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THE EFFECT OF ACUTE AMOXICILLIN EXPOSURE ON ANAEROBIC MICROBIAL COMMUNITIES - ACTIVITY, ECOLOGY AND RESISTOME

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SUBMITTED IN FULFILMENT OF THE REQUIREMENTS FOR THE DEGREE OF $Doctor \ of \ Philosophy$

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> UNIVERSITY OF GLASGOW JULY 2022

ABSTRACT

Antimicrobial resistant (AMR) genes in water treatment technologies are an issue of global concern. However, the impact of antibiotics on the anaerobic microbial communities within the decentralised wastewater treatment systems such as septic tanks is not well researched. This thesis aims to quantify the effects of an acute shock exposure of amoxicillin on the methanogenic activity, treatment efficacy, and structure of the microbial community in anaerobic granules. The work was conducted using an adapted specific methanogenic assay (SMA) to measure the rate of methane production with additional analytical methods including soluble chemical oxygen demand (sCOD) quantification, next generation sequencing (NGS), and AMR detection using Smart-Chip Real time PCR (quantitative PCR, qPCR) array. This work found that the presence of amoxicillin within the anaerobic microbial communities had measurable effects on the rate of methane production, community assembly, and the concentration of AMR genes within the effluent. However, the influence of amoxicillin on sCOD removal was limited. An additional study was conducted to determine the abiotic decay of amoxicillin within anaerobic culture medium. This study found that the oxygen scavenger L-cysteine has a strong catalytic effect on the decay of amoxicillin. As such, studies assessing the sensitivity of amoxicillin in lab grown cultures should take into account the abiotic decay within the medium used. Overall, the research demonstrates that the impact of pharmaceuticals, such as amoxicillin within decentralised waste treatment systems like septic tanks should be more widely investigated. Although septic tanks are rarely monitored, this work demonstrates that they have the potential to be point sources for the spread of AMR. Furthermore, changes to the microbial community caused by shock exposure to amoxicillin and other pharmaceuticals could lead to septic tank failure and discharge of untreated sewage into the water environment.

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LIST OF ABBREVIATIONS

ABC	ATP-binding cassette
AD	Anaerobic digestion
AMG	Antimicrobial genes
AMR	Antimicrobial resistance
ANOVA	Analysis of variance
ARA	Average relative abundance
ARG	Antimicrobial resistance gene
ASV	Amplicon sequence variant
ATP	Adenosine triphosphate
CH_4	Methane
CO_2	Carbon dioxide
COD	Carbon Oxygen Demand
СТ	Threshold cycle
DNA	Deoxyribonucleic acid
g	Gram
GC-FID	Gas chromatograph flame ionising detector
GMA	General methanogenic assay
gVSS	Grams volatile solids
H_2	Hydrogen gas
H_2S	Hydrogen disulphide
HCl	Hydrogen chloride
HPLC	High-performance liquid chromatography
КОН	Potassium hydroxide
L	
	Litre
LC-UV	Litre Liquid chromatography ultraviolet
LC-UV M	Litre Liquid chromatography ultraviolet Moles/litre
LC-UV M MATE	Litre Liquid chromatography ultraviolet Moles/litre Multidrug and toxic compound extrusion
LC-UV M MATE MBL	Litre Liquid chromatography ultraviolet Moles/litre Multidrug and toxic compound extrusion Metallo beta-lactamase
LC-UV M MATE MBL MDR	Litre Liquid chromatography ultraviolet Moles/litre Multidrug and toxic compound extrusion Metallo beta-lactamase Multi drug resistant
LC-UV M MATE MBL MDR MFS	Litre Liquid chromatography ultraviolet Moles/litre Multidrug and toxic compound extrusion Metallo beta-lactamase Multi drug resistant Major facilitator superfamily
LC-UV M MATE MBL MDR MFS MGE	Litre Liquid chromatography ultraviolet Moles/litre Multidrug and toxic compound extrusion Metallo beta-lactamase Multi drug resistant Major facilitator superfamily Mobile genetic elements
LC-UV M MATE MBL MDR MFS MGE ml	Litre Liquid chromatography ultraviolet Moles/litre Multidrug and toxic compound extrusion Metallo beta-lactamase Multi drug resistant Major facilitator superfamily Mobile genetic elements Millilitre
LC-UV M MATE MBL MDR MFS MGE ml	Litre Liquid chromatography ultraviolet Moles/litre Multidrug and toxic compound extrusion Metallo beta-lactamase Multi drug resistant Major facilitator superfamily Mobile genetic elements Millilitre N-acetyl glucosamine
LC-UV M MATE MBL MDR MFS MGE ml NAG NAM	Litre Liquid chromatography ultraviolet Moles/litre Multidrug and toxic compound extrusion Metallo beta-lactamase Multi drug resistant Major facilitator superfamily Mobile genetic elements Millilitre N-acetyl glucosamine N-acetyl muramic acid
LC-UV M MATE MBL MDR MFS MGE MI NAG	Litre Liquid chromatography ultraviolet Moles/litre Multidrug and toxic compound extrusion Metallo beta-lactamase Multi drug resistant Major facilitator superfamily Mobile genetic elements Millilitre N-acetyl glucosamine N-acetyl muramic acid Next generation sequencing

PBP	Penicillin binding protein
PCR	Polymerase chain reaction
psi	Pound per square inch
RND	Resistance-nodulation-division protein
rRNA	Ribosomal ribonucleic acid
sCOD	Soluble Chemical oxygen demand
SMA	Specific methanogenic assay
SMR	Small multidrug resistance
tCOD	Total chemical oxygen detection
UNICEF	United nations children's fund
UV Detector	Ultraviolet detector
VFA	Volatile fatty acid
WHO	World health organisation
WWT	Wastewater treatment
WWTP	Wastewater treatment plant

1

² Chapter 1 ³ Introduction

4

5

6 Antimicrobial resistant genes in the water environment is an issue of global pressing concern (Torres 7 et al., 2021). Whilst water infrastructure is recognised as key to limiting the spread of AMR, it is 8 increasingly recognised that wastewater treatment facilities can act as point sources for the 9 proliferation and spread of resistant microbes (Kümmerer, 2009a; Nguyen et al., 2021). However, much of the world is reliant on decentralised water treatment systems such as the septic tanks (Zhou 10 et al., 2014; Harada and Strande, 2016). Unlike centralised water treatment, biological treatment 11 12 within decentralised systems often utilises anaerobic microbial communities to treat wastewater (Diaz-Valbuena et al., 2011). Therefore, to understand the influence of antibiotics on decentralised 13 water treatment and septic systems, it is important to understand the effect that antibiotics have on 14 15 anaerobic cultures.

The beta-lactam antibiotic amoxicillin is one of the world's most widely prescribed antibiotic used 16 17 to treat a wide variety of conditions including pneumonia, dental abscesses, and stomach ulcers (Public Health England, 2017; Klein et al., 2019). However, whilst there has been some research into 18 19 other antibiotics on anaerobic microbial communities (Massé et al., 2000; Cetecioglu et al., 2015; 20 Aydin, 2016), there is very little research into the effects of amoxicillin. Additionally, whilst residual antibiotics entering a centralised wastewater treatment plant (WWTP) may be diluted by all the other 21 influent streams within the catchment, the narrow catchment into a septic tank means that any 22 contamination would enter in a shock load higher than would be seen in a WWTP. 23

This thesis aims to explore the effects of an acute shock exposure on amoxicillin on the methanogenic activity, treatment efficacy, and microbial structure of anaerobic granules. The research was conducted in three phases which explored (1) (Chapter 6) inhibition of methanogenic activity due to acute amoxicillin exposure (2) (Chapter 7) the development of antimicrobial resistance and changes in ecology due during acute exposure to amoxicillin; and (3) (Chapter 8) abiotic amoxicillin decay in anaerobic culture medium.

30

31 Thesis structure:

32 Chapter 3 – Methods and Materials

This chapter contains methods used within the experimental chapters within this thesis as well as method development for LC-UV and DNA extraction methods. As anaerobic culturing was used on multiple experiments within this dissertation, to minimise repetition, culture conditions are summarised in this chapter.

37 Chapter 4 - Method Optimisation and Baseline Study - Specific Methanogenic Activity Assay

A specific methanogenic assay (SMA) was conducted with the aim of establishing baseline activity in terms of maximum rate of methane production, window of maximum activity, and time required to exhaust the substrates provided. Additionally, the work aimed to optimise sample preparation methods. The findings of this chapter were used to inform experimental design subsequently.

42 Chapter 5 - Method Development for Repeat Batch Feeding of Anaerobic Granules

This study used a repeat batch assay to assess whether anaerobic granules fed on glucose and a complex broth could maintain activity over a two-week period without addition of trace nutrients. An SMA was conducted at the beginning, middle and end of the study to measure changes in activity with exposure to different substrate types. The work aimed to establish the health of the community through measuring variation in methane volume over time, the rate of methane production, and changes in the biomass in terms of grams of volatile suspended solids (gVSS) through time.

17

49 Chapter 6 – Changes in Methanogenic Activity in Response to Acute Amoxicillin Exposure

50 This chapter aimed to measure the inhibitory influence of amoxicillin on anaerobic systems experiencing a shock exposure event. Centralised wastewater systems are exposed to chronic low 51 52 concentration exposure to amoxicillin and other pharmaceuticals within the influent. However, the distribution of pharmaceuticals within small scale systems is likely to be of higher concentration and 53 irregularly timed. This study explored changes in methanogenic activity as a proxy for treatment 54 efficiency. As there was limited research into the inhibitory effects of amoxicillin in anaerobic 55 56 communities, it was initially necessary to determine how sensitive anaerobic communities are to 57 amoxicillin and at what exposure range inhibition of methanogenic activity was observed. A specific SMA was used to track how amoxicillin influenced the activity of each sub population within the 58 59 anaerobic trophic cascade. This was conducted by feeding a known concentration amoxicillin to 60 anaerobic granules along with co-substrates associated with different trophic levels: acetate, ethanol, 61 propionate, butyrate, glucose, and a proteinaceous complex broth and the gaseous substrate H_2/CO_2 . 62 As amoxicillin concentrations within influent is likely to vary highly depending on the influent type 63 and context, a range of nine amoxicillin concentrations were studied in two batches which ranged from 0.5 - 5.0 mg/L and 10 - 100 mg/L. The activity was calculated through measuring the 64 65 cumulative methanogenic output. The influence of amoxicillin was assessed through comparing the 66 rate of methane production, total methane output, as well as length of the lag time between substrate 67 addition and methane generation. Methane volume was measured using pressure transducers and 68 quantified using a GC-FID gas analyser.

69 Chapter 7 – Changes in Community Assembly and AMR Genes In Anaerobic Granules After

70 Acute Amoxicillin Exposure

Chapter 4 established that a shock exposure of amoxicillin produced a measurable inhibition of methanogenic activity within the anaerobic cultures used. This chapter aimed to explore the development of beta-lactam resistance genes as well as changes to the anaerobic community in response to shock exposure to amoxicillin. Anaerobic granules were exposed to four concentrations of amoxicillin: 1.0 mg/L, 10 mg/L, 50 mg/L and a blank along with glucose as a co-substrate.

76 Cumulative pressure was measured over a 48-hour window during which time sub-samples of each replicate test condition were removed for analysis. Effluent was collected for sCOD quantification. 77 78 Anaerobic sludge granules were flash frozen in liquid nitrogen for storage until DNA extraction. 79 Anaerobic activity was measured via methanogenesis and effluent sCOD. Changes within the 80 anaerobic community were measured using NGS profiling of 16S rRNA gene (V4 region) from whole community DNA by Earlham Institute (Norwich, England). The quantification of 81 82 antimicrobial resistant genes and 16s rRNA within the sludge was measured using the SmartChip[™] 83 Real-Time PCR system (TakaraBio, CA, USA) by Resistomap Oy (Helsinki, Finland).

84 Chapter 8 - The Influence of Anaerobic Culture Medium on The Structural Stability of

85 Amoxicillin

This chapter aimed to measure the decay rate of amoxicillin in the presence of different compounds 86 within anaerobic media in order to determine the impact of biotic versus abiotic decay within 87 anaerobic cultures. As the primary mechanism for beta-lactam resistance is the deactivation and 88 89 decay of the beta-lactam ring, it is assumed that the decay and removal of beta-lactams within cultures are primarily a biological action. This study aimed to assess the extent to which the 90 91 constituents of anaerobic culture media catalysed hydrolysis of the beta-lactam ring. The study 92 exposed amoxicillin to each of the individual components of anaerobic medium and incubated them 93 at 35°C as would be used during the culturing of anaerobic granules. Samples were taken at regular intervals at which point the amoxicillin was quantified using liquid chromatography with a UV 94 95 detector. This data was then used to determine the contextual half-life of amoxicillin for each test conditions. 96

97 Chapter 9 – Conclusions and Future Work

98 The contributions made within this thesis are summarised and discussed as they related to the 99 objectives of this work and wider understanding of beta-lactams within anaerobic communities. From 100 these conclusions, future work is discussed in term of potential experiment for further understanding 101 changes to the anaerobic community in response to beta-lactam exposure, confirmed beta-lactam 102 resistance within anaerobic systems, and improvements to study design.

² Chapter 2 ³ Literature Review

4 5

1

6 2.1 ANTIBIOTIC RESISTANCE IN WATER TREATMENT- AN ISSUE OF

7 GLOBAL CONCERN

Antibiotic resistance is an issue of increasing global concern (Torres *et al.*, 2021). Antibiotics are an
invaluable lifesaving treatment to reduce mortality associated with bacterial infection, prevent the
spread of disease, and make possible many surgical procedures regularly practiced today. As such,
the use of antibiotics has increased year after year (Center for Disease Dynamics Economics &
Policy, 2015). The increase in use of antibiotics however has been accompanied by an increase in
microbial resistance to those drugs, resulting in an overall reduction in their efficacy. Furthermore,
few new antibacterial drugs have been developed since the 1970s (Torjesen, 2013).

Estimations of the impact of antimicrobial resistance (AMR) vary. The Review on Antimicrobial 15 16 Resistance states that in the period running up to 2016 as many as 700,000 people died each year due to complications from antimicrobial resistance; and that this number could rise to 10 million by 2050 17 (O'Neill, 2016) (Figure 2-1). There are an estimated 2.8 million antibiotic resistance infections each 18 year in the United States which contribute to more than 35,000 deaths (Sriram et al., 2021). In the 19 20 European Union in 2009 over 25,000 deaths per year were attributed to AMR diseases (EMEA, 21 2009). A study by The Centre for Disease Dynamics, Economics & Policy (CDDEP) has suggested 22 that the burden of antimicrobial resistance is greater in middle and lower income countries, likely 23 due to variance in prescription policies as well as health and water infrastructure (Sriram et al., 2021).



Figure 2-1 - Predicted deaths attributed to AMR resistant diseases as of 2050. Figure taken from (O'Neill, 2016).

24 Whilst water infrastructure is recognised as key to limiting the spread of AMR, it is increasingly 25 recognised that wastewater treatment facilities can act as point sources for the proliferation and 26 spread of resistant microbes (Kümmerer, 2009a; Nguyen *et al.*, 2021). As such, the fate of antibiotics 27 in wastewater treatment and its spread is an area of active study (Singer et al., 2016; Jong et al., 28 2018; Hiller et al., 2019; Ebomah and Okoh, 2020; Nguyen et al., 2021). As the antibiotics that we 29 take leave our homes and hospitals in wastewater (Figure 2-2), they eventually enter our wastewater 30 treatment plants (WWTPs). Indeed, a wide variety of pharmaceutical compounds have been detected 31 at varying concentrations both in the influent and effluent of WWTPs (Matsuo et al., 2011; Blair et 32 al., 2013). Presently, the treatment processes at WWTPs are designed to breakdown and remove 33 digestible or settable organic materials and are not optimised for the removal of pharmaceuticals. As 34 such, these compounds may then eventually be discharged into the wider water environment. Wastewater treatment is at its heart a biological process. Exposure of microbial communities (as 35 36 might be found in WWTPs) to sublethal concentrations of antibiotics has the potential to select for antimicrobial resistance (Oberlé et al., 2012; Singer et al., 2016). Surveys of AMR within wastewater 37 treatment facilities regularly detect genes associated with resistance to antibiotics both within the 38

WWTP and in effluents (Amador *et al.*, 2015; Triggiano *et al.*, 2020; Wang *et al.*, 2020). These
effluents are usually discharged into local water ways and resistance genes are regularly detected
downstream (Singer *et al.*, 2016).

42 Whilst WWTPs are widespread, much of the world population do not have access to centralized 43 wastewater treatment and instead use decentralised treatment systems such as septic tanks and pit latrines (Zhou et al., 2018). And, while there is a significant body of research into the spread of AMR 44 in WWTPs, there is limited research into influence of antibiotics on the treatment efficiency or spread 45 46 of AMR in decentralised systems. As the technologies used in small scale and decentralised water 47 treatment systems are fundamentally different from large scale treatment facilities, understanding the 48 spread of AMR globally requires an increased understanding of AMR in decentralised wastewater 49 technologies.

50



Figure 2-2 - An example of how antibiotics move around our environment.

While humans use antibiotics medically, antibiotics are also applied to crops and livestock. This then runs off into the land and eventually into the water environment. The antibiotics we take ends up in our water treatment plants through sewage. As waste treatment effluent is usually discharged directly to the environment, any residual antibiotics within effluent will also enter the environment.

51 2.2 DECENTRALISED WASTEWATER TREATMENT

52 2.2.1 The global sanitation crisis

53 Access to effective sanitation is essential for ensuring global human health. Indeed, improving global 54 access to effective sanitation has been a priority for the World Health Organisation (WHO) for the last 30 years. Nevertheless, the latest report from the WHO on the state of the worlds sanitation found 55 56 that "over half of the world's population, 4.2 billion people, use sanitation services that leave human 57 waste untreated, threatening human and environmental health" (UNICEF & WHO, 2020) despite its 58 aim to end this practice globally by 2030 (Harada and Strande, 2016) (Figure 2-3). Despite an 59 increased reliance on centralised collection and treatment of wastewater within urban areas for the 60 last 100 years, it is currently estimated that 2.7 billion people worldwide rely on decentralised 61 sanitation technologies such as septic tanks and pit latrines (Harada and Strande, 2016). Due to their versatility and low cost, an increase in the utilisation of decentralised or onsite waste treatment 62 technologies could be a feasible path forward in the effort towards meeting the WHO sustainable 63 development goals (SDGs). There are three primary drivers towards the decentralisation of waste 64 65 treatment systems in address of the SDGs: economic, technical, and environmental (Massoud, Tarhini and Nasr, 2009). Decentralised systems are generally inexpensive to install and require less 66 maintenance than a large plant which by contrast requires large amounts of upfront funding as well 67 as ongoing maintenance and constant monitoring for continued operation (Sharma et al., 2013). 68 69 Additionally, the transport of waste to a central facility is expensive and requires the construction of pipes and pumps throughout the catchment area. These pipes are expensive to install and provide 70 opportunities for leaks and environmental contamination. Decentralisation can be a more cost-71 72 effective treatment option as cities expand into low density suburban landscapes and regions with 73 limited local budgets and restricted access to funding. Finally, decentralised wastewater treatment 74 facilities have significantly lower carbon footprints than centralised systems as they do not have the 75 electricity and fuels costs required for transfer of sewage to and treatment at WWTPs (Arias et al., 76 2020).



A. Distribution of global population not served with improved water supply in 2000

B. Global Sanitation Coverage 2017



Figure 2-3 - Distributions of Global Sanitation.

A Distribution of population not served with improved sanitation supply defined as "Use of pit latrines without a slab or platform, hanging latrines or bucket latrines" (WHO/Unicef 2000) B. Global sanitation coverage as of 2017. Each square represents a million people. (UNICEF 2020)

77 Decentralised waste treatment systems are currently used throughout the world. In industrialized 78 countries, decentralised waste treatment is commonly used in areas with low population density such 79 as rural or suburban settings. While there are many decentralised technologies available, due to its 80 simplicity and low maintenance, septic tanks are commonly utilised throughout the globe. For 81 example, roughly 20% of households in the USA are served by septic tanks (Donovan et al., 2013). It is estimated that there are 400,000 registered septic systems within England (DEFRA and EA, 82 2014). Although cities across Japan are primarily served by large centralised systems, in rural Japan, 83 reliance on septic tanks can be as high as 65% (Mizuochi et al., 2008). In developing regions or 84 85 countries, decentralisation is a cost-effective option for treatment in areas where centralised treatment systems are otherwise not available. In Thailand, off grid facilities are the primary treatment used 86 87 outside of cities (Withers et al., 2014).

88 2.2.2 Septic Tanks

89 Septic tanks are one of the oldest forms of modern sewage treatment. The first reported use of the 90 septic tank was by John Louis Mouras and Abbe Moigno in France in 1860 and consisted of a wooden 91 chamber that allowed for the separation of solids from effluent (Butler & Payne, 1995). This design was later patented by Donald Cameron in 1895. A similar design, the Imhoff cone, which allowed 92 93 for increased separation of the solids and effluent was patented in 1907. These technologies were 94 designed to treat relatively small volumes of waste before discharge directly to the vadose zone. 95 Although there have been many technological advances in centralised wastewater treatment systems 96 at an urban scale, decentralised small-scale systems are remarkably similar in design and function to 97 the initial technologies developed in 19th century (Lofrano & Brown, 2010). Modern septic tanks 98 are available in many different forms and designs; however, their treatment processes are all 99 fundamentally the same.

Septic tanks operate as settling tanks in which settleable solids are gravimetrically removed from
suspension where they are slowly anaerobically digested (Diaz-Valbuena *et al.*, 2011). Wastewater
enters the tank via a T-shaped inlet pipe or baffle which minimises mixing as influent enters the tank





Figure 2-4 - Diagram of a septic tank. Source www.owlshall.co.uk/

103 (Figure 2-4). As wastewater moves through the chamber fats, oils, or any other buoyant materials will collect on the surface where they are retained and form the scum layer. Settleable solids sink to 104 the bottom of the chamber where they form a layer of sludge, leaving behind a clarified effluent for 105 discharge. Anaerobic microbes slowly hydrolyse the organic matter from the influent. As the organic 106 107 matter is hydrolysed soluble metabolites, such as volatile fatty acids, dissolve into the clarified liquid 108 and are removed from the tank with the effluent. Generally, septic tanks only provide partial 109 degradation of the organic materials, and as a result requires additional treatment such as percolation 110 through the soils via underground pipes in a drain field (also called a soakaway) where the resident 111 soil microbiota will metabolise remaining organic materials. Nevertheless, methane and other 112 greenhouse gasses are emitted from septic tanks. The EPA estimates that septic system emissions 113 account for nearly 48.0 % of methane emissions from wastewater treatment, both centralised and 114 decentralised, in 2020 which make up 5.7 Tg of CO₂ equivalents (EPA, 2022).

115 As they have no moving parts, septic tanks do not require any additional power to operate, though they do require regular maintenance. As sludge is digested, indigestible or inorganic materials build-116 117 up in the tank and require removal, usually on an annual basis though frequency depends on the size 118 of the tank and the quality of the effluent. While septic tanks traditionally operate at ambient 119 temperatures, research into heated septic tanks has found that increasing the internal tank temperature 120 from 30°C to 40°C increased the methanogenic activity in the sludge layer of the septic tank. This 121 resulted in less total volatile solids formed, a decrease in the accumulation of settled sludge, and an 122 increase in the total methane produced (Pussayanavin et al., 2015).

The composition of influent into septic tanks varies greatly throughout the world with difference in application and usage. The main components of septic tank influents are comprised of black water and grey water. Black water is primarily comprised of excreta and flush water, while grey water is comprised of wastewater from washing, cleaning, and bathing. Brown water is comprised of a combination of the two (Metcalf and Eddy, 2003; Karia and Christian, 2013). The ratio of black water and greywater in the influent varies with region and application type, as well as throughout the day (Patterson, 2003). For example, in industrialised countries utilising septic tanks, households will

130 often have appliances generating significant greywater discharge such as showers, washing machines, and dishwaters. As a result, the influent into the septic tank will be generally less 131 132 concentrated with respect biochemical oxygen demand (BOD). By contrast, in areas with low or no 133 access to domestic water supply, where greywater is collected and reused, blackwater makes up a much greater proportion of the influent. As a result, the influent of these systems is generally much 134 135 more highly concentrated with respect to BOD. In some applications, such as vacuum toilets, flush 136 water is not used, which will concentrate the influent even further. As the biological degradation of 137 organic matter within a septic tank is primarily by anaerobic digestion, to understand how antibiotics 138 will affect decentralised septic systems, it is important to understand anaerobic digestion and 139 anaerobic communities.

140 2.3 ANAEROBIC DIGESTION

141 Anaerobic digestion is a naturally occurring biological process in which organic matter is broken 142 down into simpler compounds in the absence of oxygen and is driven by a diverse community of 143 microorganisms. Anaerobic microbial communities can be found in marine and freshwater lake 144 sediments (Hinrichs et al., 2000; Marchant et al., 2017; Martinez-cruz et al., 2018), fens and bogs 145 (Duddleston et al., 2002; Steinberg and Regan, 2008), within digestive systems (Zhai et al., 2020), 146 as well as within anthropogenic environments such as landfills (Tammemagi, 1999). Additionally 147 anaerobic digestion is used in industrial settings including for the removal of organics in wastewaters 148 (Karia and Christian, 2013) and food waste (Poggio et al., 2016), methane production for biofuels 149 (Kiselev et al., 2019), and the fermentation of cheese and alcohol (Ishtar Snoek and Yde Steensma, 150 2008; Button and Dutton, 2012).

151 2.3.1 Anaerobic Digestion Process

Anaerobic digestion is undertaken by a diverse set of microbes via a syntrophic food chain whereby the metabolites produced by some specialised members of the microbial community are the substrates required by other members within the community (Gerardi, 2003). Anaerobic microbial communities comprise a mix of bacteria and archaea. Although anaerobic digestion is only



Figure 2-5 - Schematic of four phases of biogas production. Figure adapted from: https://www.e-education.psu.edu/egee439/node/727



Figure 2-6 - Phylogenetic tree of bacteria, archaea, eucarya (Woese, 1996).

156 undertaken in the absence of oxygen, the microbes within anaerobic microbial communities have 157 varying levels of oxygen tolerance. Strict anaerobes cannot survive in the presence of oxygen. 158 Facultative anaerobes are capable of using oxygen as an electron acceptor when it is present and are capable of growing in both aerobic and anaerobic conditions (Madigan, 2014). The anaerobic 159 160 digestion process is conducted sequentially over four stages (1) hydrolysis, (2) acidogenic 161 fermentation, (3) acetogenesis and (4) methanogenesis, in which each stage is conducted by different 162 subpopulations with the anaerobic microbial community (Sanders, 2001; Gerardi, 2003) (Figure 2-163 5). While the first three stages are undertaken by bacteria, methanogenesis is undertaken exclusively 164 by archaea which sit in their own domain (Woese, 1996) (Figure 2-6).

The first stage of anaerobic digestion is hydrolysis in which complex insoluble compounds such as proteins, lipids, and complex carbohydrates are broken down into smaller soluble substances. As these large macromolecules cannot be transported across cell membranes, digestion occurs extracellularly (**Figure 2-7**). Exoenzymes are excreted from the cell wall which hydrolyse the bonds within the substrate and release the soluble monomers. In this way, proteins are broken down into amino acids, carbohydrates release soluble sugars, and lipids break into fatty acids. In the treatment of wastewater, as the wastewater influent usually contains high concentrations of complex solid

waste, the hydrolysis step is often rate limiting in that context (Bialek, Cysneiros and O'Flaherty,
2014). Hydrolysis is primarily undertaken by hydrolytic bacteria, facultative anaerobes, and strict
anaerobes which are each capable of generating enzymes that hydrolyse unique bonds within the
provided substrates. Complex substrates requiring hydrolysis often also contain essential materials
for the formation of biomass and other enzymes.

In the second phase of anaerobic digestion, acidogenic fermentation, the soluble products of 177 178 hydrolysis are brought inside the cell where they are further broken down to provide energy (ATP) and biological building materials (Madigan, 2014) (Figure 2-7). The products of fermentative 179 180 digestion usually include volatile fatty acids, alcohols, lactic acid, carbon dioxide, hydrogen, 181 ammonia and hydrogen sulphide, as well as new bacterial cells. A large diversity of facultative and 182 strict anaerobes produces an equally diverse variety of compounds in many fermentative processes. 183 However, many of the microbes responsible for this process are the same as those responsible for 184 hydrolysis (de Lemos Chernicharo, 2015). Acidogenesis is exergonic in which the reaction is the 185 accompanied by the release of energy and therefore enables the rapid growth of fermentative bacteria 186 (Seghezzo et al., 1998). As the products of acidogenesis are acidic, high levels of acidogenic activity 187 can lower the pH of the system and disrupt methanogenesis.



Figure 2-7 – Diagram of exoenzymes and endoenzymes.

Fermentation products are further oxidised by acetogenic bacteria into substrates appropriate for methanogenesis. The primary products generated by acetogenesis are acetic acid, hydrogen and carbon dioxide though the ratios of each depend on which products were produced during fermentation as well as the composition of the microbial community. If not utilised in methane formation this hydrogen can react with carbon dioxide and acetic acid to form other organic acids lowering the pH of the aqueous medium in the process (de Lemos Chernicharo, 2015).

The final stage of the anaerobic digestion process is methanogenesis in which methane and CO_2 are 194 195 formed. The production of methane is primarily conducted by methanogenic archaea, a 196 phylogenetically diverse group of strict anaerobes, with some methanogenesis conducted by bacterial 197 species under rare conditions such as in phosphate starved ocean surface conditions (Thauer et al., 2008; Carini et al., 2014). Methane can only be produced using a limited number of substrates: acetic 198 199 acid, hydrogen/CO₂, formic acid, methanol, methylamines and carbon monoxide (Gerardi, 2003). 200 When molecular sulphur is present, methanogens will preferentially produce large amounts of H_2S 201 in addition to methane (Garcia, Patel and Ollivier, 2000). There are currently more than 150 202 identified species of methanogens though more are being discovered regularly. In general 203 methanogens can be grouped according to substrate use (Jabłoński, Rodowicz and Łukaszewicz, 204 2015). (1) Hydrogenotrophic methanogens oxidise H_2 and reduce CO_2 to form methane. The 205 hydrogenotrophic group also includes formatrophs which are capable of oxidising formate to form 206 methane. (2) Methylotrophic methanogens utilise methane compounds such as methanol, 207 methylamines, and dimethylsulfide as their primary substrates. (3) Acetotrophic (aceticlastic) 208 methanogens utilise acetate as their primary substrate (Garcia, Patel and Ollivier, 2000). While the 209 specific makeup of species varies greatly from culture to culture, the three primary groups found in 210 biogas producing anaerobic digesters are Methanobacterium, Methanolinea (hydrogenotrophic) and 211 Methanosaeta (acetoclastic) (Gerardi, 2003; Trego et al., 2020).

212 **2.4 BETA LACTAMS**

213 Beta lactams are some of the most widely used antibiotics in the world (Klein et al., 2019) 214 constituting roughly 65% of the global market (Githinji et al., 2011). In 2017 approximately 44.6% 215 of all antibiotics used in the UK were penicillin class (Public Health England, 2017). Beta lactams 216 are defined by the beta lactam ring at the centre of their molecular structure. They were among the 217 first antibiotics developed after Scottish researcher Alexander Fleming first discovered penicillin in 218 1928 (Bennett and Chung, 2001). Since then, hundreds of different types of beta-lactam derivatives 219 have been developed each with a central lactam ring. There are four classes of beta lactam: 220 penicillin's, cephalosporins, monobactam and carbapenems with a primary core structure containing 221 either a single lactam ring or a double ringed structure (Figure 2-8). Although the first beta lactams 222 were isolated from natural origin, as in the case of penicillin, novel beta lactams are synthesized 223 through the addition of side chains onto the four core structures resulting in semisynthetic compounds 224 with varying properties and potencies (Kong, Schneper and Mathee, 2010).

