 4 0 2	83014	9	1	1	2	2	1	0
	83021	9	1	1	0	2 7	1	0
, , ,	83026	7	1	1	ò	4	∠ 0	2

APPENDIX 18 (cont)

Silage No	Y	MEn	CN	WT	AL	NF	Add.
83027	9	1	0	2	3	2	3
83031	9	1	1	2	Ā	1	ĩ
83036	9	0	ō	ō	ò	ō	Ô
83049	9	1	ō	ō	ŏ	2	ž
83045	9	1	ō	õ	ž	5	~
83110	9	1	õ	ž	ō	ž	्र
84018	10	1	1	2	ž	1	र
84074	10	2	1	3	ĩ	2	1
84087	10	2	ā	3	ī	2	1
85016	11	1	õ	2	Â	2	2
85017	11	1	2	2	3	1	3
85018	11	1	3	2	4	ō	ž
85022	11	2	2	2	2	Ĵ	1
85024	11	1	2	2	1	ź	2
85026	11	1	1	2	3	1	1
85030	11	2	1	2	4	0	1
85033	11	2	1	2	1	2	1
85036	11	1	1	0	0	ō	ō
85037	11	1	2	0	Q	Ó	Ō
85038	11	0	Q	0	0	0	0
85041	11	2	1	2	1	1	1
85043	11	2	2	3	0	1	1
85048	11	2	2	2	2	1	1
85057	11	0	0	0	0	0	0
86005	12	0	0	0	0	0	0
86015	12	1	2	1	4	1	0
86016	12	1	1	1	4	2	3
86018	12	1	1	1	4	2	3
86020	12	1	1	2	3	2	3
86021	12	1	1	1	4	2	3
86026	12	1	1	2	2	2	3
86029	12	2	3	1	4	1	1
86033	12	1	3	2	2	2	0
86036	12	1	0	1	2	1	2
86037	12	2	0	0	0	0	1
86042	12	2	2	0	3	1	1
86067	12	0	0	0	0	0	0

Y = Year of harvest - 3 = 1977; 4 = 1978; 5 = 1979; 6 = 1980; 7 = 1981; 8 = 1982; 9 = 1983; 10 = 1984; 11 = 1985; 12 = 1986.

MEn = Method of Ensiling -0 = No data; 1 = clamp; 2 = big bale.

CN = Cut number - 0 = No data; 1 = first cut; 2 = second cut; 3 = third cut.

WT = wilting time - 0 = No data; $1 \le 12$ hrs; 2 = > 12-40 hrs; 3 = > 40 hrs.

AL = Age of Ley (Years) - 0 = No data; 1 = ≤ 1; 2 = > 1-5; 3 = > 5; 4 = Permanent Pasture.

NF = Nitrogen Fertilisation (KgN/ha) - 0 = No data; 1 = < 100; 2 = > 100-200; 3 = >200.

Add = Additive Use - 0 = No data; 1 = None; 2 = Add-F; 3 = Sylade + F100.

Analysis of Variance of the Effect of Different Characteristics of Silages on the <u>in vivo</u> OMD **:** Laboratory Methods Relationships.

Chara- cteri- stics	Predic Models	tion	MADF	LIGA	NCOMD	PCOMD	IVOMD	NIR 8-TERM
	Model	DF	84	84	81	84	84	77
	1	MS	17.47	10.95	20.59	9.508	10 . 74	5.297
Years	Model.	DF	79	79	76	79	79	72
	2	MS	16.39	11.39	14.33	9.261	7.752	5.255
	Model	DF	74	74	71	74	74	32
	3	MS	16.46	11 . 12	13.18	8.754	7.743	3.319
siling	Model	DF	89	89	86	89	89	82
	1	MS	18.24	11.16	19.24	10 . 28	10,62	5.194
d of Ens	Model.	DF	88	88	85	88	88	81
	2	MS	18.28	11.26	19 . 43	10.31	10.46	5.229
Metho	Model.	DF	87	87	84	87	87	73
	3	MS	17.89	11.38	18.84	10.41	10.54	5.151
	Model	DF	75	75	73	75	75	68
	1	MS	18.31	9.453	16.81	9.315	10.41	5.117
t Number	Model	DF	73	73	71	73	73	66
	2	MS	16.91	7.771	16.62	8.883	10.43	5.100
Cut	Model	DF	71	71	69	71	71	52
	3	MS	16.48	7.918	16.95	9.127	10.37	4.996
۵	Model	DF	76	76	74	76	76	69
۵	1	MS	18.59	10.89	17.36	10 .77	9.596	5.350
ting Tim	Model	DF	74	74	72	74	74	67
	2	MS	1 6. 99	10.92	16.91	10.41	9.372	5.271
Wil	Model	DF	72	72	70	72	72	51
	3	MS	17.33	10.95	16.36	10 . 62	9 . 257	5.262

289

APPENDIX 19 (cont)

.

Chara- cteri- stics	Predic Models	tion	MADF	LIGA	NCOMD	PCOMD	IVOMD	NIR 8-TERM
	Model.	DF	80	80	77 .	80	80	73
		MS	18.43	9.537	18.78	10.48	9.720	4.998
Ley	Model	חד	77	77	· . · · . · · . · · . · · . · · · · · ·	. 77 5		70 `
e of	2	MS	18.99	9,736	18.69	- 10.70	8.936	4.815
Ag	Model	DF	74	74	71	74	74	46
	3	MS	19.03	9.624	18.74	10.99	8.760	4.592
u	Model	DF	77	77	75	77	77	70
cati	1	MS	19.12	11.62	18.29	10.95	10.49	5.381
ilqc	Model	DF	75	75	73	75	75	68
litive A _l	2	MS	18.33	11.37	17.13	10.65	10.57	5.237
	Model	DF	73	73	71	73	73	52
Ado	3	MS	18.62	11.55	17.30	10.93	10.71	5.078
<u></u>	Mode1	DF	80	80	77	80	80	73
Ю		MS	16.60	10.42	18.29	10.25	11.23	5.335
sati	Model.	DF	78	78	75	78	78	71
rtili	2	MS	16.79	10.41	18.61	9,915	11.24	5.475
Ч	Mode1	DF	76	76	73	76	76	55
Z	3	MS	15.82	10.59	18.66	9.623	11.46	6.286

NIR Regression Equations Produced by Stepwise Multiple Linear Regression for the Prediction of <u>in vivo</u> DOMD of the 122 Calibration Silages.

G	oeff	icient	F	Wavelength	<u>Matl</u>	n Treatment
Term	1					
B(O) B(1)	=	62.116 -894.834	113.49	2290.	2	24.4.4.
Term	2					
B(0) B(1) B(2)	= = =	61.536 -3985.890 -1191.807	37.72 183.50	1198. 2290.	1 2	16.4.4. 24.4.4.
Term	3					
B(0) B(1) B(2) B(3)	= = =	52.448 1802.993 -6231.985 -1039.753	51.35 90.77 165.87	1666. 1206. 2290.	2 1 2	24. 4. 4. 16. 4. 4. 24. 4. 4.
Term	4					
B(0) B(1) B(2) B(3) B(4)		70.706 -5956.012 -880.455 1404.923 3953.349	86.37 116.60 22.16 79.66	1210. 2290. 2258. 1666.	1 2 1 2	16. 4. 4. 24. 4. 4. 16. 4. 4. 24. 4. 4.
Term	5					
B(0) B(1) B(2) B(3) B(4) B(5)		69.150 2299.533 -6583.203 -846.068 1386.306 3718.450	6.18 94.85 109.16 22.52 70.26	1254. 1210. 2290. 2258. 1666.	2 1 2 1 2	$\begin{array}{cccccccccccccccccccccccccccccccccccc$
Term	6					
B(0) B(1) B(2) B(3) B(4) B(5) B(6)		56.523 3717.525 -4786.052 2203.421 6430.177 -5966.962 -5192.953	143.87 169.17 20.57 29.84 87.29 64.04	1666. 1550. 1194. 1254. 1210. 1842.	2 1 2 2 1 1	24. 4. 4. 16. 4. 4. 24. 4. 4. 24. 4. 4. 16. 4. 4. 16. 4. 4.