225



Figure 2-8 - *The four primary structures of beta-lactam antibiotics. The central defining beta-lactam ring is highlighted in green.* (Wikimedia Commons, 2010)

The primarily mechanism of action for beta-lactam class antibiotics is through inhibition of bacterial cell wall synthesis. Beta-lactams interfere with cell division and the maintenance of structures within growing cells by inactivating penicillin binding proteins (PBP) therefore inhibiting the final transpeptidation step of peptidoglycan synthesis (Wong *et al.*, 2021).

230 Peptidoglycan is a polymer composed of alternating units of the amide derivative N-231 acetylglucosamine (NAG) and N-acetylmuramic acid (NAM). To connect the chains of NAG and NAM together and form the peptidoglycan layer, an enzyme called a transpeptidase removes the 232 ends of the amino acid chains and catalyses the crosslinking of a peptide bond between the amino 233 234 acid chains attached. One of the primary transpeptidase enzymes responsible for this action is PBP, 235 named for its affinity to binding with the beta lactam ring on penicillin class antibiotics (Madigan, 236 2014). As beta-lactam antibiotics deactivate the PBP the cell wall loses is structural integrity and 237 lyses (Wong *et al.*, 2021) (Figure 2-9). There are a variety of different types of PBPs within each 238 cell and can differ between bacterial species. As the structure of the PBP can vary between different 239 species so too does the effectiveness of the various beta-lactam drugs (Kong, Schneper and Mathee, 240 2010).



Figure 2-9 – Diagram of peptidoglycan and the action of penicillin around PBP. (Sawa, Kooguchi and Moriyama, 2020)



Figure 2-10 – The cell wall of gram-positive bacteria, gram negative bacteria, and two types of archaea. Beta-lactam antibiotics act through synthesis inhibition and destruction of the peptidoglycan (green bar). As archaea do not contain peptidoglycan, they are naturally immune to beta-lactam antibiotics. (Gill and Brinkman, 2011)

241 Beta-lactams affect both gram positive and gram-negative cells, although the effect is different for 242 each due to the differences in the structure of the cell walls (Figure 2-10). Gram positive cells are 243 comprised of a single phospholipid bilayer with a thick layer of peptidoglycan providing structure. 244 The cross linking in the peptidoglycan of gram-positive cells are held together through a peptide inter-bridge in which the amino acids within the peptidoglycan are crosslinked at several points. 245 246 Gram negative cells, by contrast, have two phospholipid bilayers with a thin layer of peptidoglycan in between. The crosslinking of the peptidoglycan in gram-negative cells is bonded with a single 247 248 peptide bond on the amino acid on the terminal D-alanine within the amino acid chain. The outer 249 layer of gram-negative cells is primarily comprised of phospholipid bilayer with a variety of 250 transmembrane proteins such as porins, channels, and pumps which facilitate the movement of 251 materials in and out of the cell. In between the inner phospholipid bilayer and the peptidoglycan of 252 the gram-negative cells is the periplasmic space (Madigan, 2014). This is comprised of a gel like 253 matrix known as periplasm which contains a store of enzymes, including those which confer 254 resistance for susceptible bacteria. Because the peptidoglycan on gram positive cells is more readily 255 exposed, it is more susceptible to the beta lactam antibiotics. Nevertheless, beta lactams are able to 256 enter into the periplasm of gram-negative cells where they are able to act upon the peptidoglycan

257 (Wong et al., 2021). The beta lactam ring within all penicillin class antibiotics can irreversibly bind to PBP. Binding inactivates the PBP protein by changing its shape and hence its ability to catalyse 258 259 the crosslinking between amino acids within peptidoglycan. This in turn prevents the generation and 260 repair of cell walls resulting in weak points. As weaknesses in the cell wall accumulate over time, pressure within the cell causes the cell to rupture, killing the cell in the process (Wong et al., 2021). 261 262 Organisms of the archaeal domain, including those responsible for methanogenesis, are structurally diverse and have a multitude of outer membrane structures which differ greatly to those observed in 263 264 bacteria. Unlike most bacteria, archaea do not contain murein, a type of peptidoglycan (Kandler and 265 König, 1998). In their place the structure of archaeal cell walls is obtained through a para-crystalline 266 structure commonly comprised of proteins or glycoproteins, though other structures have been found 267 as well. Because archaea do not contain peptidoglycan (Figure 2-10) they are not susceptible to beta-268 lactam antibiotics (Hans and König, 1996; Khelaifia and Drancourt, 2012).

269 2.5 BETA LACTAM RESISTANCE

As beta-lactam antibiotics have been used extensively, there has been proliferation of antibiotic resistance within bacterial populations. Three distinct mechanisms for resistance to the damage that beta lactam rings cause are identified: alteration of the PBP binding site, removal though an efflux pump, and degradation of the beta-lactam active site via beta-lactamase enzymes (Nikaido and Pagès, 2012; Bush, 2013; Zango *et al.*, 2019; Park, Sutherland and Rafii, 2020).

275 2.5.1 Penicillin Binding Proteins

The antibiotic properties of beta-lactams function by competitively bonding with the PBP activation site and preventing the formation of cell walls (Madigan, 2014). However, changes to the structure of PBP can prevent beta-lactam binding resulting in resistance to this class of antibiotics. For example, one of the most studied examples of this mutation is the PBP2a which is carried on the gene mecA which and resides on a large mobile genetic element and is believed to be transmitted through horizontal transfer between staphylococcus species (King *et al.*, 2017). While the *mecA* gene can be found in a wide variety of *Staphylococcus* species it is most widely associated with 283 Methicillin-resistant *Staphylococcus aureus* (MRSA) a pathogen of concern which has been 284 observed to be resistant to several widely used antibiotics including beta-lactams (Lakhundi and 285 Zhang, 2018).

286 2.5.2 Efflux Pumps

287 Efflux pumps form part of the xenobiotic removal system in which potential bacterial stressors are 288 removed from the cell through efflux transporters within the cell membrane (Madigan, 2014). Efflux 289 transporters exist as either single component pumps which transport the target from the cytosol into 290 the periplasm; or multicompetent pumps, which capture their targets from the inner membrane or 291 periplasm and transport them outside the cell. While single component pumps contain a single part, 292 multi component systems are typically comprised of a pump, an outer membrane channel protein 293 (OMP), and an accessory membrane fusion protein such as MFP (Figure 2-11). While the majority 294 of known efflux pumps reside in bacteria, they can also be found in eukaryotic as well as archaeal 295 organisms (Elbourne et al., 2017). Major families of efflux transporters include RND, MFS (major



Figure 2-11 – Diagram of single component efflux pumps and multi component efflux pumps. (Li, Plésiat and Nikaido, 2015)
296 facilitator superfamily), MATE (multidrug and toxic compound extrusion), SMR (small multidrug 297 resistance), and ABC (ATP-binding cassette) super- families or families (Poole, 2005). 298 Efflux pumps are nonspecific in what they remove and are therefore a part of multidrug resistance 299 (MDR) however, their role in beta-lactam resistance is highly complex (Lin and Scott, 2012; Li, Plésiat and Nikaido, 2015). The influence of efflux pumps on beta-lactam resistance is actively being 300 301 investigated and appears to depend on drug properties (eg. hydrophobicity and ionic strength), 302 bacterial species, and beta-lactamase production (Li, Plésiat and Nikaido, 2015). Efflux mediated 303 beta-lactam resistance has been observed via chromosomally encoded multicomponent pumps of the 304 RND family AcrAB-TolCj for H. influenzae and MexAB-OprMk for P. aeruginosa and several Gram-negative bacteria. Additionally, the ABC family LmrA pump has been observed to confer 305 306 modest beta-lactam resistance for L. lactis (Poole, 2005).

307 2.5.3 Beta-Lactamases

308 The primary mechanism for beta lactam resistance is mediated through the production of beta-lactam 309 enzymes which catalyse the hydrolysis of the amine bond within the beta-lactam ring inactivating 310 the beta-lactam ring in the process (Madigan, 2014). Both gram negative and gram-positive cells can 311 generate beta-lactamase, but as gram negative cells are capable of storing proteins in their 312 periplasmic space, they are able to react immediately when exposed to a beta-lactam antibiotic. As a 313 result, they can have a constitutive resistance as they are immediately able to respond to exposure. 314 Gram positive microbes, on the other hand, do not have this periplasmic space, and are not able to 315 store these proteins for later use. Therefore, gram positive microbes have inductive resistance and 316 only begin generating resistance proteins in response to exposure to the antibiotic (Madigan, 2014).

Beta lactamases are classified using two methods. The Ambler System classifies beta lactams based on amino acid structure, whilst the Bush-Jacoby classification scheme classifies beta lactams based on their function (Bush and George A. Jacoby, 2010) (**Table 2-1**). In the Ambler Classification system, beta-lactamases are grouped into four classes by structure. Classes A, C, and D enzymes are comprised of serine beta-lactamases (SBL) that hydrolyse the beta-lactam ring via a serine-bound acyl intermediate within their active site. Class B are comprised of metallo-beta-lactamase enzymes

323	(MBL) which utilise zinc to directly hydroxide ion within the beta-lactam ring (Ambler, 1980). The
324	Bush – Jacoby Method classifies beta-lactamases into three groups depending on their function and
325	enzymatic susceptibility to inhibitors with each of the three groups approximately corresponding
326	with specific structural classes. Group 1 comprises enzymes targeting cephalosporins and
327	corresponds to class C beta-lactamases. The genes encoding for beta-lactamases in this group were
328	originally chromosomal (Prescott, 2013). Group 2 comprises all other beta-lactamases with a serine
329	active centre and correspond to the A and D structural classification. Group 3 comprises MBLs as
330	correspond with class B (Bush and George A Jacoby, 2010). As both classification systems convey
331	different information, they are commonly used in conjunction to describe the beta-lactamase
332	functions and activity (Sawa, Kooguchi and Moriyama, 2020) (Table 2-1).

Ambler Molecular Class	Bush Jacoby group	Distinctive Substrate	Representative Enzyme	Enzyme Type
	2a	Penicillins	PC1	
	2b	Penicillins, early cephanosporins	TEM-1, TEM-2, SHV-1	
	2be	Extended - spectrum cephalosporins, monobactams	TEM-3, SHV-2, CTX-M- 15, PER-1, VEB-1	
	2br	Penicillin	TEM-30, SHV-10	Conina hata
А	2ber	Extended - spectrum cephalosporins, monobactams	TEM-50	lactamases
	2c	Carbenicillin	PSE-1, CARB-3	
	2ce	Carbenicillin, cefepime	RTG-4	
	2e	Extended - spectrum cephalosporins	СерА	
	2f	Carbapenems	KPC-2, IMI-1, SME-1	
B1	3a	Carbapenems	IMP-1, VIM-1,CcrA, IND-!	
B2	3a	Carbapenams	L1, CAU-1, GOB-1, FEZ- 1	Metallo beta- lactamases
B3	3b	Carbapenems	CphA, Sfh-1	
С	1	Cephalosporins	AmpC, ACT-1, CMY-2, FOX-1, MIR-1	Serine beta-
	1e	Cephalosporins	GC1, CMY-37	lactamases
	2d	Cloxacillin	OXA-1, OXA-10	
D	2de	Extended - spectrum cephalosporins	OXA-11, OXA-15	Serine beta- lactamases
	2df	Carbapenams	OXA-23, OXA-48	

 Table 2-1 - Beta-lactamase classification systems adapted from Ambler (1980) and Bush and Jacoby (2010).

333 Class A beta lactamases are the most diverse group and is comprised of penicillinases, expanded spectrum beta- lactamases (ESBL)s, and carbapenemase encoding for several phyla such as 334 335 Actinobacteria, Bacteroidetes, Cyanobacteria, Firmicutes and Proteobacteria (Walther-Rasmussen 336 and Høiby, 2007). Class A beta-lactamases were one of the first classes to be identified by Ambler 337 in 1980, later classified as 2b by Bush - Jacoby. It was observed that TEM and SHV genes encode for 338 enzymes capable of rapidly degrading penicillin and early carbapenems (Ambler, 1980). Since then, 339 mutations to TEM-3, TEM- 50 and SHV-2 (Bush Jacoby Class 2be) have allowed for additional 340 capabilities to degrade ESBLs along with Cep A (2e), CTX-M, PER-1, and VIB-1 (2be). Class A 341 carbapenemases can be categorised into six separate groups comprised of GES, KPC, SME, 342 *IMI/NMC-A* enzymes, while *SHV-38* and *SFC-1* each separately constitute a group. Class 2d and 2de 343 enzymes are capable of degrading carbenicillin and are generally represented by PSE, CARB, and 344 RTG genes (Naas et al., 2017). At the time of writing there are 1701 enzymes associated with Class 345 A beta-lactamases (Naas et al., 2017).

346 Class C Beta-lactamases are primarily active against cephalosporins. Shortly after the initial Amber 347 Classification scheme was proposed the cephalosporinase gene AmpC was sequenced and did not fit 348 into either of the classes previously established (Jaurin and Grundstrom, 1981). The Bush-Jacoby 349 systems breaks Ambler Class C into two subgroups. Group 1 contain exclusively cephalosporins and 350 is represented by the AmpC, CMY, ACT, FOX, and MIR genes. Group 2 degrades cefradine and oxy-351 beta lactams and are represented by GC1 and CMY gene families. Genes in this class are highly 352 conserved and primarily occur in gram-negative bacteria where they are encoded into the 353 chromosome (Zango et al., 2019). At the time of writing there are 3578 enzymes associated with 354 Class C beta-lactamases (Naas et al., 2017).

Class D antibiotics were established by Huovinen et al in 1988 who identified the *PSE-2* serine, later named *OVA-10*, differed sufficiently in its structure from other SBLs to be designated its own class. (Huovinen, Huovinen and Jacoby, 1988). This class is comprised of oxacillinases (OXA) and are capable of degrading isoxazolyl β -lactams like methicillin and oxacillin (Medeiros, Cohenford and Jacoby, 1985). There are three Bush-Jacoby functional groups within Ambler class D. Group 2d are 360 comprised of cloxacillin- and oxacillin- hydrolysing genes *OXA-1* and *OXA-10*. In group 2de the 361 degradation substrate is extended to extended-spectrum cephalosporins (oxyimino- β -lactams), but 362 not carbapenems and contain genes *OXA-11* and *OXA-15* and are frequently encoded on plasmids 363 (Antunes and Fisher, 2014). Subgroup 2df contains carbapenem-hydrolyzing class D β -lactamases 364 (CHDLs). At the time of writing there are 1102 enzymes associated with Class D beta-lactamases 365 (Naas *et al.*, 2017)

366 Ambler Class B enzymes are comprised of carbapenem hydrolysing metallo-beta-lactamases (MBL) 367 in which the hydrolytic centres of each contain at least one active site zinc atom (Ambler, 1980). 368 Both Ambler and Bush Jacoby classification schemes subdivide MBLs into three subclasses based 369 on their functionality, sensitivity to inhibitory metal ion chelators, and number of zinc ligands within 370 their structure (Bush and George A Jacoby, 2010). However after the structural dissimilarities within 371 the NDM-1 MBL within were discovered, it was suggested that a second B1 subclass be created to 372 accommodate these unique sequences (Yong et al., 2009). Class B1 and B3 are both capable of 373 degrading most beta-lactams, though Class B2 primarily targets carbapenems. The primary genes 374 families with Class B are IMP, VIM, SPM, GIM, NDM, and FIM. Their activity can be inhibited by 375 the addition of the chelating agent EDTA. At the time of writing there are 549 enzymes associated 376 with Class B1, 23 associated with B2, and 225 associated with subclass B3 beta-lactamases (Naas et 377 al., 2017).

378 2.5.4 Transfer of beta-lactam resistance

379 There are two methods for the transfer of resistance genes between bacteria: vertical gene transfer, 380 as happens when genes are passed from parent to daughter cells through replication, and horizontal 381 gene transfer when genes are transferred from one individual to another within the same generation. 382 The three main mechanisms for horizontal gene transfer are transformation, transduction and 383 conjugation. During transformation bacterium uptake RNA or DNA, often in the form of plasmids, 384 from within the surrounding environment. Transduction occurs when bacterial DNA is moved from 385 one bacterium to another via viruses or bacteriophages. Conjugation occurs when DNA is transferred 386 between bacteria directly via pili. Horizontal gene transfer plays a central role in the spread of antimicrobial resistance as plasmids carry genes from one host to another. In this way genes
conferring resistance can spread throughout a bacterial community, even between species (Madigan,
2014).

390 2.5.5 Beta-lactam Resistance in Wastewater and Anaerobic Microbial Communities

391 The effluent from WWTPs is considered to be one of the most significant conduits for AMR 392 proliferation and the spread of beta-lactam resistant microbes into the environment (Kümmerer, 393 2009a; Ebomah and Okoh, 2020). Several studies have investigated the occurrence of antimicrobial 394 resistance genes (ARGs) in wastewater and the efficacy of wastewater treatment in reducing AMR 395 between the influent and effluent (Rizzo et al., 2013; Biswal et al., 2014; Bengtsson-Palme et al., 396 2016). Although beta-lactams are one of the most widely prescribed and consumed antibiotics, they 397 are rarely detected within WWTP settings. As beta-lactams are highly susceptible to hydrolysis it is 398 generally believed that they are broken down within the sewers in transit to the treatment facilities 399 (Tran, Reinhard and Gin, 2018). Nevertheless, many studies reported beta-lactam resistance genes 400 in most if not all influent and effluent samples tested (Laht et al., 2014; Rodriguez-Mozaz et al., 401 2015; Neudorf et al., 2017; Lin et al., 2021). The dominant genes at each plant varied between studies 402 due to natural variances in treatment, ecology and design, as well as the scope of the project. However 403 the blaTEM, blaVIM, blaCTX, blaSHV and blaOXA are broadly considered dominant within the 404 wastewater biology (An et al., 2018; Hiller et al., 2019; Subirats et al., 2019). A meta-study of AMR 405 reduction across different wastewater treatment technologies found that although beta-lactamase 406 associated genes were often most concentrated within activated sludge, they were typically lower in 407 the effluent than the influent (Hiller et al., 2019).

Within WWTPs anaerobic digesters come in a wide variety of designs as well as both mesophilic (25~45 °C) and thermophilic (50–60 °C) temperatures. Generally membrane based technologies are the most efficient and removing ARGs from the effluent as they effectively separate the sludges from the effluent where they remains within the reactor (Hiller *et al.*, 2019; Nguyen *et al.*, 2021). Do et al., (2022) found that compared to storage or composting, anaerobic was more efficient at reducing the microbial load, ARGs and MGEs in pig slurry. Additionally, anaerobic technologies have been

found to limit the spread of antimicrobial resistance compared to composting or conventional
activated sludge (Munir, Wong and Xagoraraki, 2011; Tong *et al.*, 2019). Temperature also plays a
key role in the reduction of ARGs within anaerobic digestion. Thermophilic digestion has been
shown to effectively remove 50–99% of tetracycline ARGs and class 1 integrons in lab scale and full
scale digesters (Ghosh, Ramsden and Lapara, 2009; Diehl and Lapara, 2010).

Methanogens, which are comprised exclusively of archaeal species, do not contain peptidoglycan in 419 their cell walls and as a result are naturally resistant to beta lactam antibiotics (Hans and König, 420 421 1996; Khelaifia and Drancourt, 2012). Nevertheless, as anaerobic microbial communities are 422 comprised of a combination of bacteria and archaea, anaerobic communities as a whole are 423 susceptible to, and develop resistance to, beta-lactams. Because anaerobic microbial communities are highly syntrophic, instability within one sub population can lead to instability within the whole 424 425 system. As such, when exposed to high concentrations of beta-lactam antibiotics, anaerobic microbial 426 communities have been shown to decrease in methanogenic activity (Heidari, Fatemeh Nabavi, 427 Saffari Khouzani, Mehdi Amin, et al., 2012).

While there is little research into the proliferation of resistance within anaerobic communities in a 428 429 wastewater context, many known beta-lactam resistant pathogens are known to grow under anaerobic 430 conditions and have been detected in wastewater (Zagui et al., 2020). For example Aeromonas sp, 431 which are facultative anaerobes commonly found in WWTPs, has been shown to be a carrier for beta-lactam resistant genes for all classes of the Ambler scale (Piotrowska et al., 2017). Of the 12 432 antibiotic-resistant "priority pathogens" identified by the WHO as posing the greatest threat to human 433 health, seven are beta-lactam resistant strains and 11 are capable of growing under anaerobic 434 435 conditions (Table 2-2) (WHO, 2017). However, anaerobic treatment of sewage known to contain 436 beta-lactam resistant microbes has repeatedly been shown to reduce the relative abundance and total 437 counts of AMR within the effluent compared to influent (Resende et al., 2014; Zhang et al., 2016; 438 Kanger et al., 2020). Zhang, Yang and Pruden (2015) found that biomatter within effluent from 439 mesophilic anaerobic digestion contained less than 10% the AMR genes counts relative to the 440 influent. However research by Ju (2016) found that while most ARGs were removed during

441 anaerobic digestion, the species that remained in the effluent (mainly Collinsella aerofaciens,

- 442 Streptococcus salivarius and Gordonia bronchialis) were of pathogenic importance (Ju et al., 2016).
- **443** *Table 2-2* Antibiotic-Resistant "priority pathogens" adapted from the WHO guidelines from 2017. (WHO, 2017)

Species	Resistance	Oxygen Tolerance
Priority 1: CRITICAL		
Acinetobacter baumannii,	carbapenem-resistant	aerobic
Pseudomonas aeruginosa,	carbapenem-resistant	obligate aerobe (although can grow anaerobically in the presence of nitrate) (Health protection agency, 2015)
<i>Enterobacteriaceae</i> , (including Salmonella, Escherichia coli, Klebsiella, and Shigella)	carbapenem-resistant, ESBL-producing	facultative anaerobes,
Priority 2: HIGH		
Enterococcus faecium,	vancomycin-resistant	facultative anaerobes
Staphylococcus aureus	methicillin-resistant, vancomycin- intermediate and resistant	facultative anaerobes
Helicobacter pylori,	clarithromycin-resistant	facultative anaerobes
Campylobacter spp.,	fluoroquinolone-resistant	strict anaerobe
Salmonellae,	fluoroquinolone-resistant	facultative anaerobes
Neisseria gonorrhoeae	cephalosporin-resistant, fluoroquinolone-resistant	obligate aerobic (although can grow anaerobically in the presence of nitrate)
Priority 3: MEDIUM		
Streptococcus pneumoniae,	penicillin-non-susceptible	facultative anaerobes
Haemophilus influenzae,	ampicillin-resistant.,	facultative anaerobes
Shigella spp	fluoroquinolone-resistant	facultative anaerobes

444

445 2.5.6 Methods to Quantify AMR Resistance

Early detection of antimicrobial resistance was developed from diagnostic culture techniques developed for the assessment of drug suitability and disease identification (Madigan, 2014). Culture methods aimed to measure the minimum concentration of antibiotic needed to prevent the growth of the test culture. This was useful in a clinical setting in order to develop treatment, though was also able to detect the presence of resistant pathogens (Shanmugakani *et al.*, 2020). However, there were limitations to this method. Culture techniques were only suitable for microbes which were readily culturable in plate and broth media. Additionally, only one strain could be detected at a time.

455 With the advancement of gene sequencing and polymerase chain reaction (PCR) technologies it 456 became possible for the resistance genes themselves to be detected (Hiller et al., 2019). As such 457 environmental surveys began to be conducted on antibiotic restrained within wastewater and the 458 environment. This allowed for a rapid identification of ARGs directly. There were two limitations to this technology. The equipment required was advanced and expensive. As the method was only 459 460 capable of targeting a single gene target at a time, researchers needed to know ahead of time which 461 genes they were aiming to find. Additionally, while it could detect the genes present, qPCR was not 462 able to identify which species were resistant, only if resistance genes were present within the 463 community.

464 The development of the high-throughput qPCR array in the late 2010s allowed for the simultaneous 465 detection of hundreds of mobile genetic elements (MGEs) and ARGs (Looft et al., 2012). Parallel 466 analysis cut down on the time and the cost of running the tests (Waseem et al., 2019). Furthermore, 467 as thousands of primers have been developed, the large number of tests available made it possible to 468 screen large gene sets, which is particularly useful for environmental and water samples (Stedtfeld 469 et al., 2018). However, as these systems run in parallel, all samples were run under the same PCR 470 conditions, it was not possible to optimise conditions for the amplification of each primer. This resulted in variable sensitivity across the libraries (Waseem et al., 2019). However, the technologies 471 472 are still currently widely used with several commercial available tools available, of which the Takara 473 (previously Wafergen) SmartChip instrument which can amplify up to 5184 qPCR assays per chip 474 within 3–4 h allowing up to 384 primer sets can be analysed in parallel, is the most popular (Waseem et al., 2019). 475

476 **2.6 KNOWLEDGE GAPS**

To fully understand how beta-lactams impact decentralised water treatment, it is first important tounderstand how they impact the treatment efficiency of anaerobic systems. However, there is limited

479 research into the effects of beta-lactams on anaerobic systems. When it is, the conditions of these studies are varied and often focused on higher concentrations than would be found in water treatment 480 481 facilities and the results do not always agree. In general, it was found that the impact of beta lactams 482 on treatment efficiency are limited at low concentrations. For example, Sanz found a maximum inhibition of 10 mg/L penicillin led to a partial inhibition of 25-45% methane production (Sanz, 483 484 Rodríguez and Amils, 1996). Anaerobic digesters running on pig slurry from animals with 16 mg/kg 485 of penicillin in their feed reported a 35% reduction in methane generation when compared to the 486 digesters running wastes from pigs fed a control feed (Massé et al., 2000). Conversely, Zhang et al 487 found that anaerobic systems acclimated to amoxicillin could withstand concentrations up to 60mg/L 488 before displaying any inhibition of methane generation (Zhang et al., 2015). A study by Su et al 489 (2019) did not report any inhibition of methane in anaerobic granules exposed to 100ppm amoxicillin 490 (Su et al., 2019). As these studies vary in design and application, it is difficult to make broad 491 conclusions. The aim of this thesis is to further explore how amoxicillin antibiotics influence 492 anaerobic microbial communities specifically in terms of activity, treatment efficiency, and 493 community ecology and resistome.

² Chapter 3 ³ Materials and Methods

4

1

5 The aim of this chapter is twofold. Firstly, several of the studies within this thesis include the 6 culturing of anaerobic granules, and, whilst some of the details of each experiment were unique, 7 many of the methods remained consistent across all studies in which they appeared. Therefore, this 8 chapter summarises those methods that are repeated. Secondly, method development was required 9 for amoxicillin quantification using LC-UV as well as PCR. The optimisation and development of 10 these methods are laid out in this chapter.

11 **3.1** ANAEROBIC CULTURING

12 3.1.1 Source And Characteristics of Anaerobic Sludge

The sludge used in all trials was sampled on 6th July 2018 from a full-scale mesophilic Expanded 13 14 Granular Sludge Bed (EGSB) bioreactor operated by the North British Distillery Company in Edinburgh, Scotland. The granules had historically been treating distillery waste. The sludge was 15 16 transported back to the lab in two 30L jerry cans. The process engineers at the facility reported that the digesters had not been recently fed prior to sampling. The sludge was stored in the basement of 17 18 the Rankine building on the University of Glasgow campus at room temperature and irregularly fed glucose with nutrient additives (Section 3.2.1) until use. Although the graneuls had previously been 19 20 optimised to distillery waste as their primary substrate, with the alternative and intermittent feeding 21 it is likely that they instead optimised for both glucose fermentation and / or dead cellular matter 22 within the culture.

3.1.2 Tracking Methanogenic Activity for Specific Methanogenic Assay (SMA) and General Methanogenic Assay (GMA)

25 Methanogenic activity of the granular sludge was measured using an adapted version of the pressure 26 transducer technique in which the change in headspace pressure over time within sealed serum vials 27 containing sludge samples and a specific substrate (food) is used to estimate methane production 28 rates (Colleran et al., 1992; D. Coates, F. Coughlan and Colleran, 1996). The specific methanogenic 29 activity of each test culture was measured against simple substrates: acetate, ethanol, propionate, and 30 butyrate as described by (Colleran et al., 1992). Additionally, activity was measured using a general 31 methanogenic rate assay in which glucose and a complex broth were used as substrate (modified 32 from the OECD 224 (OECD, 2007)). During the experiment, a known mass of substrate was added to a sealed bottle containing an anaerobic culture medium and granules. The change in pressure in 33 the headspace of the vials was intermittently monitored over time as the substrate was converted to 34 35 methane and carbon dioxide using a pressure transducer. The methane concentration in the headspace gas was measured at the end of the experiment (Figure 3-1). The pressure changes and methane 36 concentration were then used to determine the methanogenic activity of the sludge against a specific 37 substrate. As each substrate used can only be broken down by a subset of the population, using an 38



Figure 3-1 - The specific methanogenic assay (SMA) tracks the changes in pressure over time. The pressure data and methane quantification data are used to estimate net methane production and to estimate the activity rate of the microbes within the culture when incubated with specific substrates.

array of different substrates indicates the activity of each sub population associated with each
substrate type. While the design specifics for each experiment differed slightly, the preparation of
the bottle, granule acclimation, pressure readings, and analytical methods were all similar between
experiments. These methods are described below.

Sample bottles were prepared by soaking overnight in Virkon[™] (Lanxess) solution followed by 43 triplicate rinsing with Milli-Q® Type 1 Ultrapure Water (Milli-Q) and air drying. Within an 44 anaerobic chamber, 0.5g of sludge granules were added to 30ml bottles (Sigma Aldrich, Catalogue 45 no. 33106) along with 9.5ml of anaerobic buffer to provide approximately 10 g/L VSS per assay. 46 47 The bottles were sealed using rubber bungs and aluminium crimp caps before removing from the chamber (Sigma Aldrich Catalogue No. 508500). Once removed, the headspace gas in each bottle 48 49 was exchanged to 1 atmosphere (ATM) with a 20% CO_2 and 80% N_2 mixture by flushing for 30 50 seconds. Each test condition was prepared in triplicate. Once filled, sealed, and flushed, the culture 51 bottles containing granules and culture medium (a mixture of buffers, oxygen scavengers, indicators, and metal nutrients) were acclimated at 37°C for 3 days prior to commencing the monitoring period. 52 53 To minimise the time to initial methane production, a method development study (Chapter 5) was 54 conducted. The study demonstrated that the addition of substrate during the acclimation period 55 reduced the length of the lag phase, reducing the experiment duration and resources for the research 56 team. Thus, each substrate was also added at the offset of the acclimation phase for each of the 57 experiments. The specific substrates used differed for each study. After the acclimation incubation, 58 each bottle was opened within the anaerobic chamber. The liquid medium within was removed using a pipette and discarded. A mix of fresh anaerobic medium, buffer and nutrients were added to the 59 60 acclimated granules and amoxicillin added as appropriate to the specific test. Bottles were resealed within the anaerobic chamber. The headspace gas was flushed for 30 seconds with a 20% CO_2 and 61 80% N₂ mixture to a final pressure of 1ATM to remove any methane or oxygen present. After the 62 63 head space gas was exchanged, the relevant liquid substrate was injected into each bottle through the septa and placed into a 37°C shaking incubator with the time of feeding representing the 64 65 commencement of the experiment. The set of liquid substrates and amoxicillin concentrations used 66 were different for each experiment conducted and details are be described in the relevant chapter.

During each experiment, pressure readings were taken every 1 -4 hours using a Centrepoint pressure transducer until the pressure within each bottle began to plateau at which point the frequency of readings was reduced. At the end of each experiment the biogas within the headspace of each bottle was removed using a syringe needle through the septa. The methane content of the biogas was determined using a GC-FID biogas analyser (described in detail in section 3.3.1). The volatile suspended solids (VSS) of the biomass were measured as described in Section 3.3.2. The methanogenic activity was calculated as described in Section 3.3.3.