APPENDIX 20 (cont)

<u>(</u>	Coef	ficient	F	Wavelength	Math	Treatment
Term	7					
B(0)	=	40.814				
B(1)	==	-1895.510	23.20	1638.	2 2	4. 4. 4.
B(2)	=	914.450	100.11	2270.	2 2	4. 4. 4.
B(3)	=	6074.395	27.61	1254.	2 2	4. 4. 4.
B(4)		-4981.553	54.88	1214.	1 1	6.4.4.
B(5)	=	-4516.571	19.17	1842.	1 1	6.4.4.
B(6)	=	3192.590	80.70	2422.	2 2	4.4.4.
B(7)	=	1848.087	27.88	1666.	22	4.4.4.
Term	8					
B(0)	=	42.377				
B(1)	=	609.605	20.88	1402.	2 2	4. 4. 4.
B(2)	=	-2038.191	31.27	1638.	2 2	4. 4. 4.
B(3)	=	1000.800	134.09	2270.	2 2	4.4.4.
B(4)	==	5515.801	26.39	1254.	2 2	4. 4. 4.
B(5)	=	-4694.963	56.67	1214.	1 1	6.4.4.
B(6)	=	-5064.583	27.86	1842.	1 1	6.4.4.
B(7)	=	2551.743	51.18	2422.	2 2	4. 4. 4.
B(8)	=	2136.619	42.15	1666.	2 2	4.4.4.
Term	9					
B(0)	=	47.371				
B(1)	=	-2402.929	14.36	1226.	2 2	4.4.4.
B(2)	=	-5361.297	28.83	1842.	1 1	.6.4.4.
B(3)	=	2580.007	49.89	2422.	2 2	4. 4. 4.
B(4)	=	1472.308	21.84	1666.	2 2	4.4.4.
B(5)	H	-5306.342	44.82	1246.	1 1	6.4.4.
B(6)	=	648.392	21.84	1402.	2 2	24.4.4.
B(7)	=	-2582.279	52.21	1638.	22	24.4.4.
B(8)	=	857.777	96.35	2270.	22	4. 4. 4.
B(9)	=	5986.215	23.95	1254.	22	24.4.4.

NIR Regression Equations Produced by Stepwise Multiple Linear Regression for the Prediction of Ash Content of the 122 Calibration Silages

,

<u>c</u>	Coef	ficient	F	Wavelength	Mat	h Tre	eatr	nent
Term	1							
B(0)	=	9.814						
B(1)	=	1126.983	53.20	1354.	2	24.	4.	4.
Term	2							
B(0)	=	11.677						
B(1)	=	-839,474	21.53	1746.	2	24.	4.	4.
B(2)	=	977.457	44.59	1354.	2	24.	4.	4.
Term	3							
B(0)	=	11.416						
B(1)	=	1163.756	33.19	1350.	2	24.	4.	4.
B(2)	=	-463.117	30.41	1970.	2	24.	4.	4.
B(3)	=	-1052.198	38.20	1730.	1	16.	4.	4.
Term	4							
B(0)	=	10.978						
B(1)	=	501,344	10.23	1654.	1	16.	4.	4.
B(2)	=	1247.942	40.41	1350.	2	24.	4.	4.
B(3)	=	-428.700	27.60	1970.	2	24.	4.	4.
B(4)	=	-1594.957	45.69	1730.	1	16.	4.	4.
Term	5							
B(0)	=	9.924						
B(1)	=	-1690.048	40.29	1650.	1	16.	4.	4.
B(2)	=	1408.094	60.94	1350.	2	24.	4.	4.
B(3)	==	-511.948	47.90	1970.	2	24.	4.	4.
B(4)	=	-2638.294	82,70	1730.	1	16.	4.	4.
B(5)	=	-570.620	29.93	1594.	2	24.	4.	4.
Term	6							
B(0)	=	9.779						
B(1)	=	-2045.767	58.82	1650.	1	16.	4.	4.
B(2)	=	858.485	13.45	1350.	2	24.	4.	4.
B(3)	=	-720.777	61.39	1966.	2	24.	4.	4.
B(4)	=	-2784.137	99.19	1730.	1	16.	4.	4.
B(5)	=	-722.736	44.62	1594.	2	24.	4.	4.
B(6)	55	1011.146	14.75	1310.	1	16.	4.	4.
/					-	•	-	

.

APPENDIX 21 (cont)

<u>(</u>	Coeff	icient	F	Wavelength	<u>Math</u> T	reatment
Term	7					
B(0) B(1) B(2) B(3) B(4) B(5) B(6) B(7)		$10.279 \\ -65.058 \\ -1865.691 \\ 795.848 \\ -729.128 \\ -2615.032 \\ -688.541 \\ 1429.274$	5.97 47.27 11.91 65.44 85.80 41.52 21.34	1898. 1650. 1350. 1966. 1730. 1594. 1310.	2 24 1 16 2 24 2 24 1 16 2 24 1 16 1 16	$\begin{array}{cccccccccccccccccccccccccccccccccccc$
Term	8					
B(0) B(1) B(2) B(3) B(4) B(5) B(6) B(7) B(8)	= = = = = = = = =	$11.342 \\ -854.734 \\ -809.134 \\ -893.614 \\ 1041.033 \\ -646.456 \\ 1563.721 \\ 1284.778 \\ 565.039 \\ \end{array}$	50.21 51.14 41.18 67.04 24.17 22.13 26.06 48.43	1890. 1966. 1730. 1426. 1398. 1298. 1866. 1910.	1 16 2 24 1 16 2 24 1 16 1 16 1 16 1 16	$\begin{array}{cccccccccccccccccccccccccccccccccccc$
B(0)	-	8.290	E0.05	1010	1 10	<i>1</i> , ,
B(2) B(3) B(4) B(5) B(5) B(6) B(7) B(8) B(9)		-590.868 -980.446 -796.440 -1250.521 1136.707 -754.803 1331.748 1737.466	13.43 67.29 54.94 58.01 84.76 34.63 17.13 41.74	1650. 1890. 1966. 1730. 1426. 1398. 1298. 1866.	$ \begin{array}{cccccccccccccccccccccccccccccccccccc$	4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4

Analysis of Variance of the Prediction of in vivo DOMD Using Various Routes by NIR

1 MS uncorrected for bias = Σ (Actual DOMD - Predicted DOMD)² Α Ν Ν 2 MS corrected for bias = Σ (Actual DOMD - Predicted DOMD - Bias)² В N - 1 N - 1 3 MS corrected for intercept and slope = Σ (Actual DOMD - Intercept - Slope x Predicted DOMD)² С N - 2 N - 2 $\frac{C}{Error MS to do F Tests} = ---- = D$ С (N - 2) F Tests $\frac{(A - B)}{D}$ at DF of 1,46 To Compare 1 with 2 = (A - C) ----- at DF of 2,46 To Compare 1 with 3 =2 * D (B - C) ----- at DF of 1,46 To Compare 2 with 3 =D

295

APPENDIX 22 (cont)

Analysis of Va	ariance Table
----------------	---------------

.

	<u></u>				с	alculated	F
Routes	А	В	С	D	1 & 2	1 & 3	2 & 3
1	424.60	423.35	419.2	9.11	0.137	0.296	0.456
2	395.15	345.35	343.16	7.46	6.680	3.480	0.293
3	284.45	264.23	257.40	5.59	3.620	2,420	1.220

The Individual Values for <u>in vivo</u> DE, Chemical and Biological Parameters Used for the Prediction of <u>in vivo</u> DE.