74 **3.2 MATERIALS**

75 3.2.1 Anaerobic Medium Components and Preparation

76 The anaerobic buffer was comprised of resazurin (0.4ml/L) as an anaerobic indicator and 1-cysteine (0.56g/L) as an oxygen scavenger. The indicator and 1-cysteine were dissolved in 1 L of Milli-Q 77 water then pH adjusted using 8N NaOH until it reached 7.0 - 7.2 pH. This solution was then boiled 78 79 until colourless, indicating that there is no dissolved oxygen in the solution, while the headspace was 80 continually purged with nitrogen gas. Once the medium was completely colourless, the bottle was placed in an ice bath to cool. When the temperature of the solution reached 50°C, 3.05g of sodium 81 82 bicarbonate was added to increase the alkalinity of the medium. The buffer was then stored under 83 anaerobic conditions and used within 48 hours.

3.2.

3.2.2 Trace Metals and Additives

Some of the studies contained additional nutrient additives (modified from Shelton & Tiedje, 1984) 85 and buffer as detailed in Chapters 4, 5, 6, and 7 within the dissertation. The nutrient additives 86 included a "Nutrient Mix", "Trace nutrient mix", and phosphate buffer solution. The additive mixes 87 88 were prepared and stored separately to allow preparation at different stock concentration to make the 89 dilutions easier to prepare and more accurate, and to reduce the likelihood of interaction between the 90 chemical components in each solution. Bottles used to prepare stock solutions of each mix were soaked in a Virkon[™] solution, rinsed three time with Milli-Q water and fully dried before use. The 91 Nutrient Mix was prepared at 10x the final working concentration and contained 5.3 g/l of NH₄Cl; 92

93	0.75g/L of CaCl ₂ ; and 1g/L Mg Cl * 6H ₂ 0 in 100ml Milli-Q. The Trace nutrient mix was prepared
94	as a 500x mix of final concentration and was comprised of 20g/L of FeCl ₂ *4H ₂ ; 0.5 g/l MnCl ₂ *
95	4H ₂ O; 0.05g/l H ₃ BO ₃ ; 0.05g/l ZnCl ₂ ; 0.03 g/l CaCl ₂ ; 0.01 g/l NaMo ₄ *2H ₂ O; 0.5 g/l CaCl ₂ * 6H ₂ O;
96	0.05 g/l NiCl ₂ * $6H_20$; and 0.05 g/l SeO ₂ . The FeCl ₂ * $4H_2O$ was added to 10ml of 25% (7.7M) HCl
97	before dilution and mixture with the trace nutrient stock solution. A phosphate buffer was added to
98	the anaerobic medium to maintain 7.0pH within the bottles. The phosphate buffer was mixed at 10x
99	the final concentration and contained: 2.7g/L H ₂ KPO ₄ , and 3.5g/L of K ₂ HPO ₄ . Both nutrient mixes
100	and the buffer were autoclaved after preparation and stored in the refrigerator at 4°C for the length
101	of the experiment. The nutrient composition was adapted from Shelton & Tiedje, (1984).

102 **3.3** ANALYTICAL METHODS AND CALCULATIONS

103 3.3.1 Quantifying Methane on the GC-FID

104 To quantify the methane content of the biogas in the headspace of sample bottles, a 2.0 ml sample of 105 headspace gas was removed from each bottle using a standard 20ml syringe and a luer lock. As the 106 upper limit of detection for the equipment was 5.0% methane, each gas sample taken from the vial 107 was diluted in the syringe 1:10 with N_2 to 20ml total volume before analysis. Methane content in the 108 biogas was quantified using a gas chromatograph (7890A, Agilent, Palo Alto, CA) equipped with a 109 GS-CarbonPlot capillary column and a flame ionization detector. Instrument parameters were: inlet 110 temperature - 250 °C; detector temperature - 250 °C; H₂ flow, 30 mL/min; air flow - 400 mL/min, 111 nitrogen as a carrier gas at a constant pressure of 15 psi. The oven temperature was 150 °C for 10 112 min.

113 3.3.2 Calculating VSS

The volatile solids content of the biomass was measured in accordance with Standards Methods 2540 B and 2540 E (Association., Association. and Federation., 1998). Volatile suspended solids (VSS) are a measure of the volatile material within the sample. This material is assumed to be organic and is used as a proxy for biomass. To determine the VSS in each sample vial at the end of each test, the total contents of the culture bottle were filtered through 0.45um Whatman® membrane glass fibre filter papers using a vacuum pump and filter apparatus. The filters and sample were dried in a 100°C
oven for 48 hours then the combined mass of each filter and sample was determined by weighing.
The filters and samples were then transferred to a furnace and volatilised at 500°C for 4 hours to
remove any volatile solids and the combined mass of the filter and ash determined by weighing. The
mass of each filter was recorded before measurement and filtration to enable calculation of the dried
sample mass and ash mass.

125 The mass (g) of the volatile solids (VSS) was calculated:

VSS(g) = (Final Mass dried sample(g) - Mass Filter(g)) - (Mass Sample Ash(g) - Mass Filter(g))

126

127 3.3.3 Calibrating Pressure Transducer

Measuring changes in microbial activity using the specific methanogenic assay requires tracking 128 129 pressure changes in the headspace of culture bottles. This is achieved with the use of a pressure 130 transducer and is reported in millivolts (mv). Thus, to enable reporting of pressure in units ATM, a 131 calibration curve was obtained by injecting known volumes of gas into a container of a known 132 volume and measuring the corresponding pressures. This was done using 30ml serum vials, as were 133 used in the SMA test, with 10ml of water added to each bottles to simulate the 10 ml of medium and 134 granules present during the studies leaving 27ml of headspace. Five bottles were stoppered and 135 sealed. The stoppers were pierced with a needle and left to equilibrate for 5 seconds, after which an 136 initial reading was taken. After the initial readings were taken, 2ml of air was injected into each of 137 the bottle and a pressure reading was taken. This was repeated until a total of 30 ml was injected into each of the bottles with corresponding pressure readings taken for each step. 138

The gas volume added (mL) was then plotted against the pressure readings (mv) to produce the calibration curve (**Figure 3-2**). A best fit regression line was calculated for the collected data to produce a set of constants which could be used to calculate gas volume (ml) from experimentally measured pressure readings (mv) **Equation 2** (n=80, slope = 4.0797, intercept = 1.6624, R^2 = 0.9984).



Figure 3-2 – Calibration curve for gas volume.

The curve compared the known volume of gasses added to the bottle and the corresponding pressure readings they produced. A best fit line was produced using experimental

144

$$Volume (ml) = \frac{Pressure (mv) - 1.6624}{4.0797}$$
 Equation 2

145

146 **3.3.4** Calculating Methane Volume

147 The pressure within culture bottles was measured (as mV) using a pressure transducer. The recorded

148 values were converted into total biogas volume (ml) using the conversion calculation in **Equation 3**.

$$Total Biogas Volume (ml) = \frac{Pressure (mv) - 1.6624}{4.0797}$$
 Equation 3

149

150 The methane volume produced in each test vial was then calculated using the pressure data combined

151 with the methane concentrations as measured by the GC-FID as is described in **Equation 4**.

Total Methane
$$(ml) = Total Biogas Volume (ml) * Methane Percent (%) Equation 4$$

152

153 **3.3.5** Rate of methane production

The methanogenic activity rate was calculated for each test by determining the slope of the best fit line through the steepest straight-line section of each plot of methane produced over time. The number of points used varied for each test condition, though no fewer than 3 points were used in this calculation and details are discussed in each experimental chapter.

A best fit slope was calculated (using Excel software) for each of the consecutive points using mls of methane per gVSS as the response variable (y) and time in hours at the treatment variable (x). The best fit slope with the greatest value was taken as the maximum rate value. This value was then divided by 24 to convert the figures to the units ml of CH4 / gVSS *day.

162 **3.3.6** Lag Time

As microbial populations adjust to their new environment there is often a delay between substrate addition and activity, known as the lag phase. The addition of amoxicillin within the culture medium was also found to influence the length of lag between substrate addition and biogas production (Chapter 6). To capture the influence of the amoxicillin on lag time, it was necessary to identify the first point of biogas production. For this study, length of the lag phase was defined as the time between substrate addition and first point used to calculate the maximum rate of methane production as found in (Section 3.3.5).

170 3.3.7 Calculating Percent Inhibition

171 To determine the inhibitory effect of amoxicillin on the potential treatment capacity of the microbial 172 communities, the percent inhibition of activity was calculated. Percent inhibition calculations were 173 modified from the calculations described in section 37 The OECD: Determination of the inhibition 174 of the activity of anaerobic bacteria: reduction of gas production from anaerobically digesting 175 (sewage) sludge (OECD, 2007). Although the OECD guidelines focus on total biogas produced, this 176 study has expanded this calculation to include relative inhibition of total biogas produced (mL), total methane produced (mL), and maximum rate of methane production (ml of CH4 / gVSS *day). Total 177 biogas and total methane production (reported as mean of triplicate tests) was defined as that which 178 had been produced in a given test condition 66 hours after the addition of substrate. The maximum 179

rate of methane compares the values as calculated in section 3.3.5. Each condition compares the
relevant value for the test condition (with amoxicillin) to the control condition (without amoxicillin)
(Equation 5).

Percent inhibition =
$$\frac{Pc}{\overline{Pt}} * 100$$
 Equation 5

183

184 Pc- The mean value within the control condition.

185 \overline{Pt} – The mean of the values for the test conditions

186

187 **3.4 MOLECULAR MICROBIOLOGY METHODS**

DNA was extracted using the Fast DNA spin kit for soil (MP Biomedical) and Fast Prep -24 instrument. The kit was primarily used as directed with additional optimisation for the sample mass of homogenised granule and length of bead beating for cell lysis (as detailed in **Table 3-1** and **3-2**) to maximise DNA yield. Method optimisation was conducted using the same batch of granules from which the experimental samples were taken. DNA concentrations were determined using the Qubit 2.0 Fluorometer and Qubit dsBR (Broad Range) Assay kit (Invitrogen).

Table 3-1 - Testing granule mass (g) used in the DNA extraction kits which would result in the highest concentration of
 DNA recovery (ng/ul).

Granule Mass (g)	DNA Concentration (ng/ul)	Conclusion
0.15	164	
0.25	236	Use 0.3g sample in the future.
0.35	225	

196

197

198

199

201	Lysis Time	DNA Concentration (ng/ul)	Conclusion
202	20 sec	185	Run the lysis procedure for 40
203	30 sec	195	seconds, as recommended by the kit
204	40 sec	323	instructions.
205			

200 *Table 3-2 Testing the lysis time for the DNA extraction kits for the highest concentration of DNA recovery (ng/ul).*

The quality of the extracted DNA during optimisation was inspected for shearing on a 1% agarose gel with a SYBR® Safe DNA Gel Stain (Invitrogen Life Technologies) with visualisation again a 1kb Plus DNA ladder (Invitrogen Life Technologies) and run at 90V for 40 minutes. It was found that, although some shearing was observed, the quality of the primary DNA was adequate for use in the downstream processes: PCR for next generation sequencing and qPCR for AMR quantification.

211 3.5 AMOXICILLIN QUANTIFICATION MY LC-UV METHOD DEVELOPMENT

212 **3.5.1** LC-UV Instrumentation and Operating Conditions

The amoxicillin was analysed by a Shimadzu LC20AT liquid chromatograph (Kyoto, Japan)
equipped with an SIL 20AHT Autosampler and a DGU-20 A Degassing unit. Chromatographic
separation was achieved using a Purospher® STAR RP-18 end capped column (5µm, 150mm x 4.6
mm). Detection was conducted using an SPD-10 UV Detector. The system was controlled by
LabSolutions (Agilent Technologies).

The column was conditioned by flushing a mixture of 1/9 Methanol and Acetate Buffer (0.01M, pH
5.0) through the column for 30 minutes. At the end of each set of runs the column was flushed with
4/1 Methanol and MilliQ (Millipore) water for one hour before storage to prevent the crystallisation
of salts on the column.

The optimal operating conditions for the 25 min method were: injection volume 100µl, flow rate 0.9
ml/min, mobile phase A: acetate buffer (0.01M, pH 5.0), mobile phase B: HPLC grade methanol.
The elution gradient: 0 - 4 min 10% B, 4 - 18 min 10 - 30% B, 18 - 20 min 30% B, 20 - 25 min 10%.

All samples were analysed using pH and conductivity. Conductivity readings were taken using an

226 Orion 013605MD conductivity cell. pH readings were taken using an Orion 8156BNUWP Ross Ultra

- 227 combination pH probe. Reads were measured using the Orion 5 Start meter.
- 228 3.5.2 Reagents and Chemicals
- 229 Amoxicillin was supplied by Sigma Aldrich. Methanol was HPLC grade and supplied by Sigma

230 Aldrich. Solutions were prepared with Milli-Q® Type 1 Ultrapure Water.

- Acetate buffer used in mobile phase was comprised of: Sodium Acetate 0.55g, Acetic Acid 0.12g,
- 232 MilliQ water 1000ml. The buffer was pH adjusted to pH 5.0 by adding NaOH (1M) as needed. All
- bottles were washed with methanol and MilliQ water before use during mixing. The mixed buffer
- was filtered through a 0.45µm filter (type) before use.

235 3.5.3 Preparation of Standard Stock Solutions

236 Stock solutions were prepared by weighing 0.2g of Amoxicillin powder into 200ml of MilliQ water

to create a 1000 ppm stock. This bottle was covered with foil and stored at 3°C for the next several

- 238 weeks to be used as a base for other stock dilutions. A 10ml sample of this solution was filtered
- through a 0.45um Whatman® membrane glass fibre filter and diluted with MilliQ to create a 100
- 240 ppm stock solution which was used as the base for all other dilutions.
- 241

Table 3-3 - Stock solutions used to optimise LC-UV method.

	Final Stock Volume (mg/L)	Concentration used for dilution (mg/L)	Volume of stock used (ml)	Milli Q Added (ml)	Total Volume (ml)
	10	100	1.0	9.0	10
	1	10	1.0	9.0	10
	0.5	1	5.0	5.0	10
	0.4	1	4.0	6.0	10
	0.3	1	3.0	7.0	10
	0.2	1	2.0	8.0	10
ſ	0.1	1	1.0	9.0	10

242

243 3.5.4 Repeatability

Repeatability measures the performance of a method on repeated samples within a single day, orintra-day variation of the peak area, retention times, and migration times. Each of the five

concentrations were analysed five times, three running from lowest to highest, and two running from
highest to lowest, with two blank samples of pure MilliQ water run between each set of standards.
The average values for the retention time for the 25min method (n=25) was 4.4088 min with an RSD
of 0.95%. The RSD values for the peak areas of each of the five concentrations (n=5) ranged from
1.97 to 7.44%. Over the five sets, the mean set run migrated from 4.42 min (set 3) to 4.42 (set 5) in
which set order and mean retention time for each set are not statistically significant (r² = 0.306). This
indicates that the repeatability of this method is acceptable.

253 **3.5.5** Linearity

Linearity is determined to include a series of at least five different concentrations within the target

range (FDA, 1995; US FDA, 1996). Linearity for the 25 min method was checked from a range of 0.1

-0.5 which included the concentrations (0.0, 0.1, 0.2, 0.3, 0.4, 0.5). Linearity was measured for

peak area for each of the standards resulting in a slope of 14025 and an intercept of -238.36 with an r^2 value of 0.995.

259



260



262

261

263 **3.5.6** Limit of Detection

264 The limit of detection and quantification was calculated using the signal to noise ratio method as 265 explained in Guideline for Industry: Text on Validation of Analytical Procedures (FDA, 1995; US 266 FDA, 1996). The baseline noise and signal to noise ratio (S/N) for each sample was calculated by 267 the LabSolutions (Agilent Technologies) software. Per these guidelines, the lower limit of detection (LOD) was defined as a S/N \ge 2 or a sample that causes a peak twice the height of the background 268 noise. The lower limit of quantification (LOQ) was defined as $S/N \ge 10$. The S/N values for each 269 270 standard were fitted to a regression line. These values were used to calculate the LOD and LOQ for 271 each method. The 25 min method generated a S/N regression with a slope of 61.994, an intercept -4.6478 of, and an R² value of 0.9369. These values resulted in a calculated LOD of 0.123 and an 272 273 LOQ of 0.236.



Figure 3-4 - Signal to Noise Ratio of Calibration Curve

274

² Chapter 4

Method Optimisation and Baseline Study - Specific Methanogenic Assay

6

7

8

9 4.1 INTRODUCTION

10 The specific methanogenic assay (SMA) was utilised to study the effect of amoxicillin on methane production and to infer the relative impact of amoxicillin on different trophic groups in anaerobic 11 12 sludge granules. The SMA was developed by D. Coates, F. Coughlan, & Colleran, (1996) and Shelton & Tiedje, (1984) and tracks the change in pressure in sealed bottles containing anaerobic 13 14 medium and a known feed type to infer the rate of methane produced by the anaerobic culture. Like 15 larger ecosystems, the mixed microbial communities found in anaerobic sludge granules contain 16 microbes mediating a trophic cascade in which different members of the population are capable of 17 metabolising organic matter of differing complexities which is sequentially broken down and 18 eventually converted into predominantly methane or carbon dioxide by methanogens (Stams, 1994; 19 Gerardi, 2003; Pol et al., 2004). Within the system, a wide variety of complex polymers are 20 hydrolysed and fermented by an equally diverse set of microbes into a relatively few simple 21 compounds most notably acetate, ethanol, propionate, and butyrate. Each of these compounds are 22 metabolised and transformed by a subpopulation of specialist organisms. The SMA aims to test the 23 activity of these specific specialised subclasses within the microbial community. When the test

24 bottles are fed specific substrates of differing complexities corresponding to a specific level of the anaerobic trophic cascade, the rate of methane produced can be used to infer the relative activity of 25 26 the members of that trophic level. An imbalance in microbial activity at any point in the methanogenic food chain in natural or engineered environments can result in the build-up of a single 27 28 food source which can in turn change the environmental growth conditions, such as pH or ammonia concentration, which can be toxic to the system setting off a chain reaction further reducing activity. 29 30 Thus, the methanogenic activity rates can be used to infer the activity of the community as a whole 31 and its capacity to degrade certain substrates, as well as overall system stability.

Whilst the monitoring and measurement methods associated with the SMA are well established, the
duration of the SMA and methods for sample preparation vary with sample type and substrates used.
Therefore, to inform subsequent experimental methodology and design, a preliminary
experiment was conducted to

36 (i) establish a baseline of activity in terms of maximum rate of methane production,
37 window of maximum activity, and time required to exhaust the substrates
38 provided; and

39 (ii) to optimise the sample preparation method.

40 **4.2 METHODS**

41 4.2.1 Preliminary Study Design Overview

An array of culture bottles were prepared and monitored in which each bottle contained a single substrate. Four substrates were used: acetate, ethanol, propionate, and butyrate (**Table 4-1**). Each culture bottle contained anaerobic granules, anaerobic medium, and one substrate (or water in the case of the no substrate controls subsequently called 'blanks'). Each condition was prepared in triplicate. Activity was measured through changes in headspace pressure (mv) using a pressure transducer. The pressure (mv) data was subsequently converted into total biogas and total methane volumes (ml). The bottles were monitored intermittently for 152 hours after feeding (6 days).

Parameter	Design Value or
	Condition
Temperatures	37°C
Feed Type	Acetate (3M),
	Ethanol (3M),
	Propionate (3M),
	Butyrate (1.5M)
Replicates	3
Bottles Used	18
pH	7.0
Bottle Volume	30ml
Granule Mass	0.5 g
Total Volume of Liquid Medium (Anaerobic Medium + granules)	10ml
Length of study	152 hours (6 days)

Table 4-1 - Study Parameters for Anaerobic Cultures

50

51 4.2.2 Investigation of sample preparation: Drain v Pipette

52 Optimum SMA preparation requires a consistent food to biomass ratio in each test vial to ensure that results are comparable between assay replicates and conditions. This is achieved through consistent 53 54 culture inoculation with equal masses (gVSS/L) of sludge granules within each culture bottle. When 55 weighing sludge granules however, liquid medium may be transferred to the weighing boat along with granules which could impact consistency of measurement. Thus, here, two different methods 56 for removing the water from the sludge granules before weighing them for use in SMA assays were 57 58 tested. The "drain method" removed the water by pouring a portion of granules over a sieve such that 59 the water would flow away from the granules, which would then be transferred to a weigh boat using 60 a spatula. This is the standard method used within the lab. In an alternative method, the "pipette 61 method", the water was removed using a pipette then the granules transferred to the weigh boat using 62 a spatula. Both methods were conducted inside an anaerobic chamber to preserve anaerobic conditions during the test. Granules were re-submerged in liquid in test vials within 5 minutes of 63 64 weighing to ensure granules did not dry out prior to testing (Figure 4-1).





65

Each method posed relative pros and cons. Visual inspection determined that the pipette method 66 67 preserved a wider range of granule sizes than the drain method as smaller granules were removed when the water drained out of the sieve. This is important because emerging evidence suggests that 68 69 granules of different size, putatively at different points in the life cycle of the granule (Trego et al., 70 2020), are both physiologically and ecologically different. It is possible that some of the biomass lost 71 through the sieve are flocs rather than active granules. Nevertheless, retaining a diversity of granule 72 sizes should better preserve the microbial community composition to reflect that within the larger 73 reactor from which the granules were sampled. Furthermore, as the pipette method ensured retention 74 of more small granules, the granules sample would have comparatively more surface area which 75 could therefore result in higher rate of methane production. Whilst the pipette method was effective at preserving granule size heterogeneity, removing the water from the granules reliably and 76 77 completely, using a pipette was challenging and inconsistencies in water removal could potentially lead to variation in the water content and therefore the volatile suspended solids mass (gVSS) within 78

each bottle. As the water was removed uniformly in the drain method there should be less variationof gVSS.

To test the influence of the different sample preparation methods, the baseline study was prepared (as described in Table 4-1) using the drain method and an additional set of three acetate fed bottles were prepared using the pipette method. Other than the difference in granule handling, the two experiment sets were the same. The preparation of all bottles and reagents was in accordance with the methods described in Section 3.1 of the Materials and Methods chapter and the analytical methods, including methane quantification, gVSS quantification and maximum rate of methane calculations, were as described in Methods Section 3.3.

88 4.2.3 Statistics

The effectiveness of the Drain and Pipette methods were analysed by comparing the mean rate of methane production in samples prepared by each method, and, by comparing the mean amount of gVSS in samples prepared by each method. It is noted that one of the bottles prepared using the drain method leaked (loss of gas) during monitoring and was therefore not included in analysis of data. Thus, the conditions were compared using the Welch two sample t-test which allows for a t-test comparison with uneven sample sizes and small counts as low as 2 (de Winter, 2013).

95 4.3 RESULTS AND DISCUSSION

96 4.3.1 Methane Production over time

97 The pressure readings and methane quantification measurements were used to estimate the methane 98 production volume over time for each test condition (Figure 4-2). All bottles showed an immediate 99 increase in pressure due to differences in temperature between the bench where they were sealed and 100 the incubator (where the pressure in the fed bottles increased with no lag or delay in gas production). 101 Reported methane production volumes were blank normalise, that is the mean volume of methane 102 produced in the blank (unfed) bottles was subtracted from each of the test conditions to calculate the 103 methane yield attributable to the substrate tested. For all feed types except for propionate, measurable biogas production ceased after between 16 to 32 hours. Bottles fed propionate did not appear togenerate any further biogas after 100 hours of incubation and monitoring.

Thus, when designing future studies, measurements within the first 48 hours should be considered the most time sensitive with regular readings taken at intervals of between 1 and 4 hours in order to capture a minimum of four observations during the period of the greatest rate of gas production. Furthermore, it can be assumed that in the absence of an inhibitor, that simple VFAs such as acetate, ethanol, and butyrate will have been consumed within 48 hours of the substrate addition whereas more complex substrates such as propionate may require longer observation periods to reach complete substrate conversion.

113 4.3.2 Granule handling- Drain v Pipette

The mean specific methanogenic activity rate using acetate as substrate was determined for samples prepared by two distinct methods: drained and pipette methods. The average activity rates for the pipette and drain bottles were 235.4 (mg CH4 /gVSS *day) and 258.0 (mg CH4 /gVSS *day) respectively. Although the mean of the drain method was slightly higher than the pipette method,



Figure 4-2 - The substrates provided were converted into methane.

The accumulation of methane over time can be used to calculate the rate of methanogenic activity. The length of the experiment was determined by the time required for all the provided substrate to be converted to methane after which the rate of methanogenesis would level off

indicating that the acetate was converted to methane at a slightly faster rate, there was no statistical
difference between the rates of methane production of the two methods was observed (p-value =
0.80) (Figure 4-3). Thus, as the drain method was less time consuming, this method was adopted for
use in all future experiments. As there were no statistical difference in the drain and pipette methods
the two methods will be treated as a single set of acetate fed bottles for the rest of this analysis.





Figure 4-3 – Comparison of drain and pipette method.

A comparison of the methanogenic activity of the granule handling using the pipette versus drain method. Each condition was fed acetate. Although the rates for the drain method were slightly higher, no statistical difference was found between them. The x represents the mean value, the centre line represents the median value. Each of the boxes represent the lower (Q1) and upper (Q3) quartile.

125 4.3.3 Methanogenesis and Study Length

126 The specific methanogenic activity rate of the sludge granules was determined for each of the 127 substrates tested (acetate, ethanol, propionate and butyrate). The bottles fed with ethanol had the greatest rate of methanogenic activity rate of 274.1 (SD = 47.7 RSD = 0.2) ml CH4/ gVSS \times day 128 129 (Figure 4-4). The bottles fed with acetate, ethanol, and butyrate had similar rate of gas production with mean rates of 219.4 (SD = 57.2, RSD = 0.3), 184.7 (SD = 15.3 RSD = 0.1), and 0.44 (SD = 130 131 0.02, RSD = 0.056) ml CH4/ gVSS × day respectively. The rate of methane production in bottles fed with propionate were lower with a mean of 124.1 (SD = 15.1, RSD = 0.1). The mean rate of methane 132 production in unfed 'blank' bottles was $0.11 \text{ ml CH}_4 / \text{gVSS} \times \text{day}$ (data not shown). Despite the lack 133 134 of added substrate, unfed blank cultures generally produce small amounts of methane, likely due to 135 the use of dead cellular materials or extracellular polymeric substances (EPS) as a substrate. A one-136 way analysis of variance model with a Tukey's post hoc assessing rate of methanogenic activity in 137 terms of feed type found that only the propionate and ethanol activity rates were statistically different 138 (p < 0.05). The higher rate of methanogenesis in for the ethanol fed samples is possibly due to the 139 cultures history treating distillery wastes.



Figure 4-4 – Rate of Methanogenic activity for different feed types.

The rate of methanogenic activity for each of the provided substrates indicated the activity of the sub communities required to convert the substrate to methane.

140 **4.4 SUMMARY**

Based on this work the following conclusions were made and used to inform future experimentaldesign.

There was no significant difference in activity observed using the drain and pipette methods
 for sample preparation. As the drain method was less time consuming and more practical to
 do while in the anaerobic chamber, this method was chosen for all future experiments.

To capture the maximum activity rate in assays without inhibitors, the first 20 hours require
 regular observation at 1–4 hour intervals.

In assays without addition of inhibitors, it can be assumed that simple VFAs such as acetate,
 ethanol, and butyrate will have consumed all the available feed within 48 hours after initial
 reading. More complex substrates such as propionate may require longer periods of to reach
 complete substrate conversion. As such, the minimum window of time for observation
 should allow for at least 60 hours for an SMA.

² Chapter 5

Method Development for Repeat Batch Feeding of Anaerobic Granules

6

1

7 5.1 INTRODUCTION¹

8 In a traditional SMA, anaerobic cultures are fed once. After the provided substrates are fully 9 consumed the study ends (Angelidaki et al., 2009; Shelton & Tiedje, 1984). Thus the SMA is useful 10 for determining 'as sampled' methanogenic activity, or, for comparing short term effects of inhibitory 11 compounds on activity (Cetecioglu et al., 2012). Capturing chronic effects of inhibitory compounds 12 on adaptation of a microbial community by contrast requires longer term studies. Additionally, longer 13 term studies of inhibition necessitate the maintenance (or approximation) of steady state conditions 14 which are maintained through the cycling of anaerobic medium and the regular addition of organic 15 substrates (Gerardi, 2003; McHugh et al., 2003). This is often achieved using bioreactors. Although 16 these systems are usually automated or semi-automated, they are also large and require specialised 17 and complex equipment to run, such as pumps, heaters, agitators, fume cupboards, and gas collection 18 rigs. The complexity and physical footprint, in addition to their associated costs, reduces the potential 19 to deploy such rigs in a study underpinned by a large number of test conditions or biological

¹ Note: This section of the thesis reports the development of a repeat batch culturing assay for anaerobic granules adapted from the SMA method. The method was developed to enable longer term repeat batch culture via manual removal and replacement of anaerobic media over several weeks incubation. Although the repeat batch method was not used experimentally due to the Covid 19 lockdowns and loss of laboratory access, the results of this work, which demonstrated the value of nutrient additives and feeding during acclimation, were adopted in the design of other work reported within this thesis. Thus, the development study is reported in detail here.

20 replicates. Anaerobic culture bottles, such as those used in the SMA, require little physical space,
21 and limited specialised equipment beyond an anaerobic chamber in which to fill the bottles, a
22 pressure transducer to measure pressure changes, and a shaking incubator to heat and mix the
23 cultures. As the physical space required for this method is much smaller, it is possible to run more
24 replicates and conditions within a single study. However, as each bottle requires manual handling to
25 refresh medium and feed, manual repeat batch feeding has the potential to be labour intensive.

26 Whilst enabling observation of a large number of different test conditions, the traditional SMA assay 27 does not provide trace elements for growth as may be required in a longer-term study. If an anaerobic 28 culture is not provided with the substances required for maintenance and growth, the cells will begin 29 to die, resulting in a reduction in activity and biomass. The time required for starvation and activity suppression differs between cultures and with the substrate makeup. Nutrients that may be essential 30 31 to support the growth of many microbes, if added to cultures in such a way that they build up within 32 a system, can however become toxic and inhibitory to growth and methanogenesis (Thanh et al., 2016). Metals such as copper, nickel, zinc, cadmium, and chromium, all of which are essential at low 33 concentrations, have been shown to inhibit acetogenesis and methanogenesis when concentrations 34 are sufficiently high (Mudhoo and Kumar, 2013). Therefore, to conduct a repeat batch assay, 35 36 understanding the requirement for and impact of nutrient dosing is essential.

This study used a repeat batch assay to assess whether anaerobic granules fed on glucose and a complex broth could maintain activity over a two-week period without addition of trace nutrients. Additionally, the effect of a nutrient additive mix on the activity of anaerobic granules fed on the same substrates was investigated to determine if the presence of the additives would increase activity and overall health or build up and become toxic. The 'health' of the community was assessed by monitoring:

43 (1) Variation in the volume of methane produced over time, whereby it was assumed that steady
44 methane production indicated a healthy culture while variable or diminished methanogenesis
45 indicates an unstable culture

46 (2) Methanogenic activity rate was measured three times during the study.

47 (3) Biomass in terms of gVSS – biomass was measured at the beginning and end of the study to determine if the culture was stable or had declined and lost biomass as a result. 48

METHODS AND MATERIALS 5.2 49

Study Design - Repeat Batch Method 50 5.2.1

51 This study measured the methanogenic activity of a granular sludge using simple (glucose) and complex feeds (complex broth) with and without added essential nutrients over 20 days using a repeat 52 batch culturing method (Figure 5-1). Anaerobic granules were placed in culture bottles with 53 54 anaerobic medium, substrates, and medium and incubated at 35°C. Every 48 hours, 24ml of the 55 medium was removed and replaced with fresh medium and the appropriate substrate/ nutrient 56 combination which resulted in a 2.5day hydraulic retention time (HRT). The conductivity of the removed effluent was assessed to infer nutrient uptake. Immediately before effluent replacement, the 57 58 pressure in the headspace of the bottles was measured and the biogas was sampled methane quantification. Pressure readings were taken once every 24 hours to measure net biogas 59 accumulation. At three points during the study (day 2, day 10, and day 18) a general methanogenic 60 assay (GMA) was conducted in which biogas was measured every 2-6 hours over a 48-hour period, 61 similar to the readings as would be taken during an SMA, and the data was used to assess the 62 differences in the overall methanogenic activity rate of the granules for each test condition. The 63 primary difference between an SMA and a GMA is that an SMA uses an array of substrates with 64



Figure 5-1 - *Summary of the study design and the monitoring for each phase.*

different complexity to target individual trophic levels, while the GMA used more complex
substrates and tracks methane for the community as a whole. The gVSS of the granules was measured
at the beginning and end of the study to enable estimation of the change in biomass within each bottle
during the study period. Six test conditions were tracked: three substrate types (blank, glucose,
complex broth) and two nutrient conditions (with and without nutrient additives). All test conditions
were conducted in triplicate for a total of 18 test bottles.