*0	A)	<u></u> 21					
N	age	Z	Ē	4	Ą	Ą	A
Ъ	Цо	- с ц ц	(AD)	IG/	g	ଟ୍ରି	ĝ
	77007	•H []	≥; ***	Ц	Z	д	I
1	77087	14.7108	315,309 795 714	2,350	81.6837	62,04	70.10
1	78002	15,2107	323,/14	2.759	/1.0/62	54.18	63.60
1	78003	15 3027	201.037	2.339	84.1529 04.0500	60.19	72.20
1	78004	17.2415	271.431	1 990	00.0107 00.7475	55.35 70.70	13.10
ī	78005	16.4555	302.103	2 710	87.4948 87.4948	67 21	13.20
1	78022	13,7956	348.947	2.450	80.6958	602,21 60 28	72.70
1	78023	16.2152	298,523	2.580	80.2848	59.38	71.90
1	7802 4	10.7836	436.700	2.500	72,7164	55.37	57.60
1	78025	16.7368	320.808	2,480	77,5555	63.23	71.80
1	78047	14.0869	299.062	2,390	77.2311	62.85	70.20
1	79035	14.3929	258.645	2.150	77.7130	63.12	68.10
1	79036	16.3296	253,769	1.930	64.8770	68.98	71.80
1	79053	14.0878	299,512	2.530	78.3964	63.61	70.10
1	80021	14.2350	352,083	2,590	70.3247	62.92	72.00
1	69156 00070	14.5001	355.775	2.970	+	51.40	66.00
1	80037 80037	15.12/2	360.656	2.720	71.6216	54.12	63.20
1	00049	14.3833	203.65/	2.370	77.4521	64.25	71.70
1	00000 R0057	10.0307	303,082 703,044	2.270	15.6657	59.17	69.70
1	80099	14 7667	302.040 754 AAA	71737 71737	10 5144	55.15	69.90
i	80169	17.3772	317.499	2.070	001J144 Dù 4050	24.0Z	64.4V
i	80178	11.1271	367.015	1,120	64 0592	52 13	67.8V
1	80168	13.2132	344.457	2,580	73.0360	53.67	64.20 66 20
1	80171	15.3000	359.934	2,900	77.0950	51.95	65.70
1	80172	14.1183	392.109	2.850	71.1691	55.38	68.90
1	80175	12.3255	391.490	3.170	59.8416	51.93	63.30
2	79047	14.2556	330,000	2.745	71.7174	64.68	66.31
2	80022	15.0782	362.000	2.790	68.7019	63.89	70.80
2	80031	14.0526	415.000	2.801	54.8235	51.38	62.78
2	89170	14.8684	330,000	3.034	67.2223	57.72	61.45
2	81003	9 5316	434 666	7,170	76.0.00 47 0711	70.88	69.68
2	81005	15.1828	325 000	3,017 2,603	1),73777 77 3017	20.03	97.53
2	81006	15.0258	283 000	2.003	72.0040	67.10	70.01
2	81007	13.0605	353.000	3,105	LI 1457	59 71	07.1/ 13.15
2	81012	15.7175	261.000	1.817	84.7389	BL 23	15.45 75.06
2	81014	12.0392	385.000	3.081	56, 3630	55.34	60.61
2	81020	13.7589	318,000	2.761	63.0074	62.01	64.77
2	82004	15.6271	399.000	2.977	55.8652	54.38	60.72
2	82005	14.0136	385,000	3,176	ł	47.53	56.71
2	82008	11.8367	348,000	2.717	59.6340	53.28	60.95
2	82013	13.8817	368.000	2,913	65.1036	56.25	61.68
2	82014	12.218/	392,000	3.013	62.8944	54.66	60.44
2	82015	13.1135	353,000	2.647	/5.1913	66.47	68.83
2	82022	13.3773	342,000	2,044	68.5794	67.75	69.21
2	82027	14 2425	331,090 373 000	2,044	60.1260 44 8600	61.3/ 52.00	66.21
2	82025	14.5005	375,000 335,000	2,000	טענסיויס מנור מנ	17 00 17 01	07.27 10 70
2	82027	15.0936	334.000	2.429	66.7996	63.47	60.51
2	82031	14.1919	338,000	2.534	70,4865	65.57	68.48
2	82033	14.0947	371,000	2.825	62.1064	57.33	63.09
2	82057	13.9065	409.000	3.212	52.7029	50.53	56.04
2	82065	15,2040	408.000	2.612	44.7819	58.35	63,52
2	B2074	16.7394	393,009	2,830	62.5567	59.95	65.91
2	82079	15.9974	257.000	2.151	17.1225	76.18	72.97
2	83007	15.0119	303,000	2.448	57.6808	67.36	69.3B
2	8300B	14.9631	366.000	2,754	52.6040	53.78	59.39
2	62012 87014	17.7681	274,000 364 000	Z.053	15.9203	17.82	74.40
≤	0.5014	1010111	24.1 1000	11101	00,0001	/4,10	13.92

.

50.12
1.1.1.1

0	e No	因 別					
Pop N	Silag	<u>In vi</u> v	MADF	LIGA	NCOMD	PCOMD	TVOVI
2	83021	17.3774	334,000	2.363	65,7921	67.68	69.81
2	83026	17.0312	301.000	2.229	68.2861	68.94	70.64
2	83027	18.9344	407.000	2.385	53.5362	61.20	60.08
2	83031	10.6224	397.000	3.430	50.6876	44.33	50.70
2	83049	16.8739	313.000	2 724	71.04/8	/1.28 47 A3	/1.58
2	B3065	12.8325	343.000	2.835	63.0456	61.70	63.38
2	83110	15.9381	298.000	2.222	75.8631	75.04	71.83
2	84018	14.1509	377.000	3.045	54.3314	56.51	62.55
2	84074	11.6966	333.000	2.654	69.0694	65.48	66.10
2	85017	17.8308	230 000	2,418	68,5526 90 5739	67.02 96.79	65.76 דם לד
2	85018	13.3615	322,000	2,538	55.6531	57.81	55.62
2	85022	12.5564	348,000	3.021	57.7654	51.17	57.02
2	B5024	13.5575	312.000	2.364	69.7624	64.64	65.94
2	85026	16,5319	285,000	1.897	74.5797	76.66	75.38
2	85077	11.85/8	367,000 375 000	Z 862	58.8235	55.41	63.82
2	85036	16,7603	381.000	2,551	52,5599	00.03 61,44	63-05
2	85037	15.9694	359.000	2,582	52.0548	56.40	57.34
2	85038	16.6381	321,000	2.171	66.2528	69.86	71.74
2	85041	16.6503	207.000	2.261	74.3790	74.53	72.75
2	85043 85040	13,2948	317,000	2,283	65.7016	63.54	57.30
2	85057	14.0000	291.000	2.480	07.0V41 75 8744	83.24 73.33	75 40
2	B6005	14,7491	332,000	2.435	73,8673	66.94	69.92
2	86015	17.6207	360,000	2.710	73,4141	55.40	60.24
2	86016	19.2503	335.000	1.833	79.6474	73.94	70.92
2	86018	15.9797	374.000	2.458	71.1850	64.29	67.26
2	86020	17.0638	314.000	2,330	73.9977	69.19	72.24
2	86026	16.2298	338,000	2,4:2	75.1845	67.40 71.88	69.13 73 75
2	86029	16.3439	333.000	2,379	64.1975	60.86	67.67
2	86033	15.0564	383.000	2.640	64.0412	59.75	63.56
2	B6036	20.6300	332.000	2.965	78.6869	71.66	71.02
2	86037	15.6059	354.000	2,435	68.7220	65.18	67.06
2	86067	14,3017	278,000 378 000	1.700	83.0214 70 0005	75.19	12.58
3	1	13.6948	453.000	3,520	53.7707	45.49	56.60
3	2	14.5237	423.000	3,140	66.8303	54.70	65,10
Ż	3	13,1103	434.000	3.360	55.8668	45.45	60.00
3	4	15.2058	373.000	2.850	66.7322	59,83	64.10
3	5 7	15,6981	353,000	2.740	75.2949	64.77	72.30
3 3	/ A	10.0442 15.9908	328 000	2.080	/3.6981 77 A05=	68./3 66.00	70.50
3	9	13,8951	424.000	3.420	53.9727	45,56	55.00
3	10	14.1690	364.000	3.010	69.7237	60.02	67.30
3	11	15.9789	406.000	3.020	65.1504	53,53	64.90
3	12	15.0308	386.000	2,840	66.4041	57.70	69.00
3 7	13	15,1736	410.000	3,160	62.2961	50.56	60.50
3	15	15.4536	424.000	3,100	62 3755	23.00 49 12	02.89 59.00
3	16	15.5412	400.000	3,150	70,0059	57.30	67.50
3	18	15.3636	390.000	3.100	67.6350	60.02	68.50
3	19	13.6010	385,000	3.450	63.6916	46.63	64.70
3	20	15,3276	345.000	3,010	74.7372	65.85	73.70
১ ব	21	15.2295	349.000	2,950	71.4358	66.06	73.10
3	23	16.0908	357.000	2.760 7.840	79.4038	73.20 49.00	78.20
3	25	16.4475	379.000	2,895	69.3263	65.16	71.00
3	26	14.1493	414,000	3.240	65.6650	54,82	67.10
3	27	16.8300	381.000	3.050	69.5411	59.93	71.60
3	28	15.9624	309.000	2.530	79.7580	71.12	77.40
ა	27	1010490	422 , 000	2.142	62.2/60	55.11	61.40

•

A N H X H Z	PCOM	UMOVI
3 30 15.9848 314.000 2.595 75.7597	72.31	75.20
3 32 16.5393 336.000 2.875 76.6559	73.39	75.80
3 33 15.9537 327.000 3.225 73.2692	69.13	72.20
3 37 16.5137 382.000 2.860 69.4098	65.78	72.60
3 38 15.7178 358.000 3.185 71.9930	67.96	74.50
3 39 15.1317 367.000 2.960 64.1123	60.53	65.50
3 40 16.6439 355.000 3.000 68.6999	67.14	73.60
3 41 16.5462 414.000 3.160 64.1020	60.25	67.90
3 42 14.2392 344.000 2.875 71.9492	67.20	69.30
3 44 15.2672 351.000 2.765 68.5165	64.17	68.40
3 45 14.2956 382.000 3.185 62.8661	52.30	67.70
3 46 17.4698 368.000 2.915 67.6315	63.48	70.70
3 47 16.3028 370.000 3.060 68.4013	58.57	71.70
3 48 13.6416 410.000 3.060 60.4970	57.13	67.20
3 50 15,4287 381,000 3,165 61,9968	57,56	68.40
3 51 15.4762 389.000 2.965 62.4572	52.24	62.90

For population numbers see Table 2.1

.

Analysis of Variance of the Regression Equations for the 3 Populations Used in the Calibration Set for the Conventional Predictors of in vivo DE

Models		MADF	LIGA	NCOMD	PCOMD	CIMOVI
Model	DF	138	138	135	138	138
1	MS	2.729	2.327	2.545	2.051	2.078
Model	DF	136	136	133	136	136
2	MS	2.438	1.686	2.308	1.919	1.996
Model	DF	134	134	131	134	134
3	MS	2.453	1.696	2.294	1.873	1.949

•

Analysis of Variance of Regression Equations for the 3 Populations Used in the Prediction of in vivo DE by NIR

۰.

9-term	130 0.8328	128 0.8407	110 0.8783
8-term	131 0.8608	129 0.8661	113 0.8738
 7-term	132 0.8445	130 0.9013	116 0.9243
6-term	133 1.026	131 1.013	119 1.049
5-term	134 1.104	132 1.114	122 1.169
4-term	135 1.240	133 1.241	125 1.214
3-term	136 1.379	134 1.358	128 1.397
2-term	1 <i>37</i> 1.539	135 1.440	131 1.469
1-term	138 1.852	136 1.606	134 1.616
	DF MS	R N N	FI SM SM
Models	Mbdel 1	Model 2	Model 3
APPENDIX 26

NIR Regression Equations Produced by Stepwise Multiple Linear Regression for the Prediction of <u>in vivo</u> DE of 140 Silages.