71 **5.2.2** Culturing

72 5.2.2.1 Bottle Preparation and monitoring

73 The bottles used were 100ml in volume. All samples were prepared in triplicate including six blanks 74 (three with nutrient and three without) into to which no substrate was added. All bottles were filled under anaerobic conditions within an anaerobic chamber. Each bottle contained 1.5g of sludge (10 75 76 g/L of VSS) to which 28.5ml of anaerobic medium was added. After inoculation, the headspace in 77 the bottles was exchanged for 90 seconds with a 20% CO_2 and 80% N_2 mixture to 1 ATM to ensure 78 an anaerobic environment within each bottle. Samples were then acclimated without food for three 79 days within a 37°C incubator. The activity of cultures were monitored according to the protocol by Colleran et al (1992) as described previously (Section 3.3). 80

81 5.2.2.2 Repeat feeding

82 After acclimation, the bottles were returned to the anaerobic chamber and the liquor was completely 83 removed from each of the bottles using a pipette, leaving the granules behind. Then, 27.5ml fresh 84 anaerobic medium was added to the acclimated granules. Bottles were sealed within the anaerobic 85 chamber. The headspace was exchanged for 30 seconds with a 20% CO₂ and 80% N₂ mixture to 1 ATM. After the head space gas was exchanged, the substrate and nutrients were injected into each 86 bottle through the septa. The bottles were then incubated for 48 hours (with the monitoring period 87 88 adopted from results of Methods Section 3). Pressure readings were taken using a Centrepoint pressure transducer at 24 and 48 hours after the addition of the substrate. After the 48-hour pressure 89 90 reading biogas was sampled from the headspace for methane quantification. This process was

93 5.2.2.3 Methanogenic Activity Measurements

94 Pressure readings were taken approximately every 1-2 hours until gas production began to level off, after which readings were taken every 4-6 hours. These assays were conducted at three points in the 95 96 study (day 2, day 9 and day 18) to determine the rate of methanogenic activity for each of the test 97 conditions. After approximately 48 hours biogas and effluent samples were taken for analysis and the medium was replaced at described above. The final pressure measurements were read at 44 hours, 98 99 46 hours, and 40 hours from substrate addition for the first second and third assays respectively. As 100 the presence of added nutrients might have an effect on the blank conditions, these were treated as 101 test conditions in their own right rather than being used to blank normalise the data of the fed test 102 conditions. Therefore, in order to compare the rate of the methane in the fed and unfed conditions, 103 the reported rates were not blank normalised as is done in other studies. The method used to calculate 104 the rate of methanogenic activity was as described in Methods chapter Section 3.3.

105 5.2.3 Reagents and media

106 In contrast to the VFAs used in a typical SMA, the GMA used two different substrates as feed for 107 the microbes, one complex (beef broth, yeast extract, and glucose) and one simple (glucose). These 108 were selected because utilisation of the complex broth would require activity from all levels in the 109 methanogenic food chain whilst utilisation of glucose, a soluble sugar, would not require hydrolytic activity. The complex broth (modified from the OECD 224 (OECD, 2007)) contained a mix of 5g of 110 beef broth, 5g of yeast extract, 5g of D-glucose in 100ml of DI water. The 1.0 M glucose solution 111 was comprised of 18.16g of glucose in 100ml of DI water. As the complex broth has a shelf life of 112 24 hours, it was portioned into 2ml vials and frozen at -20c immediately after being made. Vials were 113 thawed immediately before use. The glucose solution was stored in the refrigerator at 4°C for the 114 115 length of the experiment. All mixes were autoclaved before storage to prevent contamination. The nutrient mixes, buffers, and anaerobic medium were prepared as described in section 3.2.2 and the 116 117 final solutions prepared by mixing as described in Table 5-1.
118 *Table 5-1- Substrate additives.*

There were six feeding conditions for this study with two test conditions: feed type (blank, glucose, and complex broth)
 and nutrient (with and without). The nutrient mix included several solutions which were stored separately and had to be

121 mixed on the day. The volumes of substrate used where normalised such that each contain 8mg TOC/L in each bottle

122

Test condition	Complex Broth (ml)	Glucose- 1M (ml)	Phosphate Buffer (ml)	Nutrient mix (ml)	Trace Nutrient mix (ml)	Mili-Q (ml)	Anaerobic medium (ml)	Total Volume added (ml)	
Blank + No	0	0	0	0	0	15.00	15	30	
Trace Nutrient	Ŭ	ů	Ū	0	Ũ	10100	10	20	
Blank + Trace	0	0	15	15	0.15	11.85	15	30	
Nutrient	0	0	1.5	1.5	0.15	11.05	15	50	
Glucose + No Trace Nutrient	0	0.3	0	0	0	14.70	15	30	
Glucose + Trace Nutrient	0	0.3	1.5	1.5	0.15	11.55	15	30	
Broth + No Trace Nutrient	0.36	0	0	0	0	14.64	15	30	
Broth + Trace Nutrient	0.36	0	1.5	1.5	0.15	11.49	15	30	

123

124 5.2.4 Analytical methods and Calculations

125 5.2.4.1 Methane calculations

126 Methane volume and rate was calculated as described in Method Chapter Section 3.3.

127 5.2.4.2 gVSS

Similarly to the SMA assay, volatile solids measurement was used to imply the biological content of the samples. The gVSS of the cultures was measured at the beginning and end of the study as a means of tracking changes in total biomass in response to the provided feed types. A decrease in VSS over time was assumed to indicate that the given substrate did not contain the necessary constituents for maintenance and cell growth. The gVSS methods and calculations were as described in Section 3.3.

133 **5.2.5** Statistics

All statistics were calculated using R version 4.1.2 and MS Excel. To measure variation in the volume and relative proportion of the methane produced, the standard deviation (SD) and relative standard deviation (RSD) was calculated for the methane volume (ml) for each of the feed types. Deviations were computed including all time points and all replicates during the study for each test condition. A Welch's 2 sample t-test was used to determine the influence of nutrient additives on the rate of methanogenic activity for each feed conditions and p-values reported. An ANOVA was used to

determine the influence of feed composition on difference in gVSS between the beginning and endof the study and 'X'-values reported.

142 5.3 RESULTS AND DISCUSSION

143 5.3.1 Variation in Methane Volume

144 The volume of methane during the repeat batch method was consistent for most of the test conditions 145 (Figure 5-2). The glucose with supplemented nutrient, complex broth with, and complex broth without nutrient produced mean of 281 ± 42 (RSD = 15.05%), 252 ± 42 (RSD 16.68%), and 240 ± 41 146 147 (RSD 17.26%) ml CH₄/gVSS respectively. The glucose fed cultures without supplemented nutrients 148 produced less overall methane with an average of 155 ± 68.2 ml CH4 /gVSS and production was 149 relatively more variable (RSD 44.07%) (Figure 5-3). Similarly, both complex broth conditions as 150 well as the glucose with added nutrient conditions showed an increase in net methane production 151 throughout the study, while the glucose without nutrient additives decreased in overall methane 152 output during the study. The variation in methane volume in the glucose without nutrient additives 153 is possibly due to deterioration of the community as the lack of essential nutrients prevented efficient 154 protein and enzyme production.



Figure 5-2 - *The mean volume of total methane at each changeover throughout the study. The error bars represent one standard deviation from the mean.*



Figure 5-3 - A comparison of the total methane generation at each time point measured during the study. Box plot represent 1 quartile of variation with an x representing the mean value. The effect of nutrient for each feed type was determined using a t-test (p < 0.05 = *, p < 0.01 = **, p < 0.001 = ***).

156

157 5.3.2 Methanogenic Activity

The GMA was measured at the beginning (Day 2), middle (Day 9), and end (Day 18) of the study 158 (Figure 5-4). Biogas production was nearly complete by 35 hours for all test conditions except those 159 160 that were fed with glucose and no nutrient, which did not reach stationary phase during the 48-hour 161 observation period (Figure 5-4a). A lag phase was observed for all the test conditions for the initial 162 GMA. Complex broth fed cultures, both with and without nutrient addition, began biogas production 163 between 5 to 7 hours into the experiment. The cultures with glucose and nutrient additives began 164 producing methane around 17 hours while the bottles without glucose did not begin producing 165 methane until after 20 hours into the study. The lag period was not observed in later GMA tests, 166 rather, for the remainder of the study biogas production was observed immediately after addition of 167 additional substrate for all substrates (Figure 5-4 b & c). The initial lag observed was due to the time 168 required for the available nutrient to be taken up and utilised by the microbial communities. However, 169 the lack of a lag time observed in the later study is likely due to some residual substrates remaining 170 within the granules.



Figure 5-4 - Cumulative methane produced for each of the GMA assays conducted - day 1(a) day 9 (b) and day 20(c). Each line represents the mean value for each time point. The error bars represent one standard deviation from the mean.



Figure 5-5 - *A* comparison of the rate of methanogenic activity for each of the GMAs conducted at Day1, Day 9 and day 20 of the experiment.

172 The influence of nutrient additives on methanogenic activity (Figure 5-5) varied significantly (p - 1)173 values <0.001 - 0.08) between each substrate type used. Glucose with nutrient additives had the 174 highest activity with a mean rate of 312.2, 228.96, and 275.79 ml ch4/gVSS*day for the first, second 175 and third GMAs respectively. The rate of methanogenesis declined throughout the study for cultures 176 fed glucose, regardless of nutrient supplementation. By contrast, methanogenic activity for bottles 177 fed with complex broth both with and without nutrients increased between the first and second GMA, 178 though decreased slightly for the third GMA although not significantly (p>0.05) (Table 5-2). The 179 higher rate of methanogenesis for the glucose fed bottles with added nutrients is likely related to the 180 cultures previous exposure to glucose leading to a community adapted to that as its primary substrate. 181 The slower rathe of methane production for the complex broth suggested that the community was 182 not adapted to utilise this substrate as efficiently. As the community adapted through the experiment, 183 the rate of methane production for the cultures fed the complex broth increased.

184 Table 5-2 - The influence of nutrient additives on methanogenic activity for each of the different substrate types.

185 Nutrients increased the methanogenic activity rate produced for glucose substrates such that the difference was strongly

significant for the first and final GMA (p<0.001). The nutrient decreased the activity rate for broth fed bottles, though

187 *this effect was not found to be statistically significant.* (P>0.5)

		Mean Methanogenic Activity (Ml CH4 / gVSS *day)		Difference in Mean from Nutrient addition	t	df	p-value
		With Nutrient	Without Nutrient				
	GMA 1 (Day 2)	466.5 ±73.2	154.3 ±39.2	312.20	6.5	3.06	< 0.001
Glucose	GMA 2 (Day 9)	371.0 ± 15.7	142.1 ±122.5	228.96	3.2	2.07	0.08
	GMA 3 (Day 18)	364.2 ± 16.3	88.4 ± 10.4	275.79	24.7	3.40	< 0.001
	GMA 1 (Day 2)	265.5 ± 53.5	292.4 ±5.7	-26.91	-0.9	2.46	0.4756
Complex Broth	GMA 2 (Day 9)	341.1 ±111.5	457.2 ±62.4	-116.06	-1.6	3.14	0.298
	GMA 3 (Day 18)	319.3 ±13.6	359.2 ± 20.1	-39.93	-2.9	3.51	0.05396

188

189 5.3.3 gVSS Before and After the Study

190 The volatile solids content of the granular sludge was measured at the beginning and the end of the 191 study to determine whether the substrate and nutrient conditions provided were sufficient to sustain 192 biomass (Figure 5-6). An ANOVA determined that the gVSS at the beginning and the end of the study was associated with the feed type (p-value <0.001) but not to the addition of nutrient (p-value 193 - 0.874). While the unfed bottles had 0.24g less gVSS on average by the end of the experiment, the 194 195 glucose and nutrient fed bottles had 0.006g and 0.00027g less gVSS on average respectively. T-tests 196 comparing the gVSS in the seed community at the beginning of the study with the gVSS at the end 197 of the study found that both the blank conditions were statistically different from the seed community (with nutrient p = 0.003, without nutrient p < 0.01) while all the glucose and complex broth 198 199 conditions were not statically different from the seed (p > 0.10). These results suggest that in the 200 absence of additional feed sources, the microbes in the unfed bottles utilised other biomatter in the 201 bottles as feed and for biogas production. As the complex broth contained proteins, lipids, and 202 carbohydrate all the macromolecules required for cellular construction were present. As such, the 203 biomass was maintained throughout the study for those conditions.

204



Figure 5-6 - The gVSS for each test bottle. Differences between gVSS for each feed condition. The box plots represent the first quartile, and the x represents the mean value. The p-values are from t-test comparison from the seed condition.

205 5.4 CONCLUSIONS

206 The effect of nutrients on methanogenic activity and VSS content of the biomass depended on the 207 substrate complexity. Nutrient additives had no influence on the volume, rate, or variability of 208 methane generation for cultures fed a complex substrate. Furthermore, no statistical difference in 209 biomass (VSS) was detected between the beginning and end of the study for both feed conditions. 210 Conversely the rate, volume, and stability of methane produced in cultures fed the simple substrate 211 glucose differed greatly with and without the addition of nutrient. While the addition of nutrient 212 additives resulted in methanogenic activities and volumes similar to those fed a more complex 213 substrate, the biomass at the end of the study had decreased for both of the glucose fed conditions.

214 In summary:

Nutrient additives will be used for all future analyses as their presence has an immediate
 positive effect on the rate, volume, and stability of methane generation, even in short studies.

217	•	Substrate will be added at the offset of the acclimation phase as this eliminates the lag phase.
218		This will shorten the length of the overall observation widow and increase the likelihood of
219		observing the period of maximum methanogenic activity.
220	•	The maintenance of community biomass requires the addition of complex substrates for long
221		term studies. Glucose and nutrient additives on their own are not sufficient to maintain
222		community biomass.

1

² Chapter 6

- ³ How Acute Amoxicillin
- ⁴ Exposure Effects
- ⁵ Methanogenesis within
- ⁶ Trophic Cascade of Mixed
- 7 Anaerobic Communities
- 8

9 6.1 INTRODUCTION

10 Beta-lactam antibiotics, such as penicillin and amoxicillin, are amongst the most widely used antibiotics throughout the world (Van Boeckel et al., 2014). They are used in a variety of industries 11 12 and applications including human health, veterinary care, and agriculture (Patel et al., 2019). Globally, beta-lactams make up roughly 65% of the antibiotics market, though usage varies 13 regionally (Githinji et al., 2011). As with all pharmaceuticals, some portion of ingested antibiotics 14 15 will pass through our bodies largely unchanged and enter our sewers and eventually wastewater treatment facilities (Tran, Reinhard and Gin, 2018). These facilities, which are designed to remove 16 organic matter from the influent, are not optimised for pharmaceutical removal (Patel et al., 2019). 17 18 As a result, many pharmaceuticals pass through wastewater treatment plants and are discharged into 19 the receiving water environment with the treated effluent (Kümmerer, 2009a, 2009b; Githinji et al., 20 2011; Surridge and Watson, 2012; Fekadu et al., 2019; Peña-Guzmán et al., 2019).

However, in much of the world conventional centralised wastewater treatment plants such as thosemonitored in the studies reported above are not viable due to technical, economic, or environmental

23 constraints. Rather, small or decentralised wastewater treatment technologies are used, of which septic tanks are the most common (EPA, 2002; Sharma et al., 2013). Unlike centralised systems, 24 25 influents into the tank are not diluted by external waters such as storm water or industrial effluents. 26 Thus, septic tanks often contain a higher relative proportion of blackwater than influents to centralised water treatment facilities (Conn et al., 2006). Resultantly, the concentration of organics 27 28 within septic tanks as well as pharmaceutical pollutants can be one or two orders of magnitude more 29 concentrated (Kujawa-Roeleveld and Zeeman, 2006; Luostarinen et al., 2007; Conn, Lowe, et al., 30 2010; Strande et al., 2018).

31 Due to the low detection rates of beta lactams in centralised western wastewater treatment facilities, 32 they are often cited as low risk pollutants (Kümmerer, 2009a). However, given the differences in 33 concentration between the influent to septic tanks and centralised water treatment facilities coupled 34 with the low rates of monitoring for septic systems, it is possible that the reports of beta-lactams in 35 water treatment systems may underestimate true global exposures. High concentrations of undigested 36 antibiotics in influent have the potential to harm treatment systems, the wider environment, and 37 human health in several ways. First, if concentrations are high enough to inhibit biological treatment, 38 then the quality of effluent exiting into the environment could become compromised (Amin et al., 39 2006; Cetecioglu and Orhon, 2018) and could become a risk to human health. Secondly, undigested 40 antibiotics and potentially antibiotic resistant microbes could leave the systems in the effluent to then 41 enter the environment and waterways and become vectors for disease and impact natural microbial 42 diversity in receiving environments (Kümmerer, 2009a). Due to this lack of research into amoxicillin in the water environment, in 2018 amoxicillin was added to the latest update to the Water Framework 43 44 Directives emerging pollutant watch list (Loos *et al.*, 2018). Given the global reliance on septic tank technologies, to understand the impact of beta lactams on water treatment, it is important to 45 46 understand the effect that amoxicillin would have in septic tanks specifically and anaerobic systems 47 generally.

Research into the effects of beta-lactams on anaerobic systems specifically is rarely studied. Unlike
bacteria, methanogenic archaea, the primary microbes responsible for the production of methane, are

50 naturally resistant to beta-lactam antibiotics due to their lack of peptidoglycan in their cell walls (Kandler and König, 1998; Khelaifia and Drancourt, 2012). However, anaerobic communities are 51 comprised of a wide variety of microbes (Trego *et al.*, 2020), with a potential for varying degrees of 52 53 susceptibility to beta-lactam antibiotics. When research into the impacts of beta-lactam antibiotics on anaerobic communities is conducted, the conditions of these studies are varied and often focused 54 55 on concentrations higher than we might expect in wastewater environments, and the results do not 56 always agree. In general, it was found that the impact of beta lactams on treatment efficiency are 57 limited at low concentrations. For example, Sanz found that the addition of 10 mg/L penicillin inhibited total methane production by 25-45% (Sanz, Rodríguez and Amils, 1996) when VFAs were 58 59 used as substrate. Anaerobic digesters running on pig slurry from animals with 16 mg/kg in of 60 penicillin in their feed reported a 35% reduction in methane generation when compared to the 61 digesters running wastes from pigs fed a control feed (Massé et al., 2000). Zhang et al found that 62 anaerobic systems acclimated to amoxicillin could withstand concentrations up to 60mg/L before displaying any inhibition of nitrogen removal (Zhang et al., 2015). A study by Su et al (2019) did 63 64 not report any inhibition of methane production in anaerobic granules exposed to 100mg/L amoxicillin (Su et al., 2019). 65

66 The aim of this study is to assess the impact of short-term exposure to amoxicillin on the methanogenic activity of different trophic groups within a mixed anaerobic microbial 67 68 **community**. The range of amoxicillin concentrations used in this study aimed to include those that 69 might conservatively be found in small septic tanks in western countries (0.5 - 2.5 mg/L) through to 70 higher concentrations more potentially found in areas of less regulated distribution (5 - 25 mg/L) and 71 extreme concentrations (50-100mg/L). The study asks if there is a relationship between the specific 72 methanogenic activity of each trophic group with respect the rate and volume of methane/biogas 73 production and the amoxicillin concentration in the influent.

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76

77 **6.2 Methods**

78 6.2.1 Experimental Design

79 The experimental design was based on the specific methanogenic activity (SMA) (Colleran et al., 80 1992; D. Coates, F. Coughlan and Colleran, 1996) assay, in which the rate of change of pressure 81 within sealed bottles containing the test microbial community and a range of substrates, is used to 82 assess the activity of the microbes within the culture capable of breaking down those substrates 83 (Figure 6-1). Both the rate of production and volume of methane produced were determined. To 84 explore differences in susceptibility to amoxicillin inhibition within specific groups in the trophic 85 cascade, substrates of varying complexity (H₂/CO₂, acetate, ethanol, propionate, butyrate, glucose, 86 and a protein rich complex broth) were used as substrate. The experimental set up consisted of 135 87 sealed bottles, each containing anaerobic sludge granules and anaerobic medium comprised of 88 buffers, oxygen scavengers, and an oxygen indicator as described in Sections 3.2.1 and 3.2.2.



Figure 6-1 - Experimental design.

The effect of amoxicillin on exposure was assessed through changes in 1) rate of methanogenic
activity, 2) lag time of methanogenic activity, 3) total volume of methane produced. Each condition
was compared against a control.

92 6.2.2 Reagents and Media

93 Seven substrates of varying complexity were used, targeting each level of the anaerobic trophic cascade. Activity against each substrate was assayed at each antibiotic concentration (from 0-94 95 100mg/L) and all combinations were assayed in triplicate (Figure 6-2). Six liquid substrates were 96 used: acetate (3M), ethanol (3M), propionate (3M), butyrate (1.5M), glucose (1M), and a complex broth solution at (150g/L). A mix of 20% H_2 and 80% CO_2 was used as a gaseous substrate. The 97 98 VFAs (acetate, propionate, ethanol, and butyrate) were first neutralised using 8.0M NaOH to a pH 99 of 7.0 then diluted with MiliQ water to the target concentration and then placed into vials and sealed. 100 The 1M glucose was prepared by mixing 18.0g of glucose granules with 100ml MiliQ water. The 101 complex broth was prepared by mixing glucose (10g), yeast extract (10g), and beef broth (10g) into 102 200ml MiliQ water. After preparation, the complex broth was stored at -20°c until use.

103 To understand the inhibitory effects of amoxicillin on methanogenesis within an anaerobic 104 community two feed types were used as primary substrates: acetic acid as a liquid substrate and a 105 mix of carbon dioxide and gaseous hydrogen as a gaseous substrate. Methane is formed by three



Figure 6-2 – The anaerobic trophic cascade and the associated substrates used in this study.

106 different groups of bacteria: (1) hydrogenotrophic methanogens which directly convert CO_2 and H_2 107 into methane, (2) acetoclastic methanogens which use acetate, and (3) methylotrophic methanogens 108 which degrade containing one carbon such as formate and methanol. Acetoclastic methanogenesis 109 is the most common form of methane production and it is estimated that two thirds of all biogenic 110 methane in the atmosphere is from the degradation of acetate (Gerardi, 2003; Fournier and Gogarten, 111 2008). When macromolecules are consumed during anaerobic digestion H_2 and CO_2 are commonly 112 generated as by-products (Vanwonterghem et al., 2016). As methane production is the final stage in 113 anaerobic digestion it is important to understand the influence of amoxicillin on methanogens.

To understand the inhibitory effects of amoxicillin on acidogenesis within an anaerobic community glucose was used as the primary substrate. The second state of anaerobic digestion is acidogenesis in which fermentative bacteria utilise simple sugars for energy, generating volatile fatty acids in the process. The fermentation of glucose can take a variety of different pathways which are conducted by a variety of bacterial types, though the most common products are propionate, butyrate, ethanol, acetate, formate, and CO_2 (Gerardi, 2003).

To understand the inhibitory effects of amoxicillin on hydrolysis within an anaerobic community, a 120 121 mix of proteins and carbohydrates (complex broth) was used as the primary substrate. Unlike other processes of anaerobic digestion, in which the substrates are directly consumed by the microbes, 122 123 hydrolysis is undertaken by exoenzymes excreted from the microbe and takes places outside the cell. As a result, the microbe does not directly gain energy from this step. However, the products of 124 125 hydrolysis – simple sugars, amino acids, and fatty acids – are consumed, often, by the same bacterial 126 types which produced the hydrolysing enzymes (Gerardi, 2003). The amino acids produced by the 127 hydrolysis of proteins vary greatly in size and structure. As a result the products of the amino acid 128 fermentation are equally diverse and nearly half of the products are gaseous and would be available 129 for conversion to methane without requiring acetate formation as part of the pathway (Breure, 130 Mooijman and van Andel, 1986).

To understand the inhibitory effects of amoxicillin on acetogenic bacteria within an anaerobiccommunity ethanol, propionate, and butyrate were used as primary substrates. Fermentation is

undertaken by a wide variety of bacteria which syntrophically degrade soluble sugars and amino
acids into acetate and other by-products such as CO₂ and H₂ (Stams, 1994). While all three substrates
in this study are commonly utilised in acetate production, the over production or build-up of
propionate in anaerobic systems can suppress methane formation and is sometimes used as an
indicator of system instability (Ren *et al.*, 2007). Therefore, it is important to understand how acute
exposure to amoxicillin might influence methane production and substrate utilisation of VFAs.

The base anaerobic medium was prepared as described in Methods Section 3.2.1 of the methods chapter to which antibiotic and substrate were added, the amount of which depended on the phase of the experiment. All conditions contained nutrient additives which were prepared as described in section 3.2.2 (Trace Metals and Additives) of the methods section. The make-up of each of these solutions and the proportion of each in the final mixture is outlined in **Table 6-1** below.

Compound	Unit	Concentration of Steel	Phase		
Compound	Omt	Concentration of Stock	Acclimatization	Test	
Granule	(g)		0.5	0.5	
Feed Substrate	(ml)	Various	0.1	0.1	
Antibiotic	(ml)	10 x Final Concentration	0.0	1	
Phosphate Buffer	(ml) 20x		0.5	0.5	
Nutrient mix	(ml)	20x	0.5	0.5	
T. Nutrient mix	(ml)	200x	0.05	0.05	
Anaerobic medium	(ml)	<i>1x</i>	8.35	7.35	
Total Volume added	(ml)		10	10	

144 *Table 6-1 - Components within each bottle.*

145

146 6.2.3 Conducting a Specific Methanogenic Activity Test

147 6.2.3.1 Bottle Preparation and Granule Acclimation

Bottles were prepared in accordance with methods section 1.1. Two bottle sizes were used: 30ml for liquid substrates and 60ml bottles for gaseous substrate. All samples were prepared in triplicate including three blanks for each bottle size to which no substrate was added. A method development study (See Methods section 3.1) indicated that providing the microbes with some feed during the acclimatisation phase would reduce the lag phase. This was advantageous as doing so increases the

likelihood of documenting the active phase of methane production. To commence the acclimation
phase (bottles fed and incubated with the specific substrate but in the absence of antibiotic), 0.1 ml
of substrate was injected into the bottles through the septa before transfer to an incubator. The
headspace gas in bottles prepared for gaseous substrates were pressurised for 60 sec at 1.4 psi of CO2
H₂ before transfer to an incubator. The acclimation incubation was at 37°C for 5 days.

158 6.2.3.2 Running the SMA

159 After the acclimation incubation, bottles were opened in the anaerobic chamber and the liquid 160 medium within each bottle was removed using a syringe. A mix of fresh anaerobic medium, buffer, 161 and nutrient were added to the acclimated granules. These were then sealed within the anaerobic 162 chamber. The headspace in the bottles was exchanged for 90 seconds with a 20% CO_2 and 80% N_2 163 mixture to 1 ATM to ensure an anaerobic environment within each bottle. At this point the SMA 164 was conducted as described in Methods Section 3.1.2. The appropriate amount of substrate was inject into the bottles through the septum at the beginning of each experiment. The study was conducted 165 166 over 66 hours.

167 6.2.4 Analytical Methods

168 6.2.4.1 Quantifying Methane

Methane in the headspace of bottles was quantified using a GC-FID (as described in Section 3.2 of the Methods chapter). The methane content of the biogas in each bottle was measured at the end of the study after the last pressure reading was taken, and before opening the bottles to remove the sludge granules and effluent. Due to an instrumentation error, methane quantification was not measured for the gaseous substrates using GC-FID. However, assuming that the pressure differential during the experiment is the result of gaseous substrate utilisation as described in D. Coates *et. al.* (1996).

176 6.2.4.2 gVSS

177 The gVSS methods and calculations were as described in Section 3.3.

178 6.2.4.3 Converting bottle pressure into volume of methane

To calculate the rate and total gasses produced, the pressures measured for each bottle must be converted to volume (ml) of total biogas. This is achieved through calibrating the bottle calibration and is descried in detail in the Methods Section 3.3.3.

182 6.2.4.4 Rate of methane production

Specific methanogenic activity rate is reported as ml of CH_4 / gVSS / day. The SMA rate was calculated for each test by determining the slope of the best fit line through the steepest straight-line section of each plot of methane produced over time. The number of points used varied for each test condition, though no fewer than 4 points were used in this calculation. More details about the exact calculations used can be found in the methods section 3.3.5.

188 6.2.4.1 Defining Lag Time

Lag time was calculated as defined in section 3.3.6 of the Methods

190 6.2.4.2 Total Methane Volume

191 The total methane values were taken from the total methane calculated for at 66 hours. The values192 are reported in volume (ml).

193 6.2.4.3 Results Statistics

To determine if the two runs could be treated as a single run, the rate of methanogenesis and total methane produced in the control conditions for each of the groups was assessed using a Student's T-Test. To determine the relationship between increasing amoxicillin concentration on methanogenesis and the total volume of methane an aov was conducted. Each batch was analysed separately set (Run 1: 0.5 - 5.0 mg/L amoxicillin; Run 2: 10 - 100 mg/L) using amoxicillin. All statistical analyses were performed using R statistical package and RStudio 4.1. Additionally, the percent inhibition values were calculated in comparison to controls.

201 **6.3 RESULTS**

This study was designed to explore the short-term effects of amoxicillin exposure on the activity of anaerobic microbial communities. This was observed through changes in the specific methanogenic activity rates and the volume of methane produced in response to specific substrates and amoxicillin exposure.

206 6.3.1 Measuring the Batch Effect for Each Run

Due to the number of bottles, the study was split into two runs. Run 1 contained all substrates with amoxicillin concentrations 0.5, 1.0, 2.5, and 5,0mg/L. Run 2 contained all substrates with amoxicillin concentrations 10, 25, 50, and 100mg/L. The difference between the blank adjusted specific methanogenic activity rate observed for each substrate without the addition of amoxicillin (i.e. the controls) between runs was assessed using a T-Test for each individual substrate type.

212 No statistical difference between the methanogenic activity was found for (p > 0.05) any condition 213 between the two runs except for propionate. The rate of methanogenic activity in the propionate fed 214 bottles in the first run was significantly higher than in the second run (p = 0.002) (Figure 6-3). 215 However, this is likely due to the differences in the rate of methane production in the first run 216 compared to the second run. Whilst there was also variation in methanogenic activity between the 217 replicates in the glucose and H₂/CO₂ fed bottles for each run, no significant difference was observed 218 between the rates determined. The Student's t-test determined no significant difference (p>0.05) in 219 the volume of methane produced in the controls for each substrate tested between Run 1 and Run 2. 220 (Figure 6-4).

The batch effect on the activity is likely caused by the time between the test runs and possible differences in feeding. While seed community had regularly been fed glucose for a week before the analysis began, by the time the second run was conducted the seed community would have had an additional 4 weeks of glucose. The results from Chapters 5 and 7 of this thesis suggest that that would have been enough time to drive changes in the community make up. Nevertheless, the analysis of the control conditions for the two runs indicates that although there were some differences in the rate and volume of methane generated for each of the different feed types, the runs were similar enough

to justify further analysis of amoxicillin inhibition on these cultures. The relative changes will be
reported as a change from the batch control for the respective run. Additionally, linear models
assessing the effects of amoxicillin will be run separately for each batch.