Term 1 B(0) = 11.219 B(1) = -487.390 87.00 2278. 1 16.4.4. Term 2 B(0) = 10.179 B(1) = -1017.376 29.06 1198. 1 16.4.4. B(2) = -630.472 133.69 2278. 1 16.4.4. B(2) = -630.472 133.69 2278. 1 16.4.4. E(1) = -534.728 16.94 2154. 2 24.4.4. B(2) = -1129.987 39.11 1198. 1 16.4.4. B(3) = -444.373 41.95 2278. 1 16.4.4. B(3) = -444.373 41.95 2278. 1 16.4.4. B(1) = -415.127 33.54 2278. 1 16.4.4. B(2) = 611.623 35.39 2434. 2 24.4.4. B(4) = -506.960 14.24 2154. 2 24.4.4. B(4) = -1665.40 30.78 2290. 2 24.4.4. B(2) = 559.237 27.17 2434. 2 24.4.4. B(3) = -1477.377 55.23 1246. 1 16.4.4. B(4) = -700.845 38.07 2154. 2 24.4.4. B(5) = -1477.377 55.23 1246. 1 16.4.4. B(4) = -700.845 38.07 2154. 2 24.4.4. B(5) = -700.273 23.83 1218. 2 24.4.4. B(2) = -632.249 40.50 2410. 2 24.4.4. B(3) = -229.447 57.71 2290. 2 24.4.4.4. B(4) = -700.845 38.07 2154. 2 24.4.4.4. B(5) = -700.273 23.83 1218. 2 24.4.4.4. B(4) = -700.845 38.07 2154. 2 24.4.4.4. B(5) = -700.273 23.83 1218. 2 24.4.4.4. B(5) = -700.273 23.83 1218. 2 24.4.4.4. B(6) = -681.980 27.37 2138. 2 24.4.4.4.	<u>(</u>	Coeff	licient	F	Wavelength	<u>Math Treatment</u>	
$\begin{array}{llllllllllllllllllllllllllllllllllll$	Term	1					
B(1) = -487.390 87.00 2278. 1 16. 4. 4. Term 2 B(0) = 10.179 B(1) = -1017.376 29.06 1198. 1 16. 4. 4. B(2) = -630.472 133.69 2278. 1 16. 4. 4. Term 3 B(0) = 9.724 B(1) = -534.728 16.94 2154. 2 24. 4. 4. B(2) = -1129.987 39.11 1198. 1 16. 4. 4. B(3) = -444.373 41.95 2278. 1 16. 4. 4. B(1) = -415.127 33.54 2278. 1 16. 4. 4. B(2) = 611.623 35.39 2434. 2 24. 4. 4. B(2) = 611.623 35.39 2434. 2 24. 4. 4. B(3) = -1196.941 39.77 1246. 1 16. 4. 4. B(4) = -506.960 14.24 2154. 2 24. 4. 4. B(2) = 559.237 27.17 2434. 2 24. 4. 4. B(2) = 559.237 27.17 2434. 2 24. 4. 4. B(4) = -106.540 30.78 2290. 2 24. 4. 4. B(5) = -1477.377 55.23 1246. 1 16. 4. 4. B(4) = -700.845 38.07 2154. 2 24. 4. 4. B(4) = -700.845 38.07 2154. 2 24. 4. 4. B(5) = -700.273 23.83 1218. 2 24. 4. 4. B(5) = -700.273 23.83 1218. 2 24. 4. 4. B(6) = -628.247 57.71 2290. 2 24. 4. 4. B(3) = -1242.039 40.50 2410. 2 24. 4. 4. B(4) = 979.402 61.67 2434. 2 24. 4. 4. B(5) = -1242.039 41.52 1246. 1 16. 4. 4. B(5) = -1242.039 41.52 1246. 1 16. 4. 4. B(5) = -1242.039 41.52 1246. 1 16. 4. 4. B(4) = 979.402 61.67 2434. 2 24. 4. 4. B(5) = -1242.039 41.52 1246. 1 16. 4. 4. B(5) = -1242.039 41.52 1246. 1 16. 4. 4. B(6) = -681.980 27.37 2138. 2 24. 4. 4.	B(0)	=	11.219				
Term 2 B(0) = 10.179 B(1) = -1017.376 29.06 1198. 1 16. 4. 4. B(2) = -630.472 133.69 2278. 1 16. 4. 4. Term 3 B(0) = 9.724 B(1) = -534.728 16.94 2154. 2 24. 4. 4. B(2) = -1129.987 39.11 1198. 1 16. 4. 4. B(3) = -444.373 41.95 2278. 1 16. 4. 4. B(1) = -415.127 33.54 2278. 1 16. 4. 4. B(2) = 611.623 35.39 2434. 2 24. 4. 4. B(2) = 611.623 35.39 2434. 2 24. 4. 4. B(4) = -506.960 14.24 2154. 2 24. 4. 4. B(2) = 559.237 27.17 2434. 2 24. 4. 4. B(4) = -166.540 30.78 2290. 2 24. 4. 4. B(4) = -700.845 38.07 2154. 2 24. 4. 4. B(4) = -700.845 38.07 2154. 2 24. 4. 4. B(5) = -700.273 23.83 1218. 2 24. 4. 4. B(5) = -700.273 23.83 1218. 2 24. 4. 4. B(2) = -63.249 40.50 2410. 2 24. 4. 4. B(3) = -124.039 41.52 1246. 1 16. 4. 4. B(2) = -124.249 40.50 2410. 2 24. 4. 4. B(3) = -124.039 41.52 1246. 1 16. 4. 4. B(3) = -124.039 41.52 1246. 1 16. 4. 4. B(4) = -700.273 23.83 1218. 2 24. 4. 4. B(5) = -623.249 40.50 2410. 2 24. 4. 4. B(5) = -122.039 41.52 1246. 1 16. 4. 4. B(5) = -124.039 41.52 1246. 1 16. 4. 4. B(5) = -124.039 41.52 1246. 1 16. 4. 4. B(5) = -124.039 41.52 1246. 1 16. 4. 4. B(6) = -681.980 27.37 2138. 2 24. 4. 4.	B(1)	=	-487.390	87.00	2278.	1	16.4.4.
$\begin{array}{llllllllllllllllllllllllllllllllllll$	Term	2					
B(1) = -1017.376 29.06 1198. 1 16. 4. 4. B(2) = -630.472 133.69 2278. 1 16. 4. 4. Term 3 $B(0) = 9.724 B(1) = -534.728 16.94 2154. 2 24. 4. 4. B(2) = -1129.987 39.11 1198. 1 16. 4. 4. B(3) = -444.373 41.95 2278. 1 16. 4. 4. B(3) = -444.373 41.95 2278. 1 16. 4. 4. B(1) = -415.127 33.54 2278. 1 16. 4. 4. B(2) = 611.623 35.39 2434. 2 24. 4. 4. B(3) = -1196.941 39.77 1246. 1 16. 4. 4. B(4) = -506.960 14.24 2154. 2 24. 4. 4. B(2) = 559.237 27.17 2434. 2 24. 4. 4. B(2) = 559.237 27.17 2434. 2 24. 4. 4. B(3) = -1477.377 55.23 1246. 1 16. 4. 4. B(4) = -700.845 38.07 2154. 2 24. 4. 4. B(5) = -700.273 23.83 1218. 2 24. 4. 4. B(3) = -229.447 57.71 2290. 2 24. 4. 4. B(3) = -229.447 57.71 2290. 2 24. 4. 4. B(3) = -229.447 57.71 2290. 2 24. 4. 4. B(3) = -1242.039 41.52 1246. 1 16. 4. 4. B(4) = 979.402 61.67 2434. 2 24. 4. 4. B(5) = -1242.039 41.52 1246. 1 16. 4. 4. B(6) = -681.980 27.37 2138. 2 24. 4. 4. B(6) = -681.980 27.37 2138. 2 24. 4. 4. B(5) = -1242.039 41.52 1246. 1 16. 4. 4. B(6) = -681.980 27.37 2138. 2 24. 4. 4. B(5) = -1242.039 41.52 1246. 1 16. 4. 4. B(6) = -681.980 27.37 2138. 2 24. 4. 4. B(5) = -1242.039 41.52 1246. 1 16. 4. 4. B(6) = -681.980 27.37 2138. 2 24. 4. 4. 5. B(6) = -681.980 27.37 2138. 2 24. 4. 4. 5. B(6) = -681.980 27.37 2138. 2 24. 4. 4. 5. B(6) = -681.980 27.37 2138. 2 24. 4. 4. 5. B(6) = -681.980 27.37 2138. 2 24. 4. 4. 5.$	B(0)	-	10.179				
$B(2) = -630.472 133.69 \qquad 2278. \qquad 1 16. 4. 4.$ Term 3 $B(0) = 9.724 \\ B(1) = -534.728 16.94 \qquad 2154. \qquad 2 24. 4. 4. \\ B(2) = -1129.987 39.11 \qquad 1198. \qquad 1 16. 4. 4. \\ B(3) = -444.373 41.95 \qquad 2278. \qquad 1 16. 4. 4. \\ B(3) = -444.373 41.95 \qquad 2278. \qquad 1 16. 4. 4. \\ B(1) = -415.127 33.54 \qquad 2278. \qquad 1 16. 4. 4. \\ B(2) = 611.623 35.39 \qquad 2434. \qquad 2 24. 4. 4. \\ B(3) = -1196.941 39.77 \qquad 1246. \qquad 1 16. 4. 4. \\ B(4) = -506.960 14.24 \qquad 2154. \qquad 2 24. 4. 4. \\ B(2) = 559.237 27.17 \qquad 2434. \qquad 2 24. 4. 4. \\ B(2) = 559.237 27.17 2434. \qquad 2 24. 4. 4. \\ B(3) = -1477.377 55.23 \qquad 1246. \qquad 1 16. 4. 4. \\ B(4) = -700.845 38.07 2154. \qquad 2 24. 4. 4. \\ B(5) = -700.273 23.83 \qquad 1218. \qquad 2 24. 4. 4. \\ B(2) = -623.249 40.50 2410. \qquad 2 24. 4. 4. \\ B(3) = -229.447 57.71 2290. \qquad 2 24. 4. 4. \\ B(3) = -229.447 57.71 2290. \qquad 2 24. 4. 4. \\ B(3) = -229.447 57.71 2290. \qquad 2 24. 4. 4. \\ B(4) = 979.402 61.67 2434. 2 24. 4. 4. \\ B(5) = -1242.039 41.52 1246. \qquad 1 16. 4. 4. \\ B(6) = -681.980 27.37 2138. \qquad 2 24. 4. 4. \\ B(6) = -681.980 27.37 2138. \qquad 2 24. 4. 4. \\ B(5) = -681.980 27.37 2138. \qquad 2 24. 4. 4. \\ B(5) = -681.980 27.37 2138. \qquad 2 24. 4. 4. \\ B(5) = -681.980 27.37 2138. \qquad 2 24. 4. 4. \\ B(5) = -681.980 27.37 2138. \qquad 2 24. 4. 4. \\ B(5) = -681.980 27.37 2138. \qquad 2 24. 4. 4. \\ B(5) = -681.980 27.37 2138. \qquad 2 24. 4. 4. \\ B(5) = -681.980 27.37 2138. \qquad 2 24. 4. 4. \\ B(5) = -681.980 27.37 2138. \qquad 2 24. 4. 4. \\ B(5) = -681.980 27.37 2138. \qquad 2 24. 4. 4. \\ B(5) = -681.980 27.37 2138. \qquad 2 24. 4. 4. \\ B(5) = -681.980 27.37 2138. \qquad 2 24. 4. 4. \\ B(5) = -681.980 27.37 2138. \qquad 2 24. 4. 4. \\ B(5) = -681.980 27.37 2138. \qquad 2 24. 4. 4. \\ B(5) = -681.980 27.37 2138. \qquad 2 24. 4. 4. \\ B(5) = -681.980 27.37 2138. \qquad 2 24. 4. 4. \\ B(5) = -681.980 27.37$	B(1)	=	-1017.376	29.06	1198.	1	16.4.4.