Figure 6-3- The total volume of the methane produced (ml).

The bar represents the mean value. The error bars represent one standard deviation from the mean. The solid bars represent the first run while the striped bars represent the second run. None of the conditions were statistically different from eachother. Both the left and right axis represent methane volume on different scales.



Figure 6-4 - *The rate of methanogenic activity (CH4 (ml) / gVSS * day).*

The bar represents the mean value. The error bars represent one standard deviation from the mean. The solid bars represent the first run while the striped bars represent the second run. Only the propionate values were statistically distinct from eachother. Both the left and right axis represent methanogenic activity (CH4 (ml)/gVSS *day) on different scales

232 6.3.2 Amoxicillin and Rate of Methanogenic Activity

233 This study aimed to assess how acute exposure to amoxicillin would influence the activity of different 234 subpopulations of an anaerobic community. To accomplish this, specific methanogenic activity tests 235 (SMA) were run for a suite of 7 feed types (complex broth, glucose, butyrate, propionate, ethanol, acetate, and H₂/CO₂) with and without exposure to amoxicillin. This was conducted over two runs. 236 237 The first run tested the amoxicillin concentrations 0.5, 1.0, 2.5, and 5.0 mg/L. The second run tested 238 amoxicillin concentrations 10, 25, 50, and 100 mg/L. The effect of amoxicillin on the community 239 was measured against the methanogenic activity, lag in activity, and total volume of methane 240 produced. Plots of the cumulative methane produced can be found in **Figure 6-5** plotted by feed type 241 with curves showing the response in the presence of each concentration of amoxicillin addition 242 alongside the response of the controls (bottles with feed but without amoxicillin).

243 6.3.2.1 Methanogens – Acetate and H_2/CO_2

The addition of amoxicillin had an inhibitory effect on acetolactic methanogenesis (**Figure 6-6**). The rate of methanogenesis in acetate fed bottles was lower for all amoxicillin treatment conditions. The greatest difference in rate of methane production was for the 25mg/L condition which had an average of 32% of the rate of methanogenesis compared to the control. The 100mg/L amoxicillin conditions produced methane at 52% the rate of the control. However, not all of these differences were found to be statistically significant. Only the 1mg/L, 2.5mg/L, and 5mg/L were found to be statistically lower than that of their respective control (p<0.05) (Appendix C).

251 Although the presence of amoxicillin results in consistent inhibition in the rate of methane produced 252 for both runs, there was no statistically singificant relationship between the addition of amoxicillin 253 and the rate of hydrogenotrophic methanogenesis (Figure 6-6). The greatest difference appears to be 254 a result of differences between each run in which all conditions have a slower methanogenic activity 255 in run 1 than in run 2. When these values are assessed in terms of relative percentage of the control, 256 no trend is observed for amoxicillin conditions of 5mg/L and below (Figure 6-6B). Values of 257 10mg/L through 100mg/L of amoxicillin in the substrate produce a gradual reduction in the rate of 258 methanogenesis is observed. The 50mg/L and 100mg/L have a mean of 56% and 57% of the control

92

259	rate respectively.	None	of the	observed	differences	are	statistically	different	from	the	control
260	(p>0.05) (Append	ix C).									

261 6.3.2.2 Acetogens - Ethanol, Propionate, and Butyrate

There was no clear change in the methanogenic activity of ethanol fed bottles with the addition of amoxicillin. The production of methane in the ethanol fed bottles varied greatly during both in terms of cumulative methane production (**Figure 6-6A**) as well as the maximum rate values calculated (**Figure 6-6B**). Test conditions containing 2.5, 5.0, 50, and 100mg/L of amoxicillin all produced methane at an increased rate compared to the control while bottles containing 0.5, 1, 10, and 25 mg/L of amoxicillin each produced methane at a slower rate compared to the control. None of these differences were found to be statistically significant (Appendix C).

When fed butyrate as its primary substrate, the presence of 50mg/L of amoxicillin and below appeared to produce an increase in the rate of maximum methanogenic activity (**Figure 6-6**). For these test conditions, the methanogenic activity was between 134-198% of the control. However, inhibition appeared to be observed in the 100mg/L exposure condition produced methane at 56% of the control. However, the correlation between amoxicillin concentrations and methanogenic rate was not found to be statistically significant for the first run (p >0.05) though was significant for the second run (p < 0.05) (Appendix C).

276 There are strong batch effects for the propionate fed control cultures in which the control conditions 277 significantly differed from eachother (Section 6.3.1). Therefore, it is difficult to compare the rate 278 between runs (Figure 6-6). The rate of methanogenesis was inhibited for all conditions compared to 279 the control within Run 1. In the second run, the rate of methanogenesis was relatively supressed with 280 the exception of 10 mg/L of amoxicillin which had a mean relative increate in the rate of 281 methanogenesis of 124% (Figure 6-6B). The correlation between amoxicillin concentrations and 282 methanogenic rate was not found to be statistically significant for either of the runs (p > 0.05) 283 (Appendix C).

284



Figure 6-5 - Cumulative methane produced for each SMA test run and report total methane produced (ml). Each colour represents the mean total methane for that time point. Control 1 was run with amoxicillin conditions 0.5ppm, 1ppm, 2.5ppm, 5ppm conditions. Control 2 was run with amoxicillin conditions 10ppm, 25ppm, 50pp, and 100ppm.



Figure 6-6 - Changes in methanogenic activity in response to the different amoxicillin exposure. Figure A represents the calculated rates using 5 points. Both the left and right axis represent methanogenic activity (CH4 (ml)/gVSS *day) on different scales. Figure B represents the percent change from the mean control values. All bars report the mean value for the three replicates. The error bars report one standard deviation from the mean.

286

287 6.3.2.3 Acidogens – Glucose

The presence of amoxicillin had a clear inhibitory effect on the rate of methanogenesis in the glucose fed bottles. The rate of methanogenesis in Run 1 (0.05 - 2.5 mg/L) was as faster than the control values (116 - 129%). However, within Run 2, when amoxicillin levels were 25 mg/L or greater, the rate of amoxicillin was less than the control for that run (**Figure 6-6**). While the first run was not found to be statistically significant, the increase in amoxicillin was statistically correlated to a decrease in the rate of methanogenesis (Appendix C).

294 6.3.2.4 Hydrolysers – Complex Broth

The presence of amoxicillin has had a marked difference on the timing and the rate of methanogenic activity within the complex broth fed bottles (**Figure 6-6**). Like the glucose fed bottles, the bottles fed complex broth has a slight increase in the rate of methanogenesis for the low amoxicillin exposure conditions in which the 1mg/L amoxicillin test bottles produced an average of 136% of the control. However, above this point there is a steady decrease in the rate of methanogenesis until 100mg/L
which produced methane at 20% of the control. The correlation between the increase in amoxicillin
and a decrease in methane was found to be statistically significant for both runs (p<0.05) (Appendix
C).

303 6.3.3 Total Volume of Methane

304 6.3.3.1 Methanogens – Acetate and H_2/CO_2

The addition of amoxicillin inhibited the production of methane for all amoxicillin concentrations above 0.05 mg/L (Figure 6-7). The average volume of methane produced by the 1mg/L amoxicillin concentration was only 87% of that produced in the control. The total methane volume slowly decreased until 100mg/L which produced only 56% of the total methane compared to the control by the end of the study. While the correlation between the addition of amoxicillin and the production of methane was statistically significant for the first run (p<0.05) this correlation was not significant in the second run (Appendix D).

The total methane produced in the H_2/CO_2 bottles decreased compared to the control for the amoxicillin conditions of 10 mg/L and higher. While the 10mg/L condition produced an average of 92% of the control, the 100mg/L conditions only produced an average of 52% compared to the control conditions. There was no statistical correlation between the addition of amoxicillin and a change in total methane for the first run, however there was a significant correlation for the second run (p<0.05) (Appendix D).

318



Figure 6-7 - Total Methane produced at 66 hours. Figure A represents the calculated rates using 5 points. Both the left and right axis represent methane volume (ml) on different scales. Figure B represents the percent change from the mean control values. All bars report the mean value for the three replicates. The error bars report one standard deviation from the mean.

319

320 6.3.3.2 Acetogens - ethanol, propionate, and butyrate as substrate

321 There was no clear trend in the addition of amoxicillin and a change in the total volume of methane 322 for the bottles fed with ethanol or propionate (Figure 6-7). The methane produced in the ethanol fed 323 bottles was highly variable between each condition and between the replicates. None of these values 324 were statistically different from the control. As was observed in the rate of methanogenesis in the 325 propionate bottles, the changes in methane production appear to be more linked to the set than to the 326 amoxicillin concentration. While the 0.5, 1.0, and 10mg/L were each roughly the same at the 327 respective set control, the 2.5, 5.0, 25, and 100 mg/L conditions each produced less total volume than 328 the respective set control. The addition of amoxicillin was not correlated with changes in total 329 methane production for either run when ethanol was the primary feed. No statistical corelation was

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found for the first run when propionate was fed, however there was a statistical correlation in the second run (p<0.05) (Appendix D).

The addition of amoxicillin decreased the total methane produced for the butyrate fed bottles for the 25mg/L and above. While the rate of amoxicillin was observed to be similar for most conditions, amoxicillin exposure conditions of 10 mg/L and below, methane production continued after the initial 8 hours of incubation. For the higher conditions (25 through 100mg/L) methane production stopped shortly after 10 hours. This resulted in a decrease in the net volume of methane produced for these values. No statistical correlation between the addition of amoxicillin and an increase in methane was found for Run 1, though there was one found for Run 2 (p<0.05) (Appendix D).

339 6.3.3.3 Acidogens – Glucose

340 The increase in amoxicillin resulted in a decreased in the total volume of methane in the glucose fed 341 bottles. In the concentrations below 2.5mg/L, however, there was slight increase in the total methane 342 produced, each producing roughly 121% of the control volume (Figure 6-7). For conditions of 343 5mg/L and above, there was a steady decrease in the total volume produced until the 100mg/L 344 condition, which had a mean methane production of just 39% of the control values. These changes 345 coincided with a reduction in the rate and an increase in lag of methanogenic activity observed. 346 However, due to the variation in the replicates, none of these differences were statistically significant (p > 0.05) for Run 1. However, a correlation between the addition of amoxicillin and a reduction in 347 total methane was found for Run 2 (p < 0.05) (Appendix D). 348

349 6.3.3.4 Hydrolysers – Complex Broth

In addition to the delay in timing and reduction in the rate of methanogenic activity with the addition of amoxicillin, there was a reduction in the total methane produced in the complex broth fed bottles as amoxicillin concentrations increased (**Figure 6-7**). For the amoxicillin concentrations of 5mg/L and below, there was no observable difference in the total volume of methane produced. While the 10mg/L and 25mg/L conditions both proceed the same volumes of methane as the lower concentrations, when compared to the set 2 control, they were a net increase in total methane of approximately 160% for each. However, the 50mg/L and 100mg/L conditions, there was a decrease

357	in total methane matched by a decrease in the rate of methane produced. Due to the variation in the
358	replicates, none of these differences were statistically significant ($p > 0.05$) for Run 1. However, a
359	correlation between the addition of amoxicillin and a reduction in total methane was found for Run
360	2 (p < 0.05) (Appendix D).

361 **6.3.4** Amoxicillin and Timing of Methanogenesis (Lag time)

Although there was some variation in the timing of the maximum rate of methanogenesis, no lag in activity was observed for the methanogenic or acetogenic test conditions. However, there was a clear trend between the addition of amoxicillin and a lag in activity observed for the glucose and complex broth fed test conditions (**Figure 6-5 and Figure 6-8**).

When glucose served as substrate, the addition of amoxicillin at all concentrations produced a noticeable increase in the lag time before the maximum rate of methanogenic activity (**Figure 6-8**). No lag was observed in the controls. The 0.5mg/L amoxicillin conditions had an average lag time of 18 hours of incubation, while in the 100mg/L amoxicillin conditions the lag in activity increased to 38 hours of incubation. These values corresponded with a decrease in methanogenic activity above 2.5mg/L. All lag times were statistically different from the control value (p<0.02).

For complex broth fed bottles the addition of amoxicillin was also associated with increased lag time before methanogenic activity (**Figure 6-8**). No lag was observed in the controls, however by the 2.5 mg/L amoxicillin conditions the lag in activity was an average of 4.67 hours of incubation. The 100 mg/L conditions has an average lag of 15 hours. Lag times for amoxicillin test conditions over 2.5 mg/L were statistically different from the control value (p<0.02).

The increase in lag time associated with an increase in the addition of amoxicillin is likely the result of cell death associated with an increase in amoxicillin concentrations in the feed. Research by Trego et al. (2020) suggests that microbes within the anaerobic granules which are responsible for hydrolysis and fermentation are located on the outside of the granule. As such they might come into closest contact with amoxicillin. This could result in a lag as there would be fewer members of the community breaking down the available substrate. The lag in the lag associated with fermentation may be more severe than the lag associated with hydrolysis, as fermentation is conducted by enzymes

inside the cell, while hydrolysis is conducted extracellularly (Madigan, 2014). Therefore, even in



the case of the death of the cell, the enzymes performing hydrolysis may remain active.

Lag time as defined as the timing of the first point used to calculate maximum methanogenic activity rate. All bars report the mean value for the three replicates. The error bars report one standard deviation from the mean.

386

387 6.4 DISCUSSION

The primary aim of this study was to measure how acute exposure to amoxicillin impacts the activity for each of the trophic groups within an anaerobic community in terms of the rate and timing (lag time) of methanogenesis and the total volume of methane produced from specific substrates. It was established that for all feed types tested the presence of amoxicillin resulted in a measurable change in the methanogenic activity, though the sensitivity of the microbes to amoxicillin differed between test conditions.

394 All methanogens belong to the domain of archaea and are assumed to be naturally resistant to the 395 effects of beta-lactam antibiotics due to the lack of peptidoglycan in their cell walls (Kandler and 396 König, 1998). This study hypothesised therefore that the addition of amoxicillin to the substrate 397 would not inhibit the production of methane from either acetate or hydrogen / CO₂. However, 398 exposure to amoxicillin had a measurable impact on the activity of both methanogenic substrates. 399 While there was no observed change in the rate of methanogenesis for either the hydrogenotrophic 400 or the acetoclastic methanogens, there was a reduction in the total methane produced. Although the 401 amoxicillin theoretically would not have affected the methane forming archaea present, there are

Figure 6-8 – Lag time.

402 many bacterial species which are capable of utilising acetate, hydrogen, and carbon dioxide (Ferry,
403 2015; Mutungwazi, Ijoma and Matambo, 2021). It is possible that some of the substrate would have
404 been used by those organisms towards repairing structural damage resulting from amoxicillin and
405 may not have been converted into methane.

406 The presence of amoxicillin had a minimal effect on the utilisation of the provided VFAs within this 407 study. However, it should be noted that the methanogenic activity was markedly smaller for all conditions, including the control, than as would have been expected given previous studies using the 408 409 same seed community (see Method Development Chapters 4 and 5). For example, 0.1ml of 3M 410 butyrate produced an average of 11.25 ml methane at a rate of 184.7 (CH₄ (ml)/gVSS*day) in the 411 pilot study compared to an average of 0.56 ml of methane produced at a rate a maximum rate of 15.70 (CH₄ (ml)/gVSS*day) in this study. There was nearly a year between these studies. During that 412 413 time, the culture which had previously been acclimated to distillery waste as its primary substrate 414 (ethanol and other VFAs) was instead primarily fed a glucose and nutrient mixture. This could have instead acclimated the community towards sugar fermentation as well as the hydrolysis of dead 415 416 cellular matter as its primary substrates. As a result, the community would have a higher rate of 417 methanogenesis when fed glucose fermentation and complex organic hydrolysis as was observed. As 418 a result, the lack of response to the presence of amoxicillin could be a result of this lack of activity 419 rather than resistance to beta-lactam antibiotics per se.

420 The presence of amoxicillin had a clear dose/response effect on the rate, timing, and overall volume 421 of methane produced by acidogenic and hydrolytic microbes. Furthermore, they were the only feed types associated with a clear increase in lag phase as the result of the presence of amoxicillin, 422 423 particularly in the case of glucose fermentation. The fermentation of glucose can take a variety of 424 different pathways which are conducted by a variety of bacterial types, though the most common 425 products are propionate, butyrate, ethanol, acetate, formate, and CO₂(Gerardi, 2003; Madigan, 2014). 426 The full hydrolysis of complex broth into methane requires additional steps to break down the lipids 427 and proteins within the mix. Hydrolysis itself is undertaken using exoenzymes on the outer 428 membrane of the cell often, by the same bacterial types which produced the hydrolysing enzymes

429 (Mutungwazi, Ijoma and Matambo, 2021). The changes in the rate of methane production are likely
430 the result in cellular damage because of the presence of amoxicillin within the substrate. The
431 reduction in total methane produced could be the result of the substrates being utilised to build more
432 cellular materials in response to the physical damage done to the cell walls by the amoxicillin.

433 That lag phase increases due to acute exposure to amoxicillin in the treatment of complex substrates 434 is an important finding in relation to wastewater treatments in that the lag could result in temporal decreases in treatment efficiency in anaerobic systems. Whilst so, the cause of the increased lag phase 435 was not ascertained in this study. Tentatively, the lag in methanogenic activity observed for the 436 437 glucose and complex broth fed bottles could be caused by the time required for repopulation within 438 the trophic cascade. If key members were inhibited, damaged, or killed by the presence of 439 amoxicillin, this could have prevented the methane from being completely formed until that part of 440 the community could repopulate. Additionally, if the provided substrates were required for the 441 reconstruction, this could also explain the delay in activity. Furthermore, as was discussed in section 6.3.4, the members of the community responsible for fermentation and hydrolysis are physically 442 443 located on the outside of the granule where they would be most exposed to any amoxicillin within 444 the surrounding medium (Trego et al., 2020)s. Therefore, they would potentially the most sensitive 445 as they have the most contact with the amoxicillin. While fermentation and hydrolysis are often 446 conducted by the same microbes (de Lemos Chernicharo, 2015) as hydrolysis is conducted 447 extracellularly, it is possible that enzymes could still breakdown substrates after the cell died. Finally, 448 if antimicrobial resistance genes were latent in the community, this lag could be caused by the time 449 required for these genes to be activated and the enzymes required to inactivate the amoxicillin to be 450 produced. To confirm these hypotheses, future studies which track the VFA make up within the 451 effluent throughout the study could determine which point the in the trophic cascade was most 452 inhibited by the amoxicillin. The build-up of a single VFA product could indicate inhibition of the 453 microbes responsible for their degradation. Genetic sequencing of the community using next 454 generation sequencing (NGS) could indicate which members of the community are promoted or inhibited after amoxicillin exposure. qPCR targeting genes associated with amoxicillin resistance 455 456 could also be used to determine any resistance genes present. A temporal analysis throughout an

457 exposure event / SMA which analyses these hypotheses could indicate the causes and activity during458 the lag time caused by the presence of amoxicillin.

459 Despite the wide use of beta lactams in human health, and the widespread use of anaerobic 460 communities to treat sewage, there are few studies which specifically look at the effects of amoxicillin on anaerobic microbial communities. The different studies don't always agree and 461 462 variations in study design make it difficult to compare these studies directly. For example, a study on antibiotics effect on net methane production found that 10 mg/L penicillin led to a partial 463 464 inhibition of 25-45% of total methane produced (Sanz, Rodríguez and Amils, 1996). Another study 465 which assessed different antibiotics effect on nitrogen removal rate reported that anaerobic systems acclimated to amoxicillin could withstand concentrations up to 60mg/L without inhibition (Zhang et 466 al., 2015). A study by Su et al (2019) did not report any inhibition of methane in anaerobic granules 467 468 exposed to 100mg/L amoxicillin (Su et al., 2019). This study has shown that the potential sensitivity 469 of amoxicillin on the activity in anaerobic system is highly dependent on the makeup and activity of the culture being tested and the feed being used. For example, amoxicillin in a highly proteinaceous 470 471 feed could reduce the rate of breakdown, and in the case of wastewater, the treatment efficiency of 472 the sewage. However, amoxicillin appeared to have little influence on the degradation of gaseous 473 substrates. As such, this research supports the findings of the other research currently available as 474 well as providing potential insight to the variation in findings between the studies as each of these 475 studies contained different feed substrates.

476 Furthermore, there is very little research into the concentrations of pharmaceuticals in septic tanks 477 environments. Reported amoxicillin concentrations in WWTP influent range greatly. While most 478 western countries may see an average amoxicillin in influent around 13 ng/L (Zuccato, Castiglioni 479 and Fanelli, 2005) extreme cases have been reported as high as 43,980 ng/L (Peña-Guzmán et al., 480 2019). The influent into septic tanks can be might more concentrated in terms of COD as well as the 481 contaminants it contains. Conn (2010) showed that pharmaceuticals and organics in septic tanks can 482 be 1-2 orders of magnitude higher than those found in centralised water treatment facilities (Conn, 483 Siegrist, et al., 2010). Therefore, it is not unreasonable to think that concentrations as high as 5mg/L

in extreme cases may be found in septic tanks. Based on the results of this study, as well as those
conducted by others (**Table 6-2**), it is reasonable to think that an acute exposure event of amoxicillin,
as might be found on a rural septic tank, or at a medical facility, might partially inhibit the microbial
community in receiving septic tank and thus reduce the capacity of the system to process waste.
However, the scale of the inhibition will depend on the study design.

489 490

 Table 6-2 - Average inhibition comparison from different feed complexities for concentration ranges.

 Percentages are expressed as a proportion of the average of the control.

491

Feed	Measurement	Antibiotic	0.5-2.5mg/L	5-25mg/L	50-100mg/L	Source
Acetate	Total Methane (ml)	Amoxicillin	86.82%	70.28%	59.57%	This study
Ethanol	Total Methane (ml)	Amoxicillin	105.60%	87.95%	115.60%	This study
Propionate	Total Methane (ml)	Amoxicillin	114.39%	102.93%	61.32%	This study
Butyrate	Total Methane (ml)	Amoxicillin	86.10%	84.09%	78.82%	This study
Glucose	Total Methane (ml)	Amoxicillin	120.04%	81.32%	47.34%	This study
Complex Broth	Total Methane (ml)	Amoxicillin	107.23%	143.26%	69.77%	This study
Acetate, propionate, butyrate	Total Methane (ml)	Penicillin	-	-25.45	-	(Sanz, Rodríguez and Amils, 1996)
Synthetic wastewater	Nitrogen Removal	Amoxicillin	0%	0%	0%	(Zhang et al., 2015)

492

493 While SMAs are a simple and inexpensive method for determining the activity of microbes in 494 response to substrates, they do not provide an exact identification of microbes within the culture. 495 Molecular methods would provide a greater window into the specific changes in the population in 496 response to the presence of amoxicillin on a species level. However, genetic sequencing would be 497 sensitive enough to measure changes not assessed by this SMA, such as specific changes to the 498 methanogenic pathways. Furthermore, whilst resistance to beta-lactam antibiotics within hospitals 499 and wastewater treatment facilities is well documented (Laht et al., 2014; Hultman et al., 2018), at 500 the time of writing there are no studies which specifically measure the proliferation of AMR genes 501 within an anaerobic community in response to added amoxicillin. A future study which measures 502 changes in the resistance genes throughout an anaerobic community would demonstrate an additional 503 way that these communities would change in response to the presence of amoxicillin. This is 504 particularly important to understand the spread of antimicrobial resistance genes through anaerobic

water treatment technologies as contaminated effluents may flow into the water environment wherethey can pose a danger to human and environmental health.

507 Finally, this study focused exclusively on shock exposure to amoxicillin on anaerobic communities. 508 However, in many anaerobic systems, such as septic tanks treating hospital waste or natural 509 anaerobic environments, exposure would be chronic. This would allow for adaptation to the regular 510 presence of the antibiotic. As the microbial community composition and / or resistome shifts it is possible that sensitivity to amoxicillin will shift as well. Studies comparing the chronic and acute 511 512 exposure have been conducted for other pharmaceuticals such as tetracycline (Cetecioglu et al., 513 2013). Therefore, an additional study which examines long term chronic effects of amoxicillin would be useful to understand chronic as well as acute exposure. 514

515 6.5 CONCLUSION

516 It is increasingly important to understand how the antibiotics that we use impact our water treatment 517 systems. Amoxicillin is one of the most prescribed antibiotics in the world. In areas which are not 518 reliant on centralised water treatment systems, septic tanks provide most of the waste treatment though anaerobic digestion of waste and conversion of organics into methane. Nevertheless, the 519 520 effects of amoxicillin and other beta-lactams have on these small water treatment systems is rarely 521 studied. This research showed that low concentrations of amoxicillin have the potential to 522 increase methane production in these systems though at concentrations of 5mg/L -100mg/L 523 amoxicillin, the treatment efficiency of these systems drops with the severity of inhibition 524 dependent on the substrate type. Further, the impact of amoxicillin on activity is not uniform 525 across trophic groups in an anaerobic microbial community. While the presence of amoxicillin 526 appears to result in minimal inhibition in the rate or volume of methane produced by the VFA fed 527 test conditions, the methane production requiring hydrolysis or fermentation appear to be sensitive 528 to amoxicillin at all concentrations tested. Additionally, the presence of amoxicillin resulted in an 529 increase in the lag time for maximum methanogenic activity within the hydrolytic and glucose 530 degradation pathways. This study shows that there is a measurable effect of amoxicillin in the range 531 of concentrations that might be found in septic tanks. Given that decreased treatment rates and

- 532 increased lag time prior to commencement of treatment of complex wastes could result in decreased
- treatment efficiency in septic systems, this finding is important therefore more research is needed to
- better understand this relationship for both acute effects on a longer time scale.

² Chapter 7

³ Changes In Community

- Structure and AMR Genes in
 Anaerobic Granules After
- 6 Acute Amoxicillin Exposure
- 7 8

1

9 7.1 INTRODUCTION

10 Antibiotic resistance in the wastewater sector is an issue of global importance. Long term chronic 11 exposure of highly active microbial communities to sub-lethal concentrations of a variety of 12 antibiotics, as are found in wastewater treatment plants (WWTPs), generates ideal conditions for the 13 development and proliferation of antimicrobial resistance genes (Rodriguez-Mozaz et al., 2015; Wu 14 et al., 2016; Wang et al., 2020). Centralised wastewater treatment facilities have been long observed 15 to be point sources of the spread of antimicrobial resistance genes into the water environment (Kümmerer, 2009b). As such there are many studies exploring antimicrobial resistance wastewater 16 17 microbial communities in long term bench scale studies (Aydin, Ince, et al., 2015; Aydin, Ince and Ince, 2015a). 18

However, much of the world is reliant on decentralised wastewater treatment (Libralato, Volpi
Ghirardini and Avezzù, 2012). As influent into these systems is more sporadic and concentrated than
those entering centralised WWTPs, so too is the antibiotic within those influents resulting in acute
rather than chronic antibiotic exposure. Whilst there are studies which focus on how acute exposure
of antibiotics within an anaerobic microbial community changes activity (Cetecioglu *et al.*, 2012;
Ozbayram *et al.*, 2015), there are very few which study the effect of beta lactams. Chapter 6 of this

25 thesis demonstrated that acute exposure to amoxicillin can result in immediate changes to microbial activity in terms of the methane production rate and lag time before which substrates were 26 27 metabolised into biogas and methane. A reduction in activity implies the inhibition in activity and or 28 outright mortality of some members of the community. Amoxicillin is a broad range antibiotic which inhibits the generation of penicillin binding protein (PBP) found within the cell walls of prokaryotes 29 30 (Macheboeuf et al., 2006; Madigan, 2014) though susceptibility to amoxicillin varies from species 31 to species (Gartiser et al., 2007; Shi, Leong and Ng, 2017). Furthermore, archaea, which don't 32 contain any PBP in their cell walls, are generally resistant to all beta-lactams. Anaerobic microbial communities, such as those found in anaerobic granules, are highly syntrophic whereby the food for 33 some members of the community are produced (as metabolites) by other members within the trophic 34 35 cascade. A reduction in numbers or metabolic rate of one part of the community, therefore, could result in a decline in activity and robustness of the community as a whole. 36

37 The presence of sublethal concentrations of antibiotics has shown to correlate to an increase in the 38 presence of antimicrobial resistance genes within the anaerobic communities. For example, an 39 increase in tetracycline resistance gene counts was observed in anaerobic microbial communities 40 when exposed to erythromycin, tetracycline and sulfamethoxazole (Aydin, Ince and Ince, 2015b). 41 Similar studies focusing on the development of antibiotic resistance to beta-lactam antibiotics in 42 anaerobic microbial communities are not available. Nevertheless, medical research, which focuses 43 on single strains of anaerobic bacteria have demonstrated beta-lactam resistance is present in many 44 species. For example anaerobic Bacteroides, Fusobacterium, and Clostridium species have long been shown to produce beta-lactamase enzymes (Nord and Hedberg, 1990). 45

This study aimed to expand upon the work conducted in Chapter 6 of this thesis. As the cultures fed glucose in that study appeared to be the most sensitive to the presence of amoxicillin, glucose was used as the primary substrate in this study. The previous work established that increasing amoxicillin exposure inhibited methanogenesis within anaerobic granular microbial communities. The hypotheses tested are summarised in **Table 7-1**.

51
52 Table 7-1 - Hypotheses

		Hypotheses
1.	Microbial activity	 Increasing amoxicillin would inhibit methanogenic activity. And hence the sCOD removal within the microbial communities would decrease with an increase in amoxicillin exposure.
2.	Community structure	 There will be a mixed microbial consortia for AD within the granules. The presence of amoxicillin will drive a change in microbial community composition which is distinct to that observed in the control (no-amoxicillin group). The changes in the community will have a resulting reduction in the overall robustness of the community as a whole and its ability to tolerate more change.
3.	Antibiotic resistance.	 Anti-microbial resistance genes would be present in the seed community at the offset. Exposure to amoxicillin will increase the relative abundance of beta-lactamase and other resistance associated genes.

53

54 **7.2 METHODS**

55 7.2.1 Experimental design

This study aimed to monitor changes in community structure and the proliferation of resistance genes 56 within granular anaerobic microbial communities exposed to beta-lactam antibiotics. The analytical 57 58 methods summarised in this section are explained in further detail below (Figure 7-1 and 7-2). To test how the microbial communities within anaerobic granules changed in response to an acute 59 60 exposure to amoxicillin, sub-samples of a common seed sludge were exposed to three different concentrations of amoxicillin (1mg/L, 10 mg/L, and 50 mg/L). The sub-samples were incubated 61 62 using glucose as substrate along with fed control (no amoxicillin) and unfed control sub-samples and 63 were monitored over 48 hours during which time a series of sacrificial bottles were removed for analysis to capture temporal changes (Figure 7-3). All test conditions were conducted in triplicate. 64 65 The amoxicillin concentrations selected for this study were informed by the results of Chapter 6 in 66 which measurable suppression of methanogenic activity as well variations in the lag in the start of 67 activity were observed when amoxicillin in the substrate was at 5 mg/L and greater. Conversely the inhibitory effects of 1 mg/L of amoxicillin on the timing and rate of amoxicillin were not statistically 68

69 significant. As it is possible that there may have been microbial changes in the resistome that were independent of methanogenesis, this 1 mg/L value was also included in this study. As the previous 70 71 study demonstrated the inhibitory effect was greatest for the cultures fed exclusively on glucose, the 72 cultures within this study were also fed glucose with the aim of replicating the results. The pressure 73 in the head space of the sealed serum vials in which the sludge sub-samples were incubated were 74 monitored every 2-6 hours throughout the study and headspace gas was extracted from the bottles at 75 the final time point to determine methane production for each test condition. To obtain temporal data 76 within the study, seven sets of three sacrificial bottles were prepared for each test conditions (i.e. 3 77 sacrificial bottles for each test condition at each of 7 time points). Bottles were removed (sacrificed) 78 for times correlating to Figure 7-3. For each set of sacrificial bottles sampled, the effluent and 79 granules were immediately removed and stored for later analysis. Microbial community composition 80 was determined using next generation sequencing of the V4 region of the 16s rRNA gene using the 81 Illumina Miseq sequencing platform (San Diego, CA). The SmartChip Real-Time PCR system (Takara Bio Inc., Shiga, Japan) was used to detect and quantify the AMR genes present. 82

83





Test Conditions



The acclimation was conducted over 14 days with a protein rich broth and a trace metal nutrient added to the substrate solution. The test conditions contained three different amoxicillin feed types, as well as a control, with glucose and no amoxicillin and a blank with no glucose and no amoxicillin. All were run in triplicate.