Term 3 B(0) = 9.724 $B(1) = -534.728 16.94 2154. 2 24.4.4.$ $B(2) = -1129.987 39.11 1198. 1 16.4.4.$ $B(3) = -444.373 41.95 2278. 1 16.4.4.$ $B(0) = 6.524$ $B(1) = -415.127 33.54 2278. 1 16.4.4.$ $B(2) = 611.623 35.39 2434. 2 24.4.4.$ $B(3) = -1196.941 39.77 1246. 1 16.4.4.$ $B(4) = -506.960 14.24 2154. 2 24.4.4.$ $B(4) = -506.960 14.24 2154. 2 24.4.4.$ $B(2) = 559.237 27.17 2434. 2 24.4.4.$ $B(2) = 559.237 27.17 2434. 2 24.4.4.$ $B(3) = -1477.377 55.23 1246. 1 16.4.4.$ $B(4) = -700.845 38.07 2154. 2 24.4.4.$ $B(5) = -700.273 23.83 1218. 2 24.4.4.$ $B(5) = -700.273 23.83 1218. 2 24.4.4.$ $B(2) = 623.249 40.50 2410. 2 24.4.4.$ $B(3) = -1242.039 41.52 1246. 1 16.4.4.$ $B(4) = 979.402 61.67 2434. 2 24.4.4.$	B(2)	=	-630.472	133.69	2278.	1	16.4.4.
B(0) = 9.724 $B(1) = -534.728 16.94 2154. 2 24.4.4.$ $B(2) = -1129.987 39.11 1198. 1 16.4.4.$ $B(3) = -444.373 41.95 2278. 1 16.4.4.$ $B(1) = -415.127 33.54 2278. 1 16.4.4.$ $B(2) = 611.623 35.39 2434. 2 24.4.4.$ $B(3) = -1196.941 39.77 1246. 1 16.4.4.$ $B(4) = -506.960 14.24 2154. 2 24.4.4.$ $B(4) = -506.960 14.24 2154. 2 24.4.4.$ $B(2) = 559.237 27.17 2434. 2 24.4.4.$ $B(3) = -1477.377 55.23 1246. 1 16.4.4.$ $B(4) = -700.845 38.07 2154. 2 24.4.4.$ $B(5) = -700.273 23.83 1218. 2 24.4.4.$ $B(5) = -623.249 40.50 2410. 2 24.4.4.$ $B(2) = -623.249 40.50 2410. 2 24.4.4.$ $B(3) = -1247.377 57.71 2290. 2 24.4.4.$ $B(3) = -1247.377 57.71 2290. 2 24.4.4.$ $B(4) = 979.402 61.67 2434. 2 24.4.4.$ $B(5) = -1242.039 41.52 1246. 1 16.4.4.$	Term	3					
$\begin{array}{rrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrr$	B(0)	=	9.724				
$\begin{array}{rrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrr$	B(1)	=	-534.728	16.94	2154.	2	24.4.4.
$B(3) = -444.373 41.95 2278. 1 16. 4. 4.$ $Term 4$ $B(0) = 6.524 \\ B(1) = -415.127 33.54 2278. 1 16. 4. 4.$ $B(2) = 611.623 35.39 2434. 2 24. 4. 4.$ $B(3) = -1196.941 39.77 1246. 1 16. 4. 4.$ $B(4) = -506.960 14.24 2154. 2 24. 4. 4.$ $Term 5$ $B(0) = 7.388 \\ B(1) = -166.540 30.78 2290. 2 24. 4. 4.$ $B(2) = 559.237 27.17 2434. 2 24. 4. 4.$ $B(3) = -1477.377 55.23 1246. 1 16. 4. 4.$ $B(4) = -700.845 38.07 2154. 2 24. 4. 4.$ $B(5) = -700.273 23.83 1218. 2 24. 4. 4.$ $B(2) = 623.249 40.50 2410. 2 24. 4. 4.$ $B(2) = -229.447 57.71 2290. 2 24. 4. 4.$ $B(4) = 979.402 61.67 2434. 2 24. 4. 4.$ $B(5) = -1242.039 41.52 1246. 1 16. 4. 4.$ $B(6) = -681.980 27.37 2138. 2 24. 4. 4.$	B(2)	=	-1129.987	39.11	1198.	1	16.4.4.
Term 4 $B(0) = 6.524$ $B(1) = -415.127 33.54 2278. \qquad 1 16. 4. 4.$ $B(2) = 611.623 35.39 2434. 2 24. 4. 4.$ $B(3) = -1196.941 39.77 1246. \qquad 1 16. 4. 4.$ $B(4) = -506.960 14.24 2154. \qquad 2 24. 4. 4.$ $Term 5$ $B(0) = 7.388$ $B(1) = -166.540 30.78 2290. 2 24. 4. 4.$ $B(2) = 559.237 27.17 2434. 2 24. 4. 4.$ $B(3) = -1477.377 55.23 1246. \qquad 1 16. 4. 4.$ $B(4) = -700.845 38.07 2154. 2 24. 4. 4.$ $B(5) = -700.273 23.83 1218. 2 24. 4. 4.$ $B(2) = -623.249 40.50 2410. 2 24. 4. 4.$ $B(3) = -229.447 57.71 2290. 2 24. 4. 4.$ $B(4) = 979.402 61.67 2434. 2 24. 4. 4.$ $B(5) = -1242.039 41.52 1246. 1 16. 4. 4.$ $B(6) = -681.980 27.37 2138. 2 24. 4. 4.$	B(3)	Ξ	-444.373	41.95	2278.	1	16.4.4.
$B(0) = 6.524$ $B(1) = -415.127 33.54 2278. \qquad 1 16. 4. 4.$ $B(2) = 611.623 35.39 2434. 2 24. 4. 4.$ $B(3) = -1196.941 39.77 1246. \qquad 1 16. 4. 4.$ $B(4) = -506.960 14.24 2154. 2 24. 4. 4.$ $Term 5$ $B(0) = 7.388 2290. 2 24. 4. 4.$ $B(2) = 559.237 27.17 2434. 2 24. 4. 4.$ $B(3) = -1477.377 55.23 1246. 1 16. 4. 4.$ $B(4) = -700.845 38.07 2154. 2 24. 4. 4.$ $B(5) = -700.273 23.83 1218. 2 24. 4. 4.$ $Term 6$ $B(0) = 8.237 -700.273 23.83 1218. 2 24. 4. 4.$ $B(3) = -229.447 57.71 2290. 2 24. 4. 4.$ $B(3) = -229.447 57.71 2290. 2 24. 4. 4.$ $B(4) = 979.402 61.67 2434. 2 24. 4. 4.$ $B(5) = -1242.039 41.52 1246. 1 16. 4. 4.$ $B(6) = -681.980 27.37 2138. 2 24. 4. 4.$	Term	4					
$B(1) = -415.127 33.54 2278. \qquad 1 16. 4. 4. \\ B(2) = 611.623 35.39 2434. \qquad 2 24. 4. 4. \\ B(3) = -1196.941 39.77 1246. \qquad 1 16. 4. 4. \\ B(4) = -506.960 14.24 2154. \qquad 2 24. 4. 4. \\ \hline \textbf{Term 5} \\ B(0) = 7.388 \\ B(1) = -166.540 30.78 2290. \qquad 2 24. 4. 4. \\ B(2) = 559.237 27.17 2434. 2 24. 4. 4. \\ B(3) = -1477.377 55.23 1246. \qquad 1 16. 4. 4. \\ B(4) = -700.845 38.07 2154. 2 24. 4. 4. \\ B(5) = -700.273 23.83 1218. 2 24. 4. 4. \\ B(2) = -623.249 40.50 2410. 2 24. 4. 4. \\ B(3) = -229.447 57.71 2290. 2 24. 4. 4. \\ B(4) = 979.402 61.67 2434. 2 24. 4. 4. \\ B(4) = -71242.039 41.52 1246. 1 16. 4. 4. \\ B(6) = -681.980 27.37 2138. 2 24. 4. 4. \\ B(6) = -681.980 27.37 2138. 2 24. 4. 4. \\ B(5) = -681.980 $	B(0)	=	6.524				
B(2) = 611.623 35.39 2434. 2 24.4.4. B(3) = -1196.941 39.77 1246. 1 16.4.4. B(4) = -506.960 14.24 2154. 2 24.4.4. Term 5 $B(0) = 7.388 B(1) = -166.540 30.78 2290. 2 24.4.4. B(2) = 559.237 27.17 2434. 2 24.4.4. B(3) = -1477.377 55.23 1246. 1 16.4.4. B(4) = -700.845 38.07 2154. 2 24.4.4. B(5) = -700.273 23.83 1218. 2 24.4.4. Crem 6B(0) = 8.237 B(1) = -751.372 40.69 1486. 2 24.4.4. B(2) = -623.249 40.50 2410. 2 24.4.4. B(3) = -229.447 57.71 2290. 2 24.4.4. B(4) = 979.402 61.67 2434. 2 24.4.4. B(5) = -1242.039 41.52 1246. 1 16.4.4. B(6) = -681.980 27.37 2138. 2 24.4.4. $	B(1)	11	-415.127	33.54	2278.	1	16.4.4.
$B(3) = -1196.941 39.77 1246. \qquad 1 16. 4. 4. \\ B(4) = -506.960 14.24 2154. \qquad 2 24. 4. 4. \\ \hline Term 5$ $B(0) = 7.388 \\ B(1) = -166.540 30.78 2290. \qquad 2 24. 4. 4. \\ B(2) = 559.237 27.17 2434. \qquad 2 24. 4. 4. \\ B(3) = -1477.377 55.23 1246. \qquad 1 16. 4. 4. \\ B(4) = -700.845 38.07 2154. \qquad 2 24. 4. 4. \\ B(5) = -700.273 23.83 1218. \qquad 2 24. 4. 4. \\ \hline Term 6$ $B(0) = 8.237 \\ B(1) = -751.372 40.69 1486. 2 24. 4. 4. \\ B(2) = -623.249 40.50 2410. 2 24. 4. 4. \\ B(3) = -229.447 57.71 2290. 2 24. 4. 4. \\ B(4) = 979.402 61.67 2434. 2 24. 4. 4. \\ B(5) = -1242.039 41.52 1246. 1 16. 4. 4. \\ B(6) = -681.980 27.37 2138. 2 24. 4. 4. \\ \hline $	B(2)	1	611.623	35.39	2434.	2	24. 4. 4.
$B(4) = -506.960 14.24 \qquad 2154. \qquad 2 24.4.4.$ $Ferm 5$ $B(0) = 7.388$ $B(1) = -166.540 30.78 2290. \qquad 2 24.4.4.$ $B(2) = 559.237 27.17 2434. \qquad 2 24.4.4.$ $B(3) = -1477.377 55.23 1246. \qquad 1 16.4.4.$ $B(4) = -700.845 38.07 2154. \qquad 2 24.4.4.$ $B(5) = -700.273 23.83 1218. \qquad 2 24.4.4.$ $B(5) = -623.249 40.69 1486. \qquad 2 24.4.4.$ $B(3) = -229.447 57.71 2290. \qquad 2 24.4.4.$ $B(4) = 979.402 61.67 2434. \qquad 2 24.4.4.$ $B(5) = -1242.039 41.52 1246. \qquad 1 16.4.4.$	B(3)	=	-1196.941	39.77	1246.	1	16. 4. 4.
Term 5 B(0) = 7.388 $B(1) = -166.540 30.78 2290. 2 24. 4. 4.$ $B(2) = 559.237 27.17 2434. 2 24. 4. 4.$ $B(3) = -1477.377 55.23 1246. 1 16. 4. 4.$ $B(4) = -700.845 38.07 2154. 2 24. 4. 4.$ $B(5) = -700.273 23.83 1218. 2 24. 4. 4.$ $Term 6$ $B(0) = 8.237$ $B(1) = -751.372 40.69 1486. 2 24. 4. 4.$ $B(2) = -623.249 40.50 2410. 2 24. 4. 4.$ $B(3) = -229.447 57.71 2290. 2 24. 4. 4.$ $B(4) = 979.402 61.67 2434. 2 24. 4. 4.$ $B(5) = -1242.039 41.52 1246. 1 16. 4. 4.$ $B(6) = -681.980 27.37 2138. 2 24. 4. 4.$	B(4)	=	-506.960	14.24	2154.	2	24. 4. 4.
B(0) = 7.388 $B(1) = -166.540 30.78 2290. \qquad 2 24. 4. 4.$ $B(2) = 559.237 27.17 2434. \qquad 2 24. 4. 4.$ $B(3) = -1477.377 55.23 1246. \qquad 1 16. 4. 4.$ $B(4) = -700.845 38.07 2154. \qquad 2 24. 4. 4.$ $B(5) = -700.273 23.83 1218. \qquad 2 24. 4. 4.$ Term 6 B(0) = 8.237 B(1) = -751.372 40.69 1486. 2 24. 4. 4. B(2) = -623.249 40.50 2410. 2 24. 4. 4. B(3) = -229.447 57.71 2290. 2 24. 4. 4. B(4) = 979.402 61.67 2434. 2 24. 4. 4. B(5) = -1242.039 41.52 1246. 1 16. 4. 4. B(6) = -681.980 27.37 2138. 2 24. 4. 4.	Term	5					
$B(1) = -166.540 30.78 2290. \qquad 2 24. 4. 4. \\ B(2) = 559.237 27.17 2434. \qquad 2 24. 4. 4. \\ B(3) = -1477.377 55.23 1246. \qquad 1 16. 4. 4. \\ B(4) = -700.845 38.07 2154. \qquad 2 24. 4. 4. \\ B(5) = -700.273 23.83 1218. \qquad 2 24. 4. 4. \\ \hline \textbf{Term 6}$ $B(0) = 8.237 \\ B(1) = -751.372 40.69 1486. 2 24. 4. 4. \\ B(2) = -623.249 40.50 2410. 2 24. 4. 4. \\ B(3) = -229.447 57.71 2290. 2 24. 4. 4. \\ B(4) = 979.402 61.67 2434. 2 24. 4. 4. \\ B(5) = -1242.039 41.52 1246. 1 16. 4. 4. \\ B(6) = -681.980 27.37 2138. 2 24. 4. 4. \\ \end{cases}$	B(0)	=	7.388				
B(2) = 559.237 27.17 2434. 2 24.4.4. B(3) = -1477.377 55.23 1246. 1 16.4.4. B(4) = -700.845 38.07 2154. 2 24.4.4. B(5) = -700.273 23.83 1218. 2 24.4.4. E(5) = -700.273 23.83 1218. 2 24.4.4. E(5) = -751.372 40.69 1486. 2 24.4.4. E(5) = -623.249 40.50 2410. 2 24.4.4. E(5) = -623.249 40.50 2410. 2 24.4.4. E(5) = -229.447 57.71 2290. 2 24.4.4. E(5) = -1242.039 41.52 1246. 1 16.4.4. E(6) = -681.980 27.37 2138. 2 24.4.4. E(5) = -681.980 27.37 2138. 2 24.4.4.4. E(5) 2 24.4.4.4.180 2 24.4.4.4.180 2 24.4.4.4.180 2 24.4.4.4.180 2 24.4.4.4.180 2 24.4.4.4.180 2 24.4.4.4.180 2 24.4.4.4.180 2 24.4	B(1)	=	-166.540	30.78	2290.	2	24.4.4.
$\begin{array}{rrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrr$	B(2)	=	559.237	27.17	2434.	2	24.4.4.
$\begin{array}{rcl} B(4) &=& -700.845 & 38.07 & 2154. & 2 & 24. & 4. & 4. \\ B(5) &=& -700.273 & 23.83 & 1218. & 2 & 24. & 4. & 4. \\ \hline \mbox{Term 6} \\ \hline B(0) &=& 8.237 \\ B(1) &=& -751.372 & 40.69 & 1486. & 2 & 24. & 4. & 4. \\ B(2) &=& -623.249 & 40.50 & 2410. & 2 & 24. & 4. & 4. \\ B(3) &=& -229.447 & 57.71 & 2290. & 2 & 24. & 4. & 4. \\ B(4) &=& 979.402 & 61.67 & 2434. & 2 & 24. & 4. & 4. \\ B(5) &=& -1242.039 & 41.52 & 1246. & 1 & 16. & 4. & 4. \\ B(6) &=& -681.980 & 27.37 & 2138. & 2 & 24. & 4. & 4. \end{array}$	B(3)	=	-1477.377	55.23	1246.	1	16.4.4.
B(5) = -700.273 23.83 1218. 2 24.4.4. $B(0) = 8.237$ $B(1) = -751.372 40.69 1486. 2 24.4.4.$ $B(2) = -623.249 40.50 2410. 2 24.4.4.$ $B(3) = -229.447 57.71 2290. 2 24.4.4.$ $B(4) = 979.402 61.67 2434. 2 24.4.4.$ $B(5) = -1242.039 41.52 1246. 1 16.4.4.$ $B(6) = -681.980 27.37 2138. 2 24.4.4.$	B(4)	==	-700.845	38.07	2154.	2	24. 4. 4.
Term 6 B(0) = 8.237 B(1) = -751.372 40.69 1486. 2 24.4.4. B(2) = -623.249 40.50 2410. 2 24.4.4. B(3) = -229.447 57.71 2290. 2 24.4.4. B(4) = 979.402 61.67 2434. 2 24.4.4. B(5) = -1242.039 41.52 1246. 1 16.4.4. B(6) = -681.980 27.37 2138. 2 24.4.4.	B(5)	14	-700.273	23.83	1218.	2	24. 4. 4.
$\begin{array}{rcl} B(0) &=& 8.237\\ B(1) &=& -751.372 & 40.69 & 1486. & 2 & 24. & 4. \\ B(2) &=& -623.249 & 40.50 & 2410. & 2 & 24. & 4. \\ B(3) &=& -229.447 & 57.71 & 2290. & 2 & 24. & 4. \\ B(4) &=& 979.402 & 61.67 & 2434. & 2 & 24. & 4. \\ B(5) &=& -1242.039 & 41.52 & 1246. & 1 & 16. & 4. \\ B(6) &=& -681.980 & 27.37 & 2138. & 2 & 24. & 4. \\ \end{array}$	Term	6					
B(1) = -751.372 40.69 $1486.$ 2 $24.4.4.$ $B(2) =$ -623.249 40.50 $2410.$ 2 $24.4.4.$ $B(3) =$ -229.447 57.71 $2290.$ 2 $24.4.4.$ $B(4) =$ 979.402 61.67 $2434.$ 2 $24.4.4.$ $B(5) =$ -1242.039 41.52 $1246.$ 1 $16.4.4.$ $B(6) =$ -681.980 27.37 $2138.$ 2 $24.4.4.$	B(0)	=	8.237				
B(2) = -623.249 40.50 $2410.$ 2 $24.4.4.$ $B(3) = -229.447$ 57.71 $2290.$ 2 $24.4.4.$ $B(4) = 979.402$ 61.67 $2434.$ 2 $24.4.4.$ $B(5) = -1242.039$ 41.52 $1246.$ 1 $16.4.4.$ $B(6) = -681.980$ 27.37 $2138.$ 2 $24.4.4.$	B(1)	=	-751.372	40.69	1486.	2	24.4.4.
B(3) = -229.44757.712290.224.4.4. $B(4) = 979.402$ 61.672434.224.4.4. $B(5) = -1242.039$ 41.521246.116.4.4. $B(6) = -681.980$ 27.372138.224.4.4.	B(2)	=	-623.249	40.50	2410.	2	24.4.4.
B(4) =979.40261.672434.224.4.4. $B(5) =$ -1242.03941.521246.116.4.4. $B(6) =$ -681.98027.372138.224.4.4.	B(3)	=	-229.447	57.71	2290.	2	24.4.4.
$B(5) \approx -1242.039$ 41.52 $1246.$ 1 $16.4.4.$ $B(6) \approx -681.980$ 27.37 $2138.$ 2 $24.4.4.$	B(4)	=	979.402	61.67	2434.	2	24.4.4.
B(6) = -681.980 27.37 2138. 2 24.4.4.	B(5)	=	-1242.039	41.52	1246.	1	16. 4. 4.
	B(6)	=	-681.980	27.37	2138.	2	24.4.4.