Figure 7-2 Sub sampling and analyses for each sample.

The biogas, effluent, and granules were each sampled and stored separately. The biogas was analysed immediately for methane quantification. The effluent was stored at -4C before sCOD quantification. The granules were flash frozen in the liquid nitrogen immediately after removal and stored at -80C before DNA extraction for next generation sequencing and AMR gene quantification using qPCR array.





Seven sub-sample time points at which triplicate bottles of each test condition were removed from the study obtain information about temporal variation in the effluent and microbial community. Triplicate samples were taken for each set of amoxicillin test conditions as well as a blank and a control for a total of 15 bottles removed at each sacrificial point.

88 7.2.2 Anaerobic culturing and Sampling

89 Anaerobic culturing was conducted as described in Chapter 3 section 3.1 and is summarised here. A 90 GMA was conducted using glucose as the primary substrate. One millilitre of a 1M glucose solution 91 was added to 0.5g of anaerobic granules and 9.5ml anaerobic medium containing nutrient additives, 92 for a total of 10ml. The study tested three amoxicillin concentrations with an in-bottle concentration 93 of 1 mg/L, 10 mg/L, and 50 mg/L. A total of five test conditions were run as described above. 94 Pressure readings in the headspace of each bottle were measured every 2 hours for a total duration 95 of 48 hours. The methane concentration of the headspace gas was measured at the end of the 96 experiment using a GC-FID and the rate of methane production was calculated (as described in 97 Methods Chapter Section 3.3). As the provided glucose was often consumed within 6 hours, a 4-98 point window was used to calculate maximum activity rate. Upon removal of the sacrificial bottles, 99 the bottles were opened. The effluent was decanted and filtered through a 0.2um Whatman filter and 100 transferred to a sterile 15ml tube and stored at -20C until sCOD quantification as described in Section 101 0. Anaerobic granules were transferred to 2ml Biobanking and Cell Culture Cryogenic Tubes and 102 flash frozen using liquid nitrogen before storage at -80C until DNA extraction as described in Section 103 7.2.4.1.

104 7.2.3 Measuring sCOD

The soluble COD (mg/l) of the filtered effluent was measured in accordance with Standard Methods ISO 15705 using Hach LCK 400, (0-1000 mg/L O₂) and LCK 014 (1,000-10,000 mg/L O₂) kits and quantified using a Hach DR 2800 spectrophotometer. The LCK 014 kits were initially used, however when supplies ran out, samples were diluted to fall within the range of detection for the LCK 400 kits. The theoretical sCOD of the media at T0 was calculated from theoretical sCOD for the glucose (**Equation 1**) was combined with the COD measurements of the anaerobic medium mixed and nutrient mixture.

$$1920 \frac{mg}{l} tCOD \ Glucose = \frac{6 * 32 \frac{g}{mol} \ O2 * 1800 \frac{mg}{L} \ Glucose}{180 \frac{g}{mol} \ Glucose} \qquad Equation 2$$

112

113 7.2.4 Molecular Microbiology Methods

114 7.2.4.1 DNA Extraction

115 Anaerobic granules were flash frozen in liquid nitrogen at the time of sampling and stored at -80°C. 116 The total genomic DNA was extracted from the stored samples at the end of the experiment. Samples 117 were placed on ice and allowed to thaw for 10 min before processing. Care was taken to ensure that 118 each sample only experienced one freeze thaw cycle as part of the extraction process to minimise 119 potential DNA degradation. Once thawed, each sample was homogenised within the sampling tube 120 using a sterilised glass rod to ensure representation of the entire community in the extracted DNA. 121 DNA was extracted from 0.3g of the homogenised samples using the Fast DNA spin kit for soil (MP Biomedical) and Fast Prep -24 instrument. 122

123 7.2.4.2 Preparation of barcoded amplicons for NGS

To provide insight to microbial community composition and dynamics during the study, the DNA 124 125 extracted from the anaerobic granules was prepared for next generation sequencing (NGS). Prior to 126 PCR preparation, the extracted DNA was thawed and quantified using the Qubit dsDNA BR and 127 normalised 1ng/ul using PCR grade water (Qiagen Nuclease free water Cat no. 129115) and stored at -80°C until PCR amplification. Amplicons were prepared by PCR using barcoded primers 128 targeting the V4/V5 region of the 16s rRNA gene (F515 (Caporaso et al., 2011) and R926 (Walters 129 130 et al., 2016)) using Golay barcode primers which are each synthesized with a unique barcode gene 131 sequence and Illumina sequencing adapter (Caporaso et al., 2012). PCR amplification was conducted 132 using the KAPA HiFi HotStart Ready Mix Kit (KAPA Biosystems) with the PCR conditions and 133 reactions prepared as described in Table 7-2 and 7-3 respectively. Triplicate 25ul reactions were 134 used for each sample (for a total of 75ul per sample) to maximise amount of final PCR product while minimising PCR error. Triplicate reactions were pooled and the PCR product was purified using the 135 136 Zymoclean Gel DNA Recovery Kit (Zymo Research) according to the manufacturer's specification. 137 When the yield of the PCR production was below 10ng/ul for the combined triplicated PCRs, 138 the PCR would be repeated, pooled and purified. The mean library length was 411 bp.

140 *Table 7-2 - The PCR temperature and time conditions.*

Phase	Time (min)	Temp (c°)	Repeat
Initial Denaturation	5:00	95	
Denaturation	0:20	98	
Annealing	0:30	62	X 25
Extension	0:40	72	
Final Extension	1:00	72	

141

142 *Table 7-3 - Volume of the PCR volumes used.*

Reagent	Optimisation Volumes (ul)	Test Volume (ul)
PCR Grade Water	6.6	11
Hifi taq	3	5
dNTPs	0.45	0.75
Reverse Primer (926R) 0.3uM	0.45	0.75
Taq	0.3	0.5
Forward Primer (F515.X) 0.3uM	0.45	0.75
Template / NTC	3.75	6.25
Total	15	25

148

The purified PCR product for each sample was then normalised to 10ng/ul DNA in nuclease free water. An aliquot of 2ul was removed from each normalised sample and pooled together in a 2ml tube to create a single sample for sequencing. Pooled samples were vortexed and quantified using the Qubit dsDNA HS (high sensitivity) kit. Purity was assessed visually using a 2% agarose gel (as described above). The pooled and multiplexed library was processed by the Earlham Institute (EI) (Norwich, UK) and 300bp paired-end reads for each sample was obtained on an Illumina platform.

155 7.2.4.3 Selection and Screening of AMR gene targets

The quantification of antimicrobial resistant genes was conducted using the SmartChip Real-Time PCR system (TakaraBio, CA, USA) by Resistomap Oy (Helsinki, Finland). This system contains 5184 x 100nl reaction wells which allow for 14 different sample and assay configurations allowing for up to 384 samples or assays to be run simultaneously on a single chip. To determine the selection of gene targets used, the quantification of antimicrobial resistance genes was conducted in two phases. First, an initial screening chip was prepared which surveyed five samples subset from the total sample set for a wide variety of AMR gene targets to establish which resistance genes were

present in the exposed as well as unexposed populations. The screening chip analysed five samples
for 248 possible gene targets using primers previously validated by Muziasari *et al.*, (2016, 2017).
The five samples screened were:

• Two samples from T0 to establish which genes were present in the seed population.

A single sample from time point T1 which, had been exposed to glucose but no amoxicillin,
was chosen to represent conditions at the beginning of the experiment.

• Two samples were chosen from T7 which were exposed to glucose and 50 mg/L of amoxicillin. These samples were chosen as they contain the highest amoxicillin concentrations as well as longest exposure time within this study. It was hypothesised that these samples would have the greatest relative increase in amoxicillin resistance within the study.

174 The screening chip contained gene targets from across 6 classes: 16s rRNA (2 targets), betalactamases (106 targets), integrons (8 targets), multidrug resistance genes (MDR) (73 targets), mobile 175 176 genetic elements (MGE) 53 targets, and tetracycline (6 targets). A detailed list of the gene targets 177 including their forward and reverse primers can be found in Appendix A and B. Beta-lactamases and 178 MDR class targets were both chosen to track changes in genetic resistance to beta-lactam antibiotics 179 in response to an acute exposure to amoxicillin. Integrons and MGEs concentrations have been 180 shown to correlate to an increase the count of resistance genes elsewhere and were chosen to reflect 181 this relationship (Subirats et al., 2019). As no tetracycline was added to any of the cultures, genes 182 associated with tetracycline resistant genes would not be expected to increase during this experiment. 183 As such genes associated with tetracycline were included to measure genes without selection 184 pressure.

The results produced by the screening survey were used to narrow the scope of gene targets. The relative abundance of genes detected were averaged for all samples within the screening chip to generate an average relative abundance (ARA) score for each gene. The 36 most abundant genes (i.e. those with targets with the highest ARA scores, **Table 7-4**) were then used as targets for quantification in the remaining 96 samples from the study.

Beta Lactam	Tetracycline	Integrons	MGE	MDR	191 16S rRNA
cfxA	tetQ	intl3	IS6100	mexF	16S rRNA
fox5		intI1_2	ISEcp1	oprJ	16S rRNA2
blaSFO		intI1_1	tnpA_2	mepA	
blaTMB			IS1247_1	oprD	193
blaAIM			tnpA_5	tolC_1	
blaSHV_1			repA	acrA_5	194
penA			tnpA_3	acrR_3	
blaACT			trbC	arsA	195
blaMIR			orf37-IS26		155
blaOXA48_2			IS1111		100
			IS630		196
			Tn5403		

190 <i>Table 7-4</i> - <i>The final 36 genes chosen for the secon</i>	id screening
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198 7.2.4.4 Smart Chip qPCR Array

The extracted genomic DNA was quantified using the Qubit dsDNA BR kit and sample concentration 199 200 was normalised by dilution to produce 100ul aliquots of 10ng/ul final concentration using nuclease 201 free water before shipping to Resistomap for processing. PCR conditions were as described in 202 previous publications (Muziasari et al., 2016, 2017). The reaction was conducted in the SmartChip[™] 203 by the SmartChip[™] Multisample Nano dispenser (TakaraBio, CA, USA). Each 100nl reaction was 204 comprised of SmartChip TB Green Gene Expression Master Mix (TakaraBio, CA, USA), nuclease 205 free PCR- grade water, 300 nM of each primer and a DNA template at 2 ng/ μ L (Muziasari *et al.*, 206 2016, 2017). The cycling conditions were conducted as described in Wang et al., (2014). Melting 207 curve analysis was performed for each primer set for all the samples. Melting curve analysis was 208 processed using the SmartChip[™] qPCR software. Amplicons with unspecific melting curves and 209 multiple peaks based on the slope of melting profiles were discarded from the analysis as they were 210 considered to be false positive data. The threshold cycle (CT) of 27 was set as the detection limit 211 (Muziasari et al., 2016, 2017). In addition to the experimental triplicates, three technical replicates 212 were analysed for each qPCR reaction. The genes were reported as total counts as well as relative 213 gene abundance. Relative gene abundance was calculated in proportion to the 16s rRNA gene in 214 which relative gene abundance = $2-\Delta CT$ where $\Delta CT = \Delta CT$ (detected gene) - ΔCT (16S rRNA)

(Muziasari *et al.*, 2016, 2017). Data processing and data analysis were performed using python
program by Resistomap Oy (Helsinki, Finland).

217 7.2.5 Bioinformatics and Statistical Methods

- 218 All statistical analyses were performed using R statistical package and RStudio 4.1. All analyses
- 219 were performed under the supervision and guidance of Dr. Umer Ijaz.

220 7.2.5.1 Sequencing Analysis

221 Abundance tables were generated by constructing amplicon sequencing variants (ASVs) using the 222 Oiime2 DADA2 denoising algorithm workflow with (Kozich al., 2013) et 223 (https://github.com/umerijaz/tutorials/blob/master/qiime2_tutorial.md). ASV's were taxonomically 224 classified using SILVA SSU database v138. Additionally, Qiime2 was used to generate a rooted phylogenetic tree of the observed ASVs and a final BIOM file was generated which combined 225 226 abundance information with taxonomy. This produced a 100 (samples) X 1,518 (ASVs) abundance table. Furthermore, as a prefiltering step, we removed typical contaminants such as those matching 227 228 Chloroplast and Mitochondria that unassigned and those are 229 (https://docs.qiime2.org/2022.2/tutorials/filtering/) and dropping samples with less than 5000 230 cumulative read counts, resulted in a 97 (samples) x 1,491 (ASVs) abundance table on which we 231 performed statistical analyses. The summary statistics of reads per samples is as follows: [Minimum: 232 11,568; 1st Quartile: 59,012; Median: 74,391; Mean: 84,010; 3rd Quartile: 92,976; Maximum: 233 790,867].All further statistical analysis was carried out in R Studio (v4.0.0) using this data along 234 with metadata associated with the study.

235 7.2.5.2 Methanogenic Activity and sCOD Removal

Differences in activity (methanogenic activity rate, lag time, and total biogas production volume) were assessed using a Student's T-Test which compared each amoxicillin treatment condition with the control. Soluble COD removal for each time point was assessed using a linear model which compared the sCOD value with the time of incubation associated with that sub sample.

240 7.2.5.3 Microbial community Analysis

Microbial community analysis was conducted using the 'Vegan' package within R (Oksanen et al. 2013). The standard alpha diversity indices, such as Shannon entropy, Richness, Fishers Alpha, and Pielous Evenness were calculated using the standard approach given in the Vegan package. Note that these samples were rarefied to minimum sample size. This is particularly useful for the richness estimate which gives the estimated species based on the minimum samples size. Differences in alpha diversity measurements between time points (or test conditions) were assessed using an ANOVA (Vegan's aov() script) to calculate pairwise ANOVA P-values.

Community composition variation between treatments and incubation time (Beta diversity) were assessed using pairwise Permutational Multivariate Analysis of Variance (PERMANOVA).
Principal Coordinate Analysis (PCoA) plots were used to visualise variances in dissimilarity based on Weighted UniFrac distance matrices using different distance measures (Vegan's capscale() function). The data set was grouped in terms of time points as well as amoxicillin conditions and ellipses were drawn using Vegan's ordiellipse() function that represent the 95% confidence interval of the standard errors.

The core microbiome of the total community was calculated utilising all samples using R's microbiome package as described in (McKenna *et al.*, 2020) and defined as taxa which are prevalent in 85% of samples. Analysis was performed for relative abundance at genus level.

258 7.2.5.4 Predicted taxa-function robustness

259 The taxa-function robustness proposed by Eng and Borenstein (2018) is a measure of the microbial 260 communities structure and metabolic pathways. This was used to estimate microbial community resilience and stability with the aim of quantifying an inherent component of the structure-function 261 262 relationship between taxonomic and functional profiles using community composition and of the 263 distribution of genes across genomes (Eng and Borenstein, 2018). For each sample, a perturbation 264 model (100 perturbations / sample) was applied to produce artificial microbial communities within 265 the vicinity of original microbial communities observed in each sample. The taxonomic shift of these 266 artificial perturbations were then compared to original community using phylogeny-aware weighted

267	UniFrac dissimilarity metric. For each perturbation, we also compared its functional shift from the
268	original community using cosine dissimilarity of the predicted functional profiles. To obtain
269	predicted functional profiles, the ASVs were taxonomically assigned using Greengenes database
270	(gg_13_5) for which functional KEGG orthologs and their corresponding copy numbers were
271	available. Next, the relationship between taxonomic perturbation magnitude and functional profile
272	shift for each sample (after 100 perturbations) were fitted using the linear regression model on natural
273	log-transformed data: $\ln(f) = -a + b \ln(t)$, where f is the functional shift obtained from cosine
274	dissimilarity, and t is the taxonomic shift obtained from the weighted UniFrac dissimilarity metric.
275	The two unknown coefficients: attenuation a , and buffering b are the community level robustness
276	proxies which then give the estimate of functional robustness. Two measurements are used:
277	attenuation and buffering. Attenuation is described as the expected rate at which increases in the
278	taxonomic perturbation magnitude are expected to increase functional shifts. Buffering describes
279	how large a perturbation must be before a functional profile shift becomes apparent. The attenuation
280	parameter in particular is important which is based on the slope of the response curve, with a larger
281	value suggesting that taxonomic perturbations have little impact on functional shifts. It should be
282	noted that there are practical and theoretical limitations to functional prediction methodologies.
283	Primarily, functional assignments are for <i>potential</i> function not the <i>actual</i> function as not all these
284	relationships have been validated. As our ability to sequence ever more genomes outpaces our ability
285	to verify functions practically within the laboratory, the direct relationship between sequence and
286	function has only been partially assessed and our confidence in functional predictions relied on the
287	robustness and validity of our databases (Devos and Valencia, 2000). Nevertheless, for this study we
288	are perturbing the sampling space to see how much relative shift occurs in terms of function. Because
289	we are focusing on the overall shift at community level in which the comparison is in relative terms,
290	even if a few taxa are not well resolved, the results should not be meaningfully affected. Similar
291	applications of taxa function robustness and have been recently successfully published (Nikolova et
292	al., 2021). The method uses Greengenes database (n=203,452) to resolve function instead of the
293	recent version of SILVA MOD (v 138 has n=436,680). This decision by the Eng and Borenstein
294	(2018) to use Greengenes could be the result of using 16S rRNA copy numbers (a requirement for

the algorithm for normalisation purposes), which are either easy to calculate for Greengenes, or arereadily available.

297 7.2.5.5 Taxa differential analysis

298 8 To find ASVs that are significantly different between multiple conditions, we used 299 DESeqDataSetFromMatrix() function from DESeq2 (Love, Huber and Anders, 2014) package 300 with the adjusted p-value significance cut-off of 0.05 and log2 fold change cut-off of 2. This 301 function uses negative binomial general linear model (GLM) to obtain maximum likelihood 302 estimates for ASVs log fold change between two conditions. Then Bayesian shrinkage is applied 303 to obtain shrunken log fold changes subsequently employing the Wald test for obtaining 304 significances. Although relative abundance between sample groups is widely used when 305 investigating microbiomes, there is little consensus on best practices for differential analysis 306 methods (Calgaro et al., 2020; Nearing et al., 2022). Every method has analytical biases and 307 underlying assumption. For example, rarefaction can introduce false positives, it is commonly 308 used because it simplifies analyses with various in read depth. McMurdie and Holmes, (2014) 309 argue that both proportions and rarefied counts results in a high rate of false positives. Although 310 functions such as metagenomeSeq can performs well then there are a high number of replicates, 311 they can still tend towards high false positive rates. Therefore, the paper advocates that 312 investigators avoid rarefying altogether. Nevertheless, analysis conducted by Calgaro et al., 313 (2020) which compared differences between commonly available differential analysis 314 techniques has suggested that limma voom, corncob, and DESeq2 (used in this study) had the 315 best overall performance.

316

8.1.1.1 Environmental Fitting

To see if changes in covariates (AMR qPCR genes relative abundance) has an impact on microbial community structure, we fitted smooth surfaces of the covariates on an ordination plots (PCoA in this case) using penalised splines by employing ordisurf() function from R's Vegan package (Oksanen, 2018). The method uses a generalised additive model by regressing the covariate as C ~ S(Dim1,Dim2), where C is the covariate and Dim1 and Dim2 are the ordination scores extracted

322	from PCoA and S() is a spline function. We have only shown those covariates where the model fits
323	i.e., p < 0.05.

324 8.1.1.2 Antimicrobial Resistance

To determine if the relative abundance of AMR genes in each sample correlated to incubation time, a general linear model was run correlating the relative abundance of each resistance or associated genes with incubation time. This was run separately for each amoxicillin conditions. To determine if there was a linear relationship between the resistance genes and the mobile genetic elements a GLM was run correlating the relative abundance of the resistance genes with the mobile elements and time. Samples were reported to a p-value threshold of <0.05.

331 **8.2 RESULTS**

332 8.2.1 Methanogenic Activity and sCOD Removal

333 Methane production proceeded to approximate steady state for all conditions within the 48-hour window of the study (Figure 7-4). Methane was generated at an average rate of 71.4 (SD 24.5), 56.3 334 (SD 15.2), 89.7 (SD 14.7), and 83.6 (SD 14.5) for the control, 1 mg/l, 10 mg/L and 50 mg/L 335 336 amoxicillin respectively (Table 7-4). Based on the results from Chapter 3, it was assumed that the 337 presence of amoxicillin would result in measurable changes to the methanogenic activity. However, 338 no clear relationship between the addition of amoxicillin and changes in the rate of methanogenic 339 activity were observed for the different amoxicillin concentrations provided in this study. T-tests 340 comparing the rate of methane production in each test condition against the control were not found 341 to be significant (P > 0.05) (Figure 7-5) for any of the amoxicillin concentrations used. However, 342 differences in the time of maximum activity was found to be statistically different from the control 343 value for each of the amoxicillin test conditions though there was no relationship between the 344 increase in amoxicillin and the time of maximum methanogenic activity (Figure 7-5). These results 345 differ from similar analysis as reported in Chapter 6 in which the presence of an increase in amoxicillin resulted in a decrease in methanogenic activity of the acidogenic microbial subpopulation 346 347 as well as an increase in the time between the addition of substrate and the generation of methane. Differences in the results between these studies could be the result of community drift due to 348 349 differences in culture acclimatization. While each study utilised granules from the same source, the 350 granules within this study were fed and incubated with a nutrient rich broth for two weeks before the beginning of the study. Additionally, the sludge granules had been stored with intermittent feeding 351 352 between the two studies which could also have contributed to community drift.



Figure 7-3 - COD and methane production throughout the study. The cumulative methane reported has been blank adjusted. All points report the mean value for the six replicates. The error bars report one standard deviation from the mean.



Figure 7-2 - Boxplots comparing the methanogenic activity and the first point in the activity. No statistical differences were found in the methanogenic activity. However, statistical differences were found in the first point of activity. (*** $p \le 0.001$, ** 0.001 , * <math>0.01 .) The x represents the meanvalue for the three replicates. The error bars report one standard deviation from the mean.

 Table 7-5 - The mean and standard deviation for the rate of methanogenesis and the timing of the first activity of methane production.

<i>P</i> values are for t-tests comparing each of amoxicillin test conditions to the control.					
		Control	1 mg/L	10 mg/L	50 mg/L
	п	6	6	6	6
Rate	Mean	71.39	56.26	89.66	83.58
(CH4 ml	SD	24.50	15.17	14.70	14.53
/gVSS*Day)	р		0.273	0.190	0.366
First Astivity	Mean	31.00	28.00	18.76	23.33
(hours)	SD	1.00	1.15	5.18	2.75
(liours)	p		0.001	0.003	0.001

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355 356

358 The sCOD of the effluent was quantified for each of the sacrificial vials during the study to generate 359 a time series. A t-test comparing the final sCOD values recorded for each test condition found that each of the amoxicillin exposed cultures were not statistically different from the control (p-value > 360 361 0.05) (Table 7-5). Although there was an overall drop in sCOD in the effluent from T0 to T6, the sCOD values throughout the study fluctuated over time (Figure 7-4). Broadly, effluent sCOD 362 363 occurred in three phases: initial drop in sCOD, followed by an increase, after which the effluent 364 sCOD began to fall again. The timing, rapidity, and scale of these fluctuations were different for each 365 of the feed conditions. Fluctuation in sCOD values during the observational period could be the result 366 of cell death due to the presence of amoxicillin and non-optimal substrate usage in the presence of 367 amoxicillin. As the amoxicillin was used up and the community recovered, the available substrate 368 was utilised and the overall sCOD levels converged. The initial drop in sCOD was slower for each 369 of the test conditions and even increased initially for the 50 mg/L conditions, suggesting that the 370 presence of amoxicillin inhibited glucose uptake. The sCOD of the control conditions increased 371 slowly after 18 hours and began to fall again after 30 hours. Conversely the sCOD in 10 and 50 mg/L conditions both rose and fell much faster than the second and third phases than the control conditions. 372

373 374 375

 Table 7-6 - Table comparing the mean and standard deviation of the total sCOD at T7 (48 hour).

 The p-value reports the difference from the test condition from the control.

	G - 0	G - 1	G - 10	G - 50
	mg/L	mg/L	mg/L	mg/L
Mean	2107.67	1786.00	2205.00	1814.00
SD	547.45	610.57	348.53	405.25
p-value		0.61	0.84	0.58



Figure 7-4 - Differences in the sCOD. The overall sCOD contains all the values within the study while the T7 sCOD refers only to the values at the end of the study.

377 8.2.2 Microbial Community Composition

The microbial community make up was revealed using next generation sequencing of the 16s rRNA genes targeting the V4 region. Three primary questions were asked of the data set: 1) Does the presence of amoxicillin significantly change the community structure compared to the seed community? 2) Who is present in the community? 3) Which microbes are changing in response to the presence of amoxicillin?

383 8.2.3 Community Diversity

The alpha diversity indices: Pielous evenness, richness, and Shannon diversity, were calculated for 384 385 each of the amoxicillin test conditions. This calculation included all time points for each amoxicillin 386 condition. Although it was hypothesised that the presence of amoxicillin would reduce the diversity 387 when compared to the control and seed community, this was not found to be the case. Conversely, 388 statistical differences were detected amongst the richness and evenness metrics. In both cases, the 389 greatest statistical difference from the seed community was the control condition for which a drop in 390 richness and diversity and an increase in evenness was observed (Figure 7-7). While none of the test 391 conditions are statistically different from the initial community at seed, each of the test microbial

392 communities are statistically different from the control condition (glucose fed bottles without393 amoxicillin).

The microbial composition between groups was compared by plotting samples on Principle coordinate analysis (PCoA) using weighted UniFrac distance measure. When grouped by amoxicillin and feed conditions, the control (glucose only, no amoxicillin) samples clustered distinctly from the seed community. By contrast, both the amoxicillin fed conditions and the unfed blank conditions clustered close to the seed community with the amoxicillin conditions all clustering together. There



Figure 7-6 - Alpha diversity figures.

The line in the box plots represents the mean of the values. The boxes are for the first and third quartile (Q1 and Q3) of the data. The lines at the top represent statistically significant differences between each run. (*** $p \le 0.001$, ** 0.001 < $p \le 0.01$, * 0.01< $p \le 0.05$.). Amoxicillin concentrations are in mg/L.





The principal coordinate analyses (PCoA) were generated using weighted UniFrac similarities. The ellipses represent 95% confidence interval of the standard errors.

399 did not appear to be any gradation of effect with concentration of amoxicillin exposure (Figure 7-8). Both this result and the results of the alpha diversity indices were unexpected. It was assumed 400 401 that the presence of amoxicillin would drive changes in microbial community composition. That the opposite trend was observed suggests that the presence of glucose is the primary driver of changes 402 403 in community composition, whilst the presence of amoxicillin inhibits adaptation. When the samples 404 on the PCoA are grouped by time, a clear progression is observed showing that adaptation through 405 time progresses stepwise from the seed community. PERMANOVA analysis found that both 406 amoxicillin condition and time combined explained nearly 50% of the variation in the data set and 407 were both statistically significant (p < 0.001) (Figure 7-8). It should be noted that the variation on 408 beta diversity Figure 7-8, Dim 1 on the x-axis and Dim 2 on the y-axis represents the variability as 409 explained within the first two dimensions. The PCOA does not reduce the dimensions of multivariate 410 data sets as is done in other methods such as non-metric distance scaling (NMDS). However, this 411 procedure transforms the data in such a way that the majority of the variability lies in the first few 412 dimensions. This simplified visualisation in the 2D/3D space by ignoring other dimensions. 413 Therefore, in the 2D plot, the percentage values should add up to the variability explained by these 414 two dimensions. The remining variability (50.4%) are explained by the dimensions not visualized.

415 8.2.4 Community make-up

416 The microbial community composition was largely stable and did not vary greatly between test 417 conditions such that the top 25 families consistently comprised nearly 80-90% of taxa present (Table 418 7-6). Core microbiome analysis (Figure 7-10), which reports the families present in at least 85% of 419 all samples found that all samples contained the same three most abundant families: Spirochaetaceae, 420 Aminicenantales, and Bacteroidetes vadinHA17 which were each found at roughly 18-20% relative 421 abundance. Other dominant families, which appeared in all samples, though to varying degrees of 422 dominance, were: Dysgonomonadaceae, Syntrophobacteraceae, *Methanobacteriaceae*, 423 Synergistaceae, Clostridiaceae, Caldatribacteriaceae, and Kosmotogaceae. The putative function of 424 these families is described in **Table**. The dominant archaeal species observed for all samples where 425 the Bathyarchaeia, Methanobacterium, Methanofastidiosales, Methanosaeta, and 426 Methanomethylovorans.

427 The alpha and diversity metrics (Section 8.2.3) established that while the change in the structure of the community was statistically significant, this difference was subtle. To determine which species 428 429 were most closely associated with the changes in the community, a taxa differential analysis was run 430 in which statistically significant changes in species counts were calculated from pairwise comparisons of each test condition and the seed community (Figure 7-11). Species which had at 431 432 least a 2- log fold change between conditions and adjusted p value < 0.05 were deemed significant. 433 Two species were found to increase in relative abundance in the seed community compared to the 434 seed community. Clostridium sensu stricto 1, which is responsible for the generation of H_2 and CO_2 435 from formate, increased in relative abundance for all samples, and *Petrimonas*, a hydrolytic member 436 of the Dysgonomonadaceae family capable of degrading recalcitrant polysaccharides to liberate 437 oligosaccharides or monosaccharides (Khamis et al., 2020), increased in relative abundance in the 1 438 mg/L and 50 mg/L conditions. Both species have been positively correlated to the expression of the 439 ARG sul2 and has been suggested as potential host bacteria for ARGs (Li et al., 2021). However, *Clostridium sensu stricto 1*, was found to be downregulated – in which the number of total counts 440 441 detected decreases - in the blank samples, again suggesting that the change was driven more by the presence of glucose than the presence of the amoxicillin. Increased relative abundance was 442 443 exclusively observed in taxa which were already abundant whilst decreasing relative abundance (less abundant in the test condition than in the seed community), was observed in comparatively rarer taxa 444 (dark blue bars in Figure 7-11). In general, there were more taxa downregulated in the control and 445 the 1 mg/L conditions than in the 10 mg/L and 50 mg/L conditions, however there was little overlap 446 447 within these changes.

Family	Description
Spirochaetaceae	A facultative saccharolytic rod shaped bacteria (Karami et al., 2014)
Aminicenantales	A hydrolytic bacteria which is able to degrade carbohydrate and protein matrixes
	and can provide sulphate-reducing and sulphurising bacteria with biochemical
	metabolism substrates. (Tao et al., 2021; Fan et al., 2022)
Bacteroidetes_vadinHA17	A hydrolytic and fermentative bacteria producing H ₂ , CO ₂ , fatty acids and alcohols
	(Wang et al., 2019)
Dysgonomonadaceae	A strict facultative anaerobic and fermentative bacteria capable of fermenting
	saccharides into organic acids. (Shamurad et al., 2020)
Syntrophobacteraceae	Syntrophobacteraceae is a member of the Class Syntrophobacterales. They are
	sulfate reducing strict anaerobes. (Waite et al., 2020)
Methanobacteriaceae	Methanobacteriaceae are a rod to fillementous archea are able to grow on H2+CO2
	and formate 2-propanol/ CO2. (Garcia, Patel and Ollivier, 2000)
Synergistaceae	Synergistaceae are Gram-negative, nonmotile rods who are strictly anaerobic. They
	are chemoorganotroph capable of fermenting amino acids into formate, acetate,
	propionate, H ₂ , and ammonia. They are not able to ferment carbohydrates. (Allison,
	MacGregor and Stahl, 2015)
Clostridiaceae,	Involved in acetate production / consumption. (Esquivel-Elizondo et al., 2017)
Caldatribacteriaceae,	VFA-oxidizing family of anaerobic bacteria known to metabolize fatty acids as well
	as oxidize propionate and butyrate (Khoei et al., 2021)
Kosmotogaceae	Mesophilic bacteria capable of H ₂ oxidation and thiosulfate reduction using a sulfide
	dehydrogenase. (Nesbø et al., 2019)

448 *Table 7-6* – *The most prominent families in the top 25 taxa within the community.*



Figure 7-7 - The top 25 taxa found throughout the entire sample set. Samples are grouped by substrate and amoxicillin type. The x-axis represents the hours of incubation. Each of the replicates are grouped together for each time point.



Figure 7-8 – The core microbiome.

The core microbiome as represented by the genera found in at least 85% of samples for each of the test conditions. The x-axis represents the detection limit at varying relative abundances. Each sample contains the same most abundant genera.



Family Genus Anaerolineaceae uncultured Prolixibacteraceae Mariniphaga Syntrophobacteraceae Syntrophobacter TA06 TA06 Anaerolineaceae Flexilinea Anaerovoracaceae Spirochaetaceae Spirochaeta JS1 JS1 Hungateiclostridiaceae Christensenellaceae Anaerolineaceae Clostridiaceae Clostridiaceae Prolixibacteraceae Mariniphaga Lachnospiraceae Tuzzerella TA06 uncultured TA06 uncultured uncultured uncultured Oscillospirales Dethiobacteraceae uncultured **Babel** iales **Babeliales** Hungateiclostridiaceae uncultured Oscillospirales Dysgonomonada etrimonas Clostridiaceae Clostridiaceae Dysgonomonadaceae Petrimonas Prolixibacteraceae Mariniphaga TA06 TA06 Clostridiaceae Oscillospirales Hydrogenoanaerobacterium Dethiobacteraceae uncultured Dysgonomonadaceae Clostridiaceae Petrimona Clostridium_sensu_stricto_1 Clostridiaceae Clostridium_sensu_stricto_1 Prolixibacteraceae Mariniphaga SRB2 SRB2 Anaerovoracaceae uncultured Pseudomonadaceae Clostridiaceae TA06 TA06 JS1 JS1 Oscillospirales Dethiobacteraceae Anaerolineaceae Syntrophorhabdaceae **Babeliales** Babeliales Hungateiclostridiaceae uncultured Bathyarchaeia Oscillospirales SB-5 SB-5 Pseudomonadaceae Pseudomonas Lachnospiraceae WCHB1-81 Anaerocolumna WCHB1-81 D8A-2 D8A-2 Hungateiclostridiaceae uncultured Anaerolineaceae Flexilinea Spirochaetaceae uncultured Peptococcaceae uncultured Hungateiclostridiaceae Ercella Christensenellaceae uncultured uncultured Pirellulaceae uncultured Anaerovoracaceae Gracilibacteraceae Syntrophorhabdaceae Anaerolineaceae uncultured Clostridiaceae Spirochaetaceae Spirochaeta uncultured uncultured uncultured

Family_XIII_AD3011_group Ercella Christensenellaceae_R-7_group uncultured Clostridium_sensu_stricto_1 Clostridium_sensu_stricto_1

Hydrogenoanaerobacterium Hydrogenoanaerobacterium Clostridium_sensu_stricto_1 Clostridium_sensu_stricto_1

Clostridium_sensu_stricto_1

Pseudomonas Clostridium_sensu_stricto_1 Hydrogenoanaerobacterium uncultured Levilinea Syntrophorhabdus Bathyarchaeia Hydrogenoanaerobacterium Christensenellaceae_R-7_group Family_XIII_AD3011_group Lutispora Syntrophorhabdus Clostridium_sensu_stricto_10 uncultured Candidatus Soleaferrea Paracoccus Fastidiosipila Pla4_lineage

Figure 7-9 – Taxa differential.

Ruminococcaceae

Rhodobacteraceae

Pla4_lineage

Hungateiclostridiaceae

Taxa differential which identifies the taxa with the greatest differences from the mean. The plot shows both logfold changes (green bars) as well as mean abundances (blue bars).

451 8.2.5 Taxa-Functional Robustness

452 Taxa-function robustness estimates community resilience and stability through measuring 453 distribution of genes associated with the structure-function relationship between taxonomic and 454 functional profiles which are reported via attenuation values with an increase in attenuation representing an increase in community robustness (Eng and Borenstein, 2018). As anaerobic 455 456 community structures are highly syntrophic, changes in community structure can potentially disrupt 457 functionality and overall taxa-function robustness particularly when those changes target highly 458 related species. To provide a quantitative comparison of the difference in taxa-function robustness 459 between the microbial communities exposed to amoxicillin over time, the attenuation values for each 460 treatment were assessed over time as defined by the taxa-function response curves. Attenuation 461 values report how much change a community can withstand before the overall robustness of the 462 community is compromised in which low values represent low robustness and high values represent 463 high robustness. Within this study, the robustness of the control community was found to remain 464 broadly similar throughout the study. Conversely, in the test conditions as well as the unfed blank 465 conditions a drop in overall robustness was observed between the beginning and end of the study. 466 However, the 1 mg/L condition resulted in an initial drop in robustness followed by a stabilising 467 effect after T4 (24 hours). These results indicate that although methanogenic activity in the 468 community was not observed to change, the presence of amoxicillin in all conditions resulted in a less robust community which may be more susceptible to other forms of disturbance such as 469 470 temperature fluctuation or changes in loading rate.



Figure 7-10 – *Robustness (attenuation parameters) through time for each of the samples. Greater values indicated more robustness while lower values indicate less robustness.*

471 **8.3 ANTIMICROBIAL RESISTANCE**

As the fundamental activities of antibiotics are naturally occurring, so too are the genes associated 472 473 with resistance mechanisms. As such, microbial communities often harbour a natural pool of 474 antibiotic resistance genes within the community, though they may not be activated. This study aimed 475 to establish which genes were present in the genome of the seed community and to detect the relative 476 increase in abundance of any antimicrobial resistance genes within the anaerobic granules when 477 exposed to a shock load of amoxicillin. Detection was conducted in two phases of qPCR array (as 478 described in Methods Section 0). In the initial screening phase a suite of 248 genes were tested using 479 a subset of five samples. This initial set contained genes that encode for resistance directly (beta-480 lactamases, tetracycline resistance, and multi-drug resistant genes) as well as genes that facilitate the spread of resistance (mobile genetic elements and integrons). The remaining samples were quantified 481 482 for the 34 most abundant genes identified in the first screening and two 16s rRNA primers.

483 In the second screening run, 11 genes were found in most samples (**Table 7-7**): four antimicrobial 484 resistance genes (cfxA, tetQ, mexF, and oprJ), and seven genes associated with mobile genetic 485 elements and integrons (intl3, intl1_2, IS6100, IS1247_1, ISEcp1, tnpA_2, and tnpA_5). Of these, 486 nine were found in all samples. Cfxa is a class A cephalosporinase which inactivates beta-lactams through a deactivation of the beta lactam ring. The detection and proliferation of cfxA in anaerobic 487 488 bacteria has been associated with the presence of amoxicillin (Iwahara et al., 2006). The tetQ resistance gene confers resistance against tetracycline antibiotics through modification of the 489 490 ribosome. The mexF and oprJ are associated with the tripartite efflux pump systems MexCD-OprJ 491 and MexEF-OprN. These systems are found within the gram-negative bacteria Pseudomonas aeruginosa (Terzi, Kulah and Ciftci, 2014). The mex pumps move compounds from the inner 492 493 membrane into the pump while the opr pump moves compounds from the outer membrane out of the 494 cell (Aeschlimann, 2003; Terzi, Kulah and Ciftci, 2014). While the mexF and oprJ genes are 495 associated with different systems, they are closely related, and each system has been found to confer 496 resistance for beta-lactam antibiotics (Aeschlimann, 2003).

498 *Table 7-7 - The antibiotic resistance genes detected in the qPCR array.*

Gene	Туре	Function
ofr A	Beta-lactamase	AMR Gene - Class A cephalosporinase which enzymatically deactivate the
Сјлп		beta-lactam ring. (https://card.mcmaster.ca/home)
	Tetracycline	AMR Gene - Ribosomal protection protein. Its gene is associated with a
tetQ	Pasistance	conjugative transposon and has been found in both Gram-positive and Gram-
	Resistance	negative bacteria. (<u>https://card.mcmaster.ca/home</u>)
marF		AMR Gene - multidrug efflux system MexF is the multidrug inner membrane
телт	MDR	transporter of the MexEF-OprN complex (<u>https://www.uniprot.org/</u>)
oprI	WIDK	AMR Gene - OprJ is the outer membrane channel component of the MexCD-
oprij		OprJ multidrug efflux complex. (<u>https://www.uniprot.org/</u>)
intl3		Integron-integrase gene which catalyses the insertion of gene cassettes into and
int 11 2	Integron	enables the capture and insertion of exogenous genes in to the genome (Gillings
<i>m</i> 111_2		<i>et al.</i> , 2014)
IS6100		Insertion sequence (IS) – transposable elements found in bacterial species and
IS1247_1		plasmids that encode gene for mobilisation and insertion (Varani <i>et al.</i> 2021)
ISEcp1		plasmas that choose gene for moonsation and insertion (varian et al., 2021)
tnpA_2	MGE	
		Transposon - Mediates transposition of transposon tnp by a 'cut and paste'
tun A 5	5	mechanism (Steiniger-White et al., 2002). TnpA transposons are associated with
inpA_3		IS21, IS6, IS6100, IS1216, and ISEcp1 insertion genes (Muziasari et al., 2017)

499

500 Mobile genetic elements (MGEs) are DNA segments that encode enzymes to copy and transport 501 genetic material both intracellularly and extracellularly promoting the horizontal transfer of genetic 502 material including the spread of ARGs (Pazda et al., 2019). Insertion sequences (IS) and transposons 503 (Tn) are small portions of DNA that transport themselves (and associated genes) to new locations. 504 Integrons (int) use site-specific recombination to catalyse the insertion of gene cassettes into DNA 505 and enables the capture and insertion of exogenous genes in to the genome (Partridge et al., 2018). 506 This study detected the insertion sequences IS6100, IS1247_1, and ISEcp1 as well as the transposons 507 tnpA_2 and tnpA_5. The tnpA transposons have been found to be associated with IS21, IS6, IS6100, 508 IS1216, and ISEcp1 insertion genes (Muziasari et al., 2017). Furthermore, the intl3 and intl1_2 genes

509 associated with the Integron-integrase gene were detected in the samples. The presence of *int11* genes 510 has been suggested as a proxy for the presence of antimicrobial resistance (Gillings et al., 2014). 511 As expected, AMR genes were detected in the seed community despite no known previous exposure 512 to amoxicillin or any other known antibiotics. Within these samples, the beta-lactamase gene cfxA 513 was found to be the most prevalent comprising 75.39% of all the detected gene counts and was 514 observed to increase in relative abundance in amoxicillin exposed microbial communities. The 515 second most abundant gene detected was the multi drug resistant mexF gene at 6.82%. There 516 appeared to be a general increase in the total number of AMR genes detected relative to the 16S gene 517 though there was a peak at T6 (30 hours) followed by a decline in relative abundance which was 518 most notable for *cfxA* (Figure 7-13).



Figure 7-11 – *Abundance of Genes Relative the 16S Gene Over Time*

(A) The pie chart represents the proportion of the total AMR gene counts detected in all samples as a percentage of all counts. Bet – beta lactam genes, MGE – mobile gene element, Tet – tetracycline resistance gene, MDR – multidrug resistance gene. Each stacked bar chart shows the AMR gene counts relative to total 16s gene counts through time for each test conditions. (B) Bar chart B includes all the genes detected. (C) Bar chart C represents contains all genes except the cfxA gene.

519 8.4 DISCUSSION

520 This study aims to assess whether a shock exposure to the antibiotic amoxicillin within the influent 521 of anaerobic systems, as might be found within decentralised wastewater treatment, could inhibit 522 methanogenesis and sCOD removal. As previous work presented in Chapter 3 demonstrated that 1 523 mg/L, 10 mg/L, and 50 mg/L of amoxicillin concentrations can decrease methanogenic activity, it 524 was assumed that an increase in amoxicillin exposure would result in a corresponding reduction in sCOD removal throughout the study. The sCOD over time was observed to increase and decrease 525 526 from the beginning and end of the study. However, the sCOD of the final effluent values taken at 48 527 hours were statistically similar to eachother, indicating that although overall sCOD concentration 528 during the study was different between test conditions (Figure 7-6), the overall ability to remove 529 sCOD at 48 hours was not greatly impacted by the presence of the amoxicillin. It should be noted 530 that these results agree with the work conducted by Massé et al., (2000) which reported that the 531 presence of penicillin did not have a significant effect of sCOD removal over several fill draw cycles 532 of an anaerobic SBR.

533 There are two possible drivers for the fluctuations in sCOD observed in this study. The initial drop 534 in sCOD in the control samples is likely caused by the utilisation of the glucose substrate in the 535 absence of amoxicillin. However, the presence of amoxicillin inhibited the full conversion of glucose 536 to methane in which intermediate products would remain in solution. This, along with damaged 537 cellular material, could have resulted in an increase in the sCOD observed. Amoxicillin inhibits 538 growth by irreversibly bonding with the peptidoglycan in the cellular membrane (Madigan, 2014). 539 This damages cell walls, inhibits the repair of the damage, and prevents the production of new cells 540 until all the free amoxicillin is used or destroyed and new PBP enzymes can be created. Once the 541 free amoxicillin is no longer active, the microbes would be available to utilise any free substrates, 542 which may explain the delayed drop in sCOD in the amoxicillin test conditions. The conversion of 543 glucose into methane requires a wide variety of microbes able to both utilise glucose as well as all 544 its by-products. As this damage inhibits cellular function, the glucose present would not be able to be fully converted into methane as efficiently. Further study measuring variations in RNA or 545

546 exclusively measuring living cellular materials within the effluent can confirm this hypothesis.
547 Ultimately, the cause of fluctuating sCOD over the course of this experiment is unknown. This
548 warrants further study which monitors changes in sCOD over time during amoxicillin exposure
549 events. To determine if the increase in sCOD is due to a build-up of methane production
550 intermediates, a study which includes VFA quantification in the effluent can determine which trophic
551 classes are being inhibited.

552 Anaerobic microbial communities are known to grow at a slower pace than aerobic microbial 553 communities as the anaerobic metabolic pathways yield less energy for growth using alternative 554 electron acceptors to oxygen (Gerardi, 2003). As such, many studies of anaerobic microbial 555 communities are conducted over days and weeks with highly intermittent sampling conducted over days. This misses the any immediate effects within the few hours of exposure. The hydraulic 556 557 retention time of decentralised WWT systems such as septic tanks can be on the order of days. 558 Therefore, the immediate aftereffects within the first few hours of amoxicillin can have knock on impacts on treatment efficiency during that time. Thus, studying response over short time intervals 559 560 is important.

561 This study showed that changes in the community can be detected within hours after an amoxicillin 562 exposure event. A two-to-three-fold increase in the relative net presence of AMR resistance genes 563 within effluent can be detected when amoxicillin in the influent is as low as 1 mg/L (Figure 7-13). 564 Changes to the overall community structure resulting from the presence of amoxicillin was observable from the first point at 6 hours from inoculation (Figure 7-8). However, as this study 565 focused on the use of 16s rRNA, which both captures the DNA within live as well as dead cells the 566 567 full extent of the variation may not have been captured. As such, the community shifts observed may 568 underestimate the real changes within the community, ie, only those living after an acute exposure 569 to amoxicillin. Future work which focuses on the RNA within the active communities could further 570 develop these results. In the absence of amoxicillin, the community shifted with the presence of 571 glucose substrate, which was not seen to the same extent when amoxicillin was present (Figure 7-572 7). This suggests, that even if methanogenic activity occurs at the same rate with and without

amoxicillin, small concentrations (1mg/L and less) can have an effect on the community make up.
Finally, a decrease in community robustness was observed within 24 hours of exposure to amoxicillin
for all tested conditions indicating communities exposed to amoxicillin can be sensitive to further
stresses (Figure 7-12).

577 Changes in the resistome were also detectable within the first 48 hours. As predicted, AMR genes 578 were detected within the seed community which had not been exposed to any known antibiotics 579 within the three years prior to the study. A linear model found that beta-lactam resistance gene cfxA 580 was positively correlated to the diversity and time within study. The relative abundance of the efflux 581 pump encoding mexF and orpJ were also found to linearly correlate to time for the 50 mg/L exposure 582 condition. Although *mexF* and *orpJ* each encode for different efflux pumps they both belong to the RND efflux pump family which are capable of conferring resistance to amoxicillin (Iman Islamieh 583 584 et al., 2018). Despite net increases in the relative AMR gene counts for T5 (30h) and T6 (36h), the 585 total AMR counts appeared to decrease for T7 (48h). This suggests that while AMR genes might be produced in response to the presence of amoxicillin, once the antibiotic is no longer active (through 586 587 decay or non-reversable PBP bonding) these genes are no longer selected for. Further research over 588 a longer time period or the inclusion of RNA sequencing for activated genes could explain the 589 changes in AMR genes observed in T7.

590 In conclusion, this research suggests that acute exposure to concentrations of amoxicillin used in this 591 study have a detrimental effect on an anaerobic activity and structure of microbial communities. 592 While there wasn't a change in the methanogenesis or sCOD removal within the first 48 hours, there were fluctuations in sCOD removal within the first 12 hours appeared to be inhibited. As septic tanks 593 594 are continuous flow, even a temporary decrease in treatment efficiency can result in the discharge of 595 untreated waste into uncontrolled environments. Furthermore, even small concentrations of 596 amoxicillin (1 mg/L) within the influent have been shown to have a measurable effect on the 597 microbial community. Most notably this includes the presence of AMR genes in the community and 598 potentially the effluent, as well as a decrease in community robustness making the system more 599 susceptible to knock on effects on performance. If effluent quality decreased, such than an increase

in organics are released in the effluent, these could expose the surrounding environment to AMR
genes. It should be noted that there were several specific differences in the design between this study
and an in-situ septic tank. For example, the temperature in traditional septic tanks, composition of
the substrate in the influent, and the pattern of exposure. However, research involving heated septic
tanks (Koottatep *et al.*, 2018) are currently being conducted, for which the results of this study would
be relevant.

606 8.5 CONCLUSIONS

While many studies looking into the effects that antibiotics can have on anaerobic communities look 607 at longer time scales with data points on the scale of days (Cetecioglu, 2014; Aydin, Ince and Ince, 608 609 2015b), this study demonstrates that changes in the treatment efficiency, microbial community, and 610 relative abundance of AMR genes can be detected within the first 48 hours of exposure. Changes in 611 the patterns of sCOD removal was detected within the first 12 hours with an increasing effect with 612 increasing amoxicillin exposure conditions. There was a measurable effect on the microbial community for all amoxicillin concentrations present (1mg/L - 50mg/l). Furthermore, a two-to-613 614 three-fold increase in the relative net presence of AMR resistance genes within effluent can be 615 detected in when amoxicillin in the influent is as low as 1 mg/L within the first 36h of incubation. 616 Consequently, decentralised waste treatment technologies relying on anaerobic microbial 617 communities for treatment, such as septic tanks, should be considered potential point source 618 pollutants for the spread of antibiotics and AMR genes into the wider environment, particularly 619 within the first 48 hours after exposure.

620 **8.6 FUTURE WORK**

The results of this research suggest that the presence of amoxicillin can make septic tanks more vulnerable to additional stresses, and therefore, further work is required before drawing concrete conclusions and generalisations of these on different scenarios with varying environmental conditions. Based on our analyses, future studies could include: While this study focused on the first 48 hours after acute exposure, studies which focus on
the effects during longer time scales can capture recovery, and determine if there is a point
in which amoxicillin exposed conditions and the control converge in terms of community
and waste treatment efficiency. It would be important for these studies to include
observations on an hourly rather than daily basis.

Septic tanks insitu can have a variety of temperatures depending on region and time of year.
 This study utilised mesophilic temperatures, which are rarely observed in insitu septic tanks
 (though some research is currently underway developing heated septic tanks). As
 biochemistry can change with different temperature ranges, future studies could repeat this
 work for communities adapted to different temperature ranges, closer to those as would be
 found in operating septic tanks.

This work demonstrated that there were changes in the sCOD during the first 12 hours and
to the community for all amoxicillin exposure conditions. Future studies which focused on
the analysing changes in function through KEGG ortholog analysis would indicate
specifically which pathways are inhibited by the presence of amoxicillin within the
community.

641

² Chapter 8

The Structural Stability of Amoxicillin within Anaerobic Culture Medium

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8 9.1 INTRODUCTION

9 Pharmaceutical micropollutants entering water treatment facilities and waterways is an issue of 10 global concern. Beta-lactam antibiotics, including penicillin and amoxicillin, are amongst the most 11 widely used antibiotics in the world comprising 50-70% of total antibiotics used in most countries 12 (Surridge and Watson, 2012; Loos et al., 2018). However, this widespread usage also leads to increased risks to the aquatic environment through water discharge from treatment facilities as well 13 as runoff from agricultural settings (Kümmerer, 2009a, 2009b). The spread of amoxicillin within the 14 15 environment can lead to the proliferation of antimicrobial resistance (Zhang, Zhang and Fang, 2009; 16 Rizzo et al., 2013) in turn leading to a decrease in the therapeutic potential of these lifesaving drugs (Zaman et al., 2017). It has been suggested that if antimicrobial resistance is allowed to spread, the 17 resulting loss of our antimicrobial drugs could be as large a threat to humanity as climate change 18 19 (Torjesen, 2013). To curb the spread of amoxicillin and other pharmaceuticals into the environment and the resulting wave of antibiotic resistance which follows, it is important to understand both how 20 21 these drugs behave as they move through our water treatment systems and the resident microbes responsible for treatment. 22

23 The toxicity of pharmaceuticals on culturable microbes is measured in a variety of ways. Techniques such as disc diffusion and tube dilution are traditionally used when studying the antibiotic effect on 24 single organisms capable of growing on agar medium (Madigan, 2014). However, many microbes 25 are not readily grown on plates including those who grow in syntrophic pairs, groups, or in complex 26 microbial communities found such as those found in wastewater treatment processes. In these cases, 27 the toxicity is measured by changes in microbial activity when microbes or microbial communities 28 are exposed to known concentrations of the target compound (OECD, 2007). Anaerobic microbial 29 30 communities, which underpin processing in anaerobic digesters for wastewater treatment, require specialised culturing conditions in which methanogenic activity inhibition is measured through 31 32 relative changes in the volume and rate of methane formation (Alexy, Kümpel and Kümmerer, 2004; OECD, 2007; Cetecioglu et al., 2013; Aydin, Ince, et al., 2015). 33

34 The degradation pathways of amoxicillin have been understood as far back as 1971 (Hou and Poole, 35 1971). The antibiotic activity of beta-lactams operates primarily through its characteristic fourmembered lactam ring structure (Loos et al., 2018) (Figure 8-1). When this structure is hydrolysed, 36 either through enzymatic actions - which are the primary mechanism of biological beta lactam 37 resistance - or through chemical or photodegradation, the beta lactam ring opens and deactivates the 38 39 molecule. (Vahdat and Sunderland, 2007; Längin et al., 2009). Research into the stability of beta 40 lactam drugs have shown that the rate of decay of the beta lactam ring is influenced by factors such 41 as pH, concentration and conductivity of the buffer, and the temperature of the matrix as a whole 42 (Chadha, Kashid and Jain, 2003; Pratama, Peterson and Palmese, 2012; Braschi et al., 2013). As the focus of this work is drug delivery, most of this work primarily utilised buffers with the aim of pH 43 44 modification.

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Figure 9-1 - Amoxicillin is abiotically degraded through the hydrolysis of the beta-lactam ring (blue). This ring is the active site of the molecule which interacts with the peptidoglycan within the microbial cell wall. It can abiotically decay through a nucleophilic attack on the carbonyl group within the beta-lactam ring (pink). The example above shows hydrolysis with a water molecule, though other nucleophilic compounds can attack using the same mechanism.

52 However, studies which measure the antibiotics efficacy and potency in terms of growth inhibition 53 rarely consider the impact of abiotic decay. A wide variety of compounds are added to culture media 54 used to sustain anaerobic microbial communities in laboratory cultures during inhibition tests. 55 Anaerobic culture medium can include oxygen scavengers, salt nutrients, indicators, stabilisers, as 56 well as buffers (Shelton and Tiedje, 1984; Aydin, Shahi, et al., 2015). Yet, the influence of media 57 components on the structural stability of the analyte of interest within culture-based toxicity studies 58 are rarely if ever considered. When decay is discussed, it is assumed that degradation is primarily 59 biological in nature (Alexy, Kümpel and Kümmerer, 2004; Hijosa-Valsero et al., 2011). Therefore, 60 the influence of the individual components within culture media on the decay rate of beta-lactam antibiotics independent from the biological degradation is unknown. 61

The aim of this study was to determine the effect of commonly used chemical components of anaerobic culture medium on the decay rate of amoxicillin in solution at concentrations used in other chapters of this thesis. Understanding the role that commonly used additives within anaerobic culture medium have on the degradation rates of amoxicillin will help to disambiguate between the biological and chemical degradation of amoxicillin within these studies. This study theorised that the rate of amoxicillin degradation is determined by the chemical structures of other solutes in the solution.

69 9.2 MATERIALS AND METHODS

70 9.2.1 Overview of experiment Design

71 To determine how anaerobic culture medium components influence abiotic amoxicillin decay, the 72 decay rate of amoxicillin was tested in combination with each component individually. The chemical 73 components tested included commonly used buffers, organic and inorganic nutrients and an oxygen 74 scavenger (Figure 8-2). Amoxicillin, at a range of initial concentrations 0.0008, 0.0040, 0.0080 mM 75 which correlates to 1 mg/l, 5 mg/l, and 10 mg/l to reflect concentrations that are used in Chapters 6 76 and 7, was stored in solution with individual components of the anaerobic medium at (i) at 35° C, the 77 most common incubation temperature for anaerobic culture of wastewater microbial communities 78 (Figure 8-3), and (ii) at -20°C, a common temperature for storage of samples arising from such 79 studies (Figure 8-4). To determine decay rates, samples were taken periodically, and amoxicillin concentration was quantified using LC-UV (Figure 8-2). 80

81



Figure 9-2 - Summary of the experimental design and logical flow of this study.

It is often assumed that as anaerobic media is pH neutral (7.0 pH) that chemical degradation of amoxicillin within culture media would be minimal. However, this assumption has not been thoroughly tested. This study measures the rate of amoxicillin decay in the presence of the individual components of anaerobic media.


Figure 9-3 - Experiment No 1 – Decay of amoxicillin at 35°C.

This experiment measures the decay of amoxicillin when incubated in the presence of one of the components found in anaerobic media adjusted to a pH of 7.0. Two additional buffers were also added to show how pH variation can alter amoxicillin stability and change the rate of decay.

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Figure 9-4 - *Experiment No 2* – *Decay of amoxicillin when stored at -20°C. This experiment measured the decay of amoxicillin at different pH values at temperatures used for storage.*

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During the course of experimentation, it was established that when 1-cysteine was in excess,
amoxicillin decayed at a greater rate than could be detected under the initial experimental conditions.
Therefore, an additional experiment was conducted in which varied 1-cysteine concentrations (0.125-
87 8mg/l) were incubated with amoxicillin (at 10mg/l) (Figure 8-5). The aim of this work was to
determine the concentration of 1-cysteine at which amoxicillin decay rates could be monitored using
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Figure 9-5 - *Experiment No 3 – Measuring the rate of decay of amoxicillin in the presence of differing concentrations of l-cysteine.*

Experiment no 1 found that the decay of amoxicillin was instantaneous in the presence of 8mM of 1-cysteine and an accurate rate measurement was not able to be taken. As the rate of amoxicillin decay is dependent on the concentration of 1-cysteine, lowering the concentration of a 1-cysteine should also slow the rate of decay of amoxicillin. As such, this experiment aims to determine the ratio of amoxicillin to 1-cysteine in which amoxicillin decay can be measured.

- 89 manual sampling and amoxicillin detected using an HPLC. This work was conducted at 35°C to
- 90 mimic the incubation temperatures preferred by mesophilic microbes.
- 91 9.2.2 Reagents

92 9.2.2.1 Test compounds selected from anaerobic growth medium

93 The test compounds and concentrations used in this study were chosen to reflect those used in 94 anaerobic medium specific to the SMA test and common to biological activity inhibition assays 95 reported in the literature (Colleran et al., 1992; D. Coates, F. Coughlan and Colleran, 1996; 96 Cetecioglu, 2014). In total, the effect of seven compounds on amoxicillin decay was tested in this 97 study (Table 8-1) at a concentration of 40mM each with the exception of glucose and l-cysteine 98 which were used at concentrations relevant to culturing methods. Three of the seven compounds 99 were buffers. Carbonate and phosphate are both used in anaerobic medium for the purpose of pH 100 buffering (Shelton and Tiedje, 1984; OECD, 1997). Acetate is commonly used as a buffer in 101 chromatography (Marco et al., 2017), though it is readily degradable by methanogens during

102	anaerobic culturing and used as a feed source in experimentation (Heidari, Fatemeh Nabavi, Saffari
103	Khouzani and Mehdi Amin, 2012; Aydin, Ince, et al., 2015). L-Cysteine is an oxygen scavenger used
104	in anaerobic culturing (Rymovicz et al., 2011). Sodium chloride was used to represent inorganic
105	nutrients added to cultures. The standard trace nutrient mix was not used for this experiment as the
106	metal salts could potentially damage the LC column. Glucose is highly bioavailable and a commonly
107	used substrate. Milli-Q water was used as control. This study hypothesised that as the Milli-Q water
108	has few dissolved substances, that this would produce the slowest decay rate. All seven compounds
109	were adjusted to a pH of 7.0 to remove pH as a variable and hence enable direct comparison.
110	Additionally, acetate and carbonate buffers were used at pH=5.0 and pH=9.0 respectively. This pH
111	range represents that typical of anaerobic systems used for wastewater treatment, and enabled
112	observation of the influence of pH on the rate of decay. All mixes were autoclaved before the addition
113	of amoxicillin stock. All conditions were tested at a temperature of 35°C, though Milli-Q® Type 1
114	Ultrapure Water, acetate (5.0pH), phosphate (7.0pH), and carbonate (9.0pH) were additionally tested
115	at -20°C (Figure 8-3 and 8-4).

Compound	pH	Function	Molarity	Conductivity (uS)	Mix	
Water	7.0	Primary Matrix	NA	4.0	NA	
		Buffer			NaC ₂ H ₄ O ₂	2.209 g
Acetate Buffer	5.0	Easter	40 mM	1619	$C_2H_4O_2$	0.785 g
		rood source			MilliQ	1000ml
A astata Duffar	7.0	Buffer	40 mM	25.90	NaC ₂ H ₄ O ₂	3.28 g
Acetate Duffer	7.0	Food source	40 11111	2380	MilliQ	1000ml
		Buffer			K ₂ HPO ₄	3.737 g
Phosphate Buffer	7.0		40 mM	4460	KH ₂ PO ₄	2.523 g
					MilliQ	1000ml
Carl an ata Driffan	7.0	Buffer	40 mM	2528	NaHCO3	3.36 g
Carbonate Buller				2328	MilliQ	1000ml
					NaHCO ₃	3.822 g
Carbonate Buffer	9.0	Buffer	40 mM	3140	Na ₂ CO ₃	0.477 g
					MilliQ	1000ml
Glucose	7.0	Food Source	1 M	9200	C6H12O6	180g
Glucose	/.0	1 ood Source	1 1/1	5200	MilliQ	1000ml
NaCl	7.0	Nutrient additive	40 mM	5.6	NaCl	2.34
11401	/.0			5.0	MilliQ	1000ml
L-Cysteine	7.0	Oxygen Scavenger	8 mM	37	L-Cysetine0.969g	
Lojstenie	'.0	Saygen Seavenger			MilliQ	1000ml

116 *Table 9-1* - *Test compounds and solutions used during the experiment. Conductivity tested at* 20°*C*.