-

.

APPENDIX 26 (cont)

	Coef	ficient	<u>F</u>	<u>Wavelength</u>	<u>Math T</u>	reatment				
Ter	m 7									
B(0) =	8.654								
B(1) =	-568.823	33.09	2198.	2 24	. 4. 4.				
B(2) =	-1433.322	226.52	2138.	2 24	. 4. 4.				
B(3) =	-1404.757	79.92	1210.	1 16	. 4. 4.				
B(4) =	-1064.857	54.17	1710.	1 16	. 4. 4.				
B(5) =	-596.431	46.41	2406.	2 2.4	. 4. 4.				
B(6) =	-725.474	15.58	1754.	2 2.4	. 4. 4.				
B(7) =	763.223	4 9. 39	2434.	2 24	. 4. 4.				
Ter	m 8									
B(0) =	8.357								
B(1) =	-549.836	8.06	1754.	2 24	. 4. 4.				
B(2) =	733.533	46.82	2434.	2 24	. 4. 4.				
B(3) =	-820.179	6.17	1246.	1 16	. 4. 4.				
B(4) =	-560.672	33.37	2198.	2 24	. 4. 4.				
B(5) =	-1298.805	144.70	2138.	2 24	. 4. 4.				
B(6) =	-938.330	14.92	1210.	1 16	. 4. 4.				
B(7) =	-1226.279	61.72	1710.	1 16	. 4. 4.				
B(8) =	-505.501	29.32	2406.	2 24	. 4. 4.				
Term 9										
B(0) =	9.883								
B(1) =	-899.840	7.53	1250.	1 16	. 4. 4.				
B(2) =	-690,570	38.33	2198.	2 24	. 4. 4.				
B(3) =	-1401.194	151.99	2138.	2 24	. 4. 4.				
B(4) =	-1079.015	24.02	1210.	1 16	. 4. 4.				
B(5) =	-1678.402	47.10	1710.	1 16	. 4. 4.				
B(6) =	-623.062	37,02	2402.	2 24	. 4. 4.				
B(7) =	104.815	7.05	2306.	2 24	. 4. 4.				
B(8) =	-836.475	18.26	1754.	2 24	. 4. 4.				
B(9) =	692.997	49.44	2434.	2 24	. 4. 4.				

د د اد اد به يو د د اد اد اد به يو

ì ì