118 9.2.2.2 Amoxicillin solution preparation

The concentrations of amoxicillin (1mg/l, 5 mg/l, and 10mg/l) used in this study were chosen to reflect concentrations used in other chapters (Chapter 6 and Chapter 7) in this thesis. Working solutions were prepared by mixing pure amoxicillin (Sigma Aldrich) with autoclaved Milli-Q water to 100 mg/l then diluted with the appropriate test solution to obtain 1.0 mg/l, 5.0 mg/l, and 10 mg/l. If necessary, solutions were pH adjusted dropwise using dilute HCl or KOH until reaching the desired pH. All bottles used during mixing were rinsed with methanol and MilliQ before use.

125 9.2.2.3 L-Cysteine solution preparation

To measure the interactions between l-cysteine and amoxicillin, an additional study was run with seven different amoxicillin concentrations. The aim of this work was not to determine the rate of decay, but rather to determine the maximum ratio of l-cysteine to amoxicillin which will produce a decay curve that is detectable on the LC-UV. Therefore, l-cysteine was prepared as an 8.0 mg/l stock solution by dilution in MilliQ water and serial dilutions were made from the stock again using MilliQ water to produce the concentrations given in **Table 8-2**.

Table 9-2- L-Cysteine and amoxicillin concentrations and the ratios of cysteine to amoxicillin used during
 experimentation.

L-Cys	teine	Amoxi	Ratio	
Concentration	Molarity (mM)	Concentration	Molarity (mM)	LC : AMOX
0.125 mg/l	0.0001	10 mg/l	0.0027	0.04 : 1
0.25 mg/l	0.0002	10 mg/l	0.0027	0.08 : 1
0.5 mg/l	0.0004	10 mg/l	0.0027	0.15 : 1
1.0 mg/l	0.0008	10 mg/l	0.0027	0.30 : 1
2.0 mg/l	0.0017	10 mg/l	0.0027	0.60 : 1
4.0 mg/l	0.0033	10 mg/l	0.0027	1.21 : 1
8.0 mg/l	0.0066	10 mg/l	0.0027	2.41 : 1

134

135 9.2.3 Incubation, storage and sampling

The experiments were conducted at two different temperatures, 35°C to simulate culturing
conditions, and -20°C to simulate frozen storage conditions. For experiments incubated at 35°C, a

45 ml working solution was prepared for each test (containing the test compound plus amoxicillin).
Incubation bottles (30mL serum vials) were soaked in Virkon TM (Lanxess) before use, after which
they were rinsed three times and dried. The 45mL working solution was then sub-divided between 3
bottles (15mL in each) such that each test condition was prepared in triplicate. Sealed bottles were
placed in a 35°C incubator and sampled approximately every 8 - 12 hours until 7 timepoints were
measured. To sample, a 1.0 ml aliquot was removed through the septa and placed in a sample vial
and immediately quantified using the LC-UV.

To test frozen storage conditions, 30ml of each dilution condition was prepared. A set of the freshly prepared mixtures were immediately analysed. The rest was distributed into 1.5ml aliquots and placed in a -20°C freezer. Samples were removed, thawed, and quantified using an LC-UV on days 0, 15, 147, and 177. Each condition was tested in triplicate.

149 9.2.4 Chromatographic Analysis

150 The amoxicillin was quantified using a Shimadzu LC20AT liquid chromatograph (Kyoto, Japan) 151 equipped with an SIL 20AHT Autosampler and a DGU-20 A Degassing unit. Chromatographic 152 separation was achieved using a Purospher® STAR RP-18 end capped column (5µm, 150mm x 4.6 153 mm) Merck KGaA, Darmstadt, Germany. Detection was conducted using an SPD-10 UV Detector 154 using wavelengths 295nm. The system was controlled by LabSolutions (Agilent Technologies). The 155 injection volume was 100µl. The flow rate was 0.9 ml/min. The mobile phases used were A: acetate 156 buffer (0.01M, pH 5.0) and B: HPLC grade methanol. The elution gradient: 0 - 4 min 10% B, 4 - 18 157 min 10 - 30% B, 18 - 20 min 30% B, 20 - 25 min 10%.

The optimisation and method validation for the LC-UV analysis is detailed in the Methods chapter. Briefly, the validation was based on protocols described by the International Conference of Harmonization (ICH) Guidelines (FDA, 1995; US FDA, 1996) and Liquid Chromatography method development methods by Injac et al, (2009) and Martínez-Huelamo et al. (2009). The method validity was determined by preparing and quantifying a range of standard solutions of amoxicillin in MilliQ (0.1 – 10mg/l) in which linearity r²> 0.999. The method LOD was 0.123 mgL-1 and LOQ was 0.236 mgL-1 as defined by S/N≥2 for LOD and S/N ≥ 10 for LOQ.

165 9.2.5 Statistical Analysis

166 The decay of chemical compounds can be described using two interrelated variables, (1) the decay 167 constant k and (2) half-life. (Ebbing and Gammon, 2016). Test compound concentrations used were 168 normalised to 40mM to maintain consistency across conditions (except for l-cysteine which was 169 8mM and glucose which used 1M to mimic concentrations used in other studies within this 170 dissertation. Thus, the concentrations of medium were several orders of magnitude more 171 concentrated than the amoxicillin concentrations used. It has been established both by theory (Ebbing 172 and Gammon, 2016) as well as by observation (Vahdat, 2000; Chadha, Kashid and Jain, 2003; 173 Kakimoto and Funamizu, 2007; Vahdat and Sunderland, 2007) that while the decay of amoxicillin 174 in the presence of other compounds with which it can react, the decay reaction as a whole is a second 175 order reaction such that the decay of depends both on the concentration of reactant as well as the 176 concentration of buffer. However, when the concentration of one of the reacting materials is in great 177 excess relative to the other (as occurs in this context) a second order reaction will behave like a first 178 order reaction as the concentration of the other solutes is effectively constant. As a result, the decay 179 in this study is said to be pseudo first order and reported as k'. As such, all values for the decay constant k are truly for the pseudo first order decay of k' and will be calculated using first order 180 181 decay calculations.

182 The second order rate reactions is calculated as follows:

$$Rate = -\frac{d[A]}{dt} = k[A][B]$$
 Equation

183

However, when B is >> than A, it can be treated as a constant. This produces a pseudo first order
reaction. Letting k' = [B]k give you the pseudo first order reaction. For this reason, the calculation
of the decay constant k and the half-life times will treat the decay as first order. The pseudo first
order rate reaction can be calculated as follows:

$$Rate = -\frac{d[A]}{dt} = k'[A]$$
 Equation 3

189 Rearranging this give you,

$$\frac{d[A]}{[A]} = -k'dt$$

190

191 Integrating this with respect to time gives,

$$\ln[A] - \ln[A_0] = -k't \qquad Equation 5$$

192

Equation 4 can be used to determine the rate at any given time in which *t* is the amount of time passed since the initial measurement, [*A*] is the concentration at that time, and [*A*₀] is the initial concentration. In this equation, *k* is the rate constant. While this constant should not change as the concentration of the target compound decays, *k* will change with temperature and will not be constant between the 35°C and -20°C studies and will need to be calculated separately for each temperature.

Equation 4

The value for k can be determined experimentally by plotting the natural log of the concentration against time. When the decay is first order, this relationship should produce a straight line (**Equation** 5) in which k is the slope of the line and $ln[A_0]$ is the constant b.

$$\ln[A] = -k't + \ln[A_0] \qquad Equation 6$$

201

For this study, the slope of this line, and therefore k', was determined by calculating the best fit line from the observed data. To assess the linearity of this line, a regression analysis was conducted for each of the test conditions and the r-square value was determined for the triplicate runs of each of the decay conditions at all time points. This was obtained separately for each of the different amoxicillin concentrations used.

207 The half-life was calculated using the value of *k* obtained using the following equation:

$$half \ life = \frac{\ln 2}{k'}$$

Equation 7

209 **9.3 RESULTS**

210 9.3.1 Amoxicillin Decay on 35°C

This research found that the rate of decay of amoxicillin was different depending on the differing composition of each of the matrices utilised in the study (**Figures 8-6, 8-7,** and **8-8, Table 8-3**). The slowest decay observed was in the glucose solution, though the decay results were highly variable in which the 10 mg/L (27.367 uM) amoxicillin had a k' of 0.0012 and a half-life of 563.31h, while the 1 mg/L (13.69 uM) amoxicillin had a k' of 0.0004 and a half-life of 1804.22 hours. The amoxicillin concentration detected within glucose solution was highly variable both during the study, as determined by the low R² values, as well as between replicates resulting in high p-values (**Table 8-**

218 **8**).

219 Amoxicillin in MilliQ water, acetate (7pH), acetate (5pH), and NaCl all had similar rates of decay 220 for each of the amoxicillin concentrations used. Amoxicillin in MilliQ decayed with a half-life of 544.09, 454.44, and 393.85 for 1 mg/L (27.37 uM), 5 mg/L (13.68 uM), and 10mg/L (2.74 uM) 221 222 respectively. Amoxicillin in 7pH acetate solution decayed with a half-life of 463.01, 466.71 and 289.96 for 1 mg/L (27.37 uM), 5 mg/L (13.68 uM), and 10mg/L (2.74 uM) respectively. Amoxicillin 223 224 in acetate at 5pH decayed with a half-life of 415.20, 415.20, and 354.38 hours for 1 mg/L, 5 mg/L, 225 and 10mg/L respectively. Amoxicillin in NaCl decayed with a half-life of 241.93, 325.42, and 288.15 226 hours for 1 mg/L, 5 mg/L, and 10mg/L respectively. Of these, the 5pH acetate solutions has the least 227 variance between replicates and samples ($R^2 = 0.697 - 0.928$) while the NaCl Solution had the 228 greatest in which there was variation within replicates ($R^2 = 0.520 - 0.749$).

Amoxicillin decayed at a similar rate in the 7pH carbonate solution and the phosphate solutions with half-lives ranging from 103.82-129.53 hours and 71.41-124.64 hours for the 7pH carbonate and phosphate solutions respectively. In both cases, the rate of decay was the slowest for the 1mg/L (2.74 uM) amoxicillin concentration, though this difference was most pronounced within the phosphate solutions in which the 1mg/L (2.74 uM). Amoxicillin in the 9pH carbonate solution decayed rapidly with mean half-lives ranging from 23.85 - 34.38 hours.

Amoxicillin in l-cysteine decayed so rapidly that an accurate measurement was not able to be taken. By the time the first sample was taken at 2 hours, the amoxicillin concentrations were less than 10% of the initial concentration and were below the limit of detection for some of the samples. After this point, the amoxicillin concentrations remained somewhat stable though, many samples were below the lower limit of detection.

240



Figure 9-6 - Amoxicillin decay at 35°C.

The Y axis for each figure measures the percent change from the initial concentration. The x-axis measures time. The trend lines measure exponential decay. 241 *Table 9-3* - Decay rates for amoxicillin in each of the solutions provided.

242 *P*-values are for liner regression analysis of the natural log of amoxicillin concentrations through time (hours). The slope of this was used to calculate the k' and half-life (*** $p \le 0.001$, ** 0.001 , * <math>0.01 .)

245 of this was used to calculate the k-and half-tipe ($(1, p \le 0.001, 1, 0.001$

Buffer type	Amoxicillin	1/	Half-Life	V	D ²	Standard	D l	
(pH)	Mg/L	ulvi	(M/ hour)	K'	K-	Error	P-valu	e
Chucoso	10	27.37	563.31	0.0012	0.477	0.0003	0.001	**
(7.0 pH)	5	13.68	1047.90	0.0007	0.109	0.0005	0.182	
(7.0 pH)	1	2.74	1804.22	0.0004	0.012	0.0009	0.671	
MiliO	10	27.37	393.85	0.0018	0.634	0.0003	< 0.001	***
(7.0 pH)	5	13.68	454.44	0.0015	0.965	0.0001	< 0.001	***
(, F)	1	2.74	544.09	0.0013	0.450	0.0004	0.002	**
Acotato	10	27.37	289.96	0.0024	0.731	0.0003	< 0.001	***
(7.0 nH)	5	13.68	466.71	0.0015	0.537	0.0003	< 0.001	***
(7.0 pm)	1	2.74	463.01	0.0015	0.723	0.0002	< 0.001	***
Acetate (5.0 pH)	10	27.37	354.38	0.0020	0.928	0.0001	< 0.001	***
	5	13.68	373.11	0.0019	0.877	0.0002	< 0.001	***
	1	2.74	415.20	0.0017	0.697	0.0003	< 0.001	***
Carbonate	10	27.37	129.53	0.0054	0.731	0.0007	< 0.001	***
(7.0 nH)	5	13.68	134.42	0.0052	0.758	0.0007	< 0.001	***
(//0 pri)	1	2.74	103.82	0.0067	0.905	0.0005	< 0.001	***
Phosnhata	10	27.37	124.64	0.0056	0.997	0.0001	< 0.001	***
(7.0 nH)	5	13.68	114.63	0.0060	0.991	0.0001	< 0.001	***
(7.0 pm)	1	2.74	71.41	0.0097	0.856	0.0010	< 0.001	***
Carbonate	10	27.37	23.85	0.0291	0.822	0.0034	< 0.001	***
(9.0 pH)	5	13.68	34.38	0.0202	0.970	0.0009	< 0.001	***
	1	2.74	31.52	0.0220	0.853	0.0024	< 0.001	***
NaCl	10	27.40	288.15	0.0024	0.520	0.0005	< 0.001	***
(7.0 nH)	5	13.70	325.42	0.0021	0.668	0.0003	< 0.001	***
(, h)	1	2.70	241.93	0.0029	0.749	0.0004	< 0.001	***

245



Figure 9-8 - The top figure plots the half-lives of amoxicillin in each of the conditions. Each bar represents a different concentration of amoxicillin used in the experiment. The bottom figure plots of the calculated k' values. The error bars in the bottle figure represent the standard error within the regression analysis used to calculate k'.



Figure 9-7 - Regression plot analysing the conductivity and concentration of the each of the solutions used in the experiment against the half life in amoxicillin within those solutions. This graph uses all the different solutions as well as all the half life calculations in a single pool.

248	To determine if the conductivity or concentration of the solutions used were correlated to the rate of
249	amoxicillin decay, a regression model was run between these values and half live for all amoxicillin
250	concentrations and solution types. These models found that conductivity $(R^2 - 0.27)$ and
251	concentration ($\mathbf{R}^2 - 0.04$) were both poor predicters for the rate of amoxicillin decay (Figure 8-8).

252 9.3.1.1 Amoxicillin Decay at -20 C

253 The results show that when frozen at -20°C, amoxicillin in pure milliQ water had the maximum half-254 life of 475.90 days (R² - 0.704) (Figure 8-9, Table 8-4). The decay rate of amoxicillin in acetate was 255 slightly shorter with an average half-life of 400.84 days. As with the decay at 35°C described in the 256 previous section, the decay rate of amoxicillin in the presence of phosphate and carbonate buffers 257 was much more rapid than for the other conditions studied. The decay of amoxicillin within 258 phosphate buffer was consistent between the three different concentrations in which half-lives ranged 259 from 32.49-33.34 days. Amoxicillin within carbonate buffer decayed at a similar rate, though the runs were not as consistent between the different concentrations used and half-lives ranged from 14.8 260 261 – 25.16 days.

²⁶³ *P*-values are for liner regression analysis of the natural log of amoxicillin concentrations through time (hours). The **264** *slope of this was used to calculate the k' and half-life (*** p \le 0.001, ** 0.001 , * <math>0.01 .)* **265**

Buffer type (pH)	ppm	uM	Half-Life (days)	k	R ²	Standard Error	P-value
	10	27.37	422.307	0.002	0.936	0.000	0.000 ***
Water (7.0 pH)	5	13.68	466.442	0.001	0.963	0.000	0.000 ***
	1	2.74	476.907	0.001	0.704	0.000	0.001 ***
	10	27.37	366.196	0.002	0.930	0.000	0.000 ***
Acetate (5.0 pH)	5	13.68	400.840	0.002	0.987	0.000	0.000 ***
	1	2.74	331.808	0.002	0.747	0.000	0.000 ***
	10	27.37	33.336	0.021	0.999	0.000	0.000 ***
Phosphate (7.0 pH)	5	13.68	32.961	0.021	0.998	0.000	0.000 ***
	1	2.74	332.492	0.002	0.748	0.000	0.000 ***
	10	27.37	25.161	0.028	0.994	0.001	0.000 ***
Carbonate (9.0 pH)	5	13.68	20.936	0.033	0.962	0.003	0.001 ***
_	1	2.74	14.800	0.047	0.829	0.011	0.012 *

⁻⁻⁻

²⁶² *Table 9-4* - *Decay rates for amoxicillin in each of the solutions provided at -20°C.*



Figure 9-10 - The top figure plots the half-lives of amoxicillin in each of the conditions. Each bar represents a different concentration of amoxicillin used in the experiment. The bottom figure plots of the calculated k' values. The error bars in the bottle figure represent the standard error within the regression analysis used to calculate k'.



Figure 9-9 - Amoxicillin decay at -20°C.

The Y axis for each figure measures the percent change from the initial concentration. The x-axis measures time in days. The trend lines measure exponential decay.

267 9.3.1.2 L-Cysteine Decay

The results from section 8.3.1 reported that when in a solution of 8mM l-cysteine, 10mg/L (2.74 mM) amoxicillin decayed to the LOQ so rapidly an accurate measurement was not able to be taken. It was suspected that this was the result of an interaction between the amoxicillin and the l-cysteine. As such, this experiment measured changes in the decay rate of amoxicillin with varying concentrations of amoxicillin.

273 The decay rate of amoxicillin was observed to increase with an increase in the concertation of 1-274 cysteine in solution (Table 8-5, Figure 8-11). When l-cysteine solution was at a concentration of 275 0.125 mg/l the half-life of amoxicillin was 452.05 hours, which was comparable to the decay rates 276 observed in MilliQ water the 35°C degradation conditions (for example, 10 mg/L of amoxicillin which decayed in milliQ water with a half-life of 393.85 hours). However, the rate of amoxicillin 277 decay increased with increased 1-cysteine concertation. When the 1-cysteine solution was above 278 279 2mg/l, amoxicillin concentrations were below the limit of detection and were not able to be reliably calculated. The greatest ratio tested in which amoxicillin was reliably detectable was for 1mg/l of 1-280 281 cysteine at a ratio of 0.3x the l-cysteine to amoxicillin. The shape of the decay curves was not linear 282 when the natural log of amoxicillin was plotted against time, suggesting that this relationship was 283 not a pseudo first order relationship, as was found in previous sections. This is likely because, unlike 284 in previous sections, the concentrations of l-cysteine solutions were not in excess of amoxicillin. It 285 is possible that the amoxicillin and l-cysteine interact with eachother during decay in a second order 286 reaction.

Concentration of L-Cysteine		ration of Amoxicillin							
mg/L	uM	uM	LC:Amox	K'	Standard Error	R ²	Half-life (hour)	P-va	alue
0.125 ppm	0.10	2.74	0.04	0.002	0.00	0.26	452.95	0.13	
0.25 ppm	0.21	2.74	0.08	0.003	0.00	0.33	208.03	0.08	
0.5 ppm	0.41	2.74	0.15	0.018	0.00	0.79	38.34	0.00	***
1 ppm	0.83	2.74	0.30	0.035	0.01	0.74	19.65	0.00	**
2 ppm	1.65	2.74	0.60	0.192	0.04	0.91	3.61	0.01	*
4 ppm	3.30	2.74	1.21	0.188	0.04	0.90	3.68	0.01	*
8 ppm	6.60	2.74	2.41]	Below Detecti	on		

287 *Table 9-5* – *Decay rates for amoxicillin at different concentrations of l-cysteine.*



Figure 9-11 – Summary of l-cysteine decay.

(A) The decay curves for each of the different l-cystine concentrations. The Y axis represents the natural log amoxicillin concentration (mg/L) for each time point. (B) The half-life of amoxicillin for each l-cysteine concentration used. (C) The k' values for the decay of amoxicillin in different l-cystine values. The error bars in the bottle figure represent the standard error within the regression analysis used to calculate k'.

291 **9.4 DISCUSSION**

292 9.4.1 The decay of amoxicillin in anaerobic medium components at 35°C

293 This study found that the decay rate of amoxicillin was different when in the presence of each of the 294 individual components of anaerobic culture medium. Previous research into the rate of decay of 295 amoxicillin found a relationship between the pH of the buffer used and the rate of decay for 296 amoxicillin (Vahdat and Sunderland, 2009). However, these studies did not consider how the 297 chemical properties of the buffer medium might influence decay rate. Both enzymatic and abiotic 298 decay amoxicillin through a nucleophilic substitution on the carbonyl group of the beta-lactam (Hirte 299 et al., 2016; Zango et al., 2019). When decay is abiotic, the rate of this reaction is determined by the 300 concentration of the reactant and its nucleophilicity (King et al., 2017). Nucleophiles are attracted to 301 positively charged electrophilic sites within a molecule (Figure 8-12). The nucleophilic strength of 302 a reactant is often correlated to the pKa of the deprotonated base of its acid base couple. For example, 303 the decay rate of amoxicillin increases with a rise in pH, as the concentration of nucleophilic 304 hydroxide ions also increases (Chadha, Kashid and Jain, 2003). However, the different chemical 305 structures, even at the same pH can have differing nucleophilicity. It is possible that the differences 306 in the decay rate of amoxicillin found in this study are related to the differences in the nucleophilicity 307 and pKa of each of the medium components tested.



Figure 9-12 - The nucleophilic decay of the beta-lactam ring.

308

The nucleophilicity, or willingness of a nucleophile to donate its electrons, is determined by various
conditions such as charge, electronegativity, and steric hinderance (Hemming, 2001; Fu, 2017;
Hamlin, Swart and Bickelhaupt, 2018) (Figure 8-13). As atoms or molecules become more electron

rich their ability to donate electrons increases. Therefore, the negatively charged molecular forms
and conjugate bases are better nucleophiles than their neutrally charged or protonated counterparts.
Similarly, there is an inverse relationship between an increase in electronegativity and how tightly
bound an electron pair is. Therefore, as the electronegativity increases the nucleophilicity decreases.
Additionally, steric hindrance restricts the access to the protons, and therefore slows the reaction rate
and effective nucleophilicity as well (Edwards and Pearson, 1962; Hemming, 2001; Mayr and Ofial,
2008).

Charge – The conjugate base is a stronger nucleophile

$$HO: > H_2O: HS: > H_2S: H_2S: H_3$$

Electronegativity - Increasing electron density decreases nucleophilicity



Steric Hinderance – the bulkiness of a molecule reduces accessibility and slows the rate of reactions and nucleophilicity



Figure 9-13- Examples of conditions which can influence nucleophilicity.

319

Acetate has a single site of potential ionisation and is moderately nucleophilic with a pKa of 4.76. At a pH of 7, the acetate ion is dominant, however at a pH of 5, the deprotonated acetic acid ion only makes up 68.9% (**Figure 8-14**). Amoxicillin decayed at a similar rate within each of the acetate buffers with half-lives ranging from 289.96-463.01 hours at a pH of 7 and 354.38-415.20 for a pH of 5. Theoretically, this result should be reversed with accelerated decay with the increased presence of the deprotonated ion. However, Vahdat (2000) has reported that the stability of amoxicillin decay was U-shaped in which lower pH values also resulted in amoxicillin decay, though stability was at
its maximum between a pH of 5 and 7 and produced roughly similar decay rates within an acetate
buffer.

Phosphate and carbonate ions are known to be strong nucleophiles, and each contain multiple active sites. Phosphate has four ionic states, though at a pH of 7.0 the ionic states PO_4H^{2-} and $PO_4H_2^{-}$ are dominant and roughly equal in concentration (**Figure 8-14**). PO_4H^{2-} is only mildly nucleophilic with a pKa of 2.2. However, $PO_4H_2^{-}$ is more nucleophilic with a pKa of 7.2. The highest charged state of PO_4^{3-} has a pKa of 12.4 and is highly nucleophilic, though is absent at a pH of 7.0.

334 Despite the strong nucleophilicity of the phosphate ions, the bulkiness of the tetrahedral shape of the 335 molecule sterically hinders intermolecular interactions and therefore inhibits the reason and therefore 336 the nucleophilicity. Amoxicillin was observed to decay rapidly relative to water with a maximum 337 half-life of 124.64 hours or 5.2 days. The rate of decay of amoxicillin differed between the pH 7.0 338 and pH 9.0 carbonate buffers. When in a pH of 7.0 carbonate solution amoxicillin decays at a similar 339 rate to phosphate, with a half-life of 134.42 hours or 5.6 days. However, when the pH is increased to 340 9 the relative decay rate increased resulting in a maximum half-life of 34.38 hours or 1.4 days. There are three forms of the carbonate ion: CO_3H_2 , CO_3H^2 , and $CO_3^{2^2}$ with a pKa of 6.35 and 10.3 341 342 for the first and second transition respectively. At a pH of 7.0 the fully protonated carbonate CO_3H_2 . 343 the weaker nucleophile, is dominant and makes up 91% of the solution. However, at a pH of 9.0, a deprotonated bicarbonate CO_3H^{-} a stronger nucleophile, is dominant at 84.8% of the solution. The 344 345 increased reactivity of the ions presents at pH 9.0 explains the differences in the decay of the two pH 346 conditions for carbonate buffer.

The most rapid amoxicillin decay observed was amoxicillin in the l-cysteine solution which decayed so rapidly that an accurate measurement was not able to be taken in the initial study as most of the measurements taken were below the limit of detection. Cysteine is a complex amino acid which contains both a carboxylic acid, an amine, and a thiol, each of which are strong nucleophiles. When cysteine is dissolved it does not exist as a neutral species as the thiol, amine, and carboxyl groups can each be ionised (**Figure 8-14**). At a pH of 7.0, the solution was dominated by the

353 $NH_3^+CHSHCOO^-$ ion which comprised 94.8% of the ions. The deprotonated hydroxyl group has a pka of 1.7. deprotonating the thiol has a pKa of 8.39, and deprotonating the amine has a pKa of 10.77. 354 355 As such each are strong nucleophiles and would readily attack the carboxyl group on the beta lactam. 356 Additionally, hydrogen bonding between the lone pairs of electrons on the carbonyl and a protonated 357 amine group would position the two molecules in such a way to increase the reactivity of the reaction. 358 This would explain the extremely rapid rates of decay observed for the decay of amoxicillin in the 359 presence of 1-cysteine. 360 Amoxicillin in NaCl solution was observed to decay with a maximum half-life of 325 hours. This is 361 faster than the decay rate of amoxicillin in water or acetate buffer conditions. While neither chloride

or sodium ions participate in nucleophilic substitutions, chlorine ions have been shown to catalyse nucleophilic substitution reactions within seawater (Elliott and Sherwood, 1993; Hazra et al., 2019). For example, Hazra et al (2019) found that the Cl- ions in sea water were effective catalysis for the oxidation of alcohol to carboxylic acids or ketones and aromatic amines to imines in water. This was done through a combination of radial and ionic mechanisms. It is possible that similar mechanism is operating within this study to decay the alcohols and amines present within amoxicillin.



Figure 9-14 - Species distribution plots for the different components used within the study. Each of the test conditions are indicated in hashed grey bar with the percentage of each species present at that pH indicated with the associated colour. All charts and calculations were made using the CurTiPot-pH and acid-base titration curves: analysis and simulation freeware, version 4.2 (Gutz, 2014).

369 9.4.2 The decay of amoxicillin in anaerobic medium components at -20°C

Previous research has established that the catalytic influence of the buffer medium behaves 370 371 differently when frozen than in a liquid state (Vahdat, 2000; Vahdat and Sunderland, 2009). The 372 Arrhenius equation calculates the change in decay rate constant k with a change in temperature. 373 Research by Vahdat (2007) which assessed changes in the decay rate of amoxicillin in acetate and 374 phosphate buffers indicated that temperatures below freezing can diverge from the rate predicted by 375 the Arrhenius equation. A similar effect was observed in this study as the rate of decay for the 376 phosphate and carbonate buffers was disproportionately shorter when frozen than for amoxicillin in 377 the acetate or MilliO water. Several mechanisms were proposed for this variation. As the phase of 378 liquid changes during freezing, dissolved solids concentrate in the liquid fraction until a point of 379 supersaturation before complete freezing. This would create solute concentration higher than those 380 within the original solution. As the decay rate of the amoxicillin is correlated to the concentration of 381 the other solutes within solution, an increase of concentration would increase the rate of amoxicillin 382 decay as well. It is also possible that differences in the eutectic temperature of a species, the lowest 383 temperature that a solution can reach before freezing, could selectively crystalise or precipitate 384 different buffer constituents under frozen conditions (Vahdat and Sunderland, 2007). This could lead 385 to changes in the pH or concentration also changing the effective rate of decay as different parts of the solution selectively freeze out of the solution causing the remaining components to change their 386 387 relative concentrations.

388 9.4.3 Amoxicillin decay in the presence of l-cysteine

389 In the initial set of analyses (Section 8.3.1) amoxicillin decayed to below the LOQ before the first 390 sampling time point (6h) and therefore accurate quantification was not possible. As the amoxicillin 391 and l-cysteine could potentially interact with each other in a second order reaction, reducing the 392 concentration of l-cysteine in the solutions was hypothesised to also reduce the rate of decay for 393 amoxicillin. This study aimed to determine the maximum ratio at which an accurate measurement for the decay of amoxicillin can be calculated using the LC-UV method of quantification of 394 395 amoxicillin. L-cysteine is generally used in anaerobic media as an oxygen scavenger however, the solutions used in this experiment were not anaerobic. The effect that amoxicillin would have on the 396

397 functional properties of cysteine were not studied. However, if the interaction between amoxicillin and cysteine changed the structure of l-cysteine it is possible that higher concentrations of amoxicillin 398 399 could inhibit the ability for l-cysteine to capture oxygen during culturing, in turn preferentially 400 damaging the methanogenic archaea within the culture. Furthermore, the shape of the decay curves 401 were not linear when the natural log of amoxicillin concentration was plotted against time suggesting 402 that the decay relationship was not a pseudo first order reaction. This is likely due to the fact that, 403 unlike in previous sections, the concentrations of 1-cysteine solutions were not in excess of 404 amoxicillin. It is possible that the amoxicillin and l-cysteine interact with eachother during decay in 405 a second order reaction.

406 9.5 CONCLUSIONS

407 This study demonstrated that the decay of amoxicillin is driven more by the chemistry of the other solutes than by pH alone. This study focused on compounds used in anaerobic culture media. While 408 409 some of the compounds tested included buffers (acetate, phosphate, and carbonate) for controlling 410 the pH during the experiment, the rest were non-buffering components within media. While there are 411 many studies which looked at the change in decay rates with a change in pH, the specifics of these 412 experiments varied in many parameters such as temperature, concentration, and buffer type. As a 413 result, it is difficult to directly compare results from this research. For example, similarly to the work 414 in this study, Chadha, Kashid and Jain (2003) and Vahdat and Sunderland (2007) also measured the 415 decay of amoxicillin in a 7.0 phosphate buffer at 35-37°C. However, each study used different 416 relative concentrations of buffer to amoxicillin. As a result, the calculated decay constant k is 417 different for each study and can cannot be directly compared. Kakimoto and Funamizu (2007) found 418 that ionic strength, measured in eC conductivity, was one of the most important factors responsible 419 for the degradation of amoxicillin, however no correlation between conductivity and rate of 420 amoxicillin decay was found in this study. This is likely due to the fact that in the research conducted by Kakimoto and Funamizu (2007) the changes in conductivity were correlated to differences in 421 concentration, which are known to correlate to changes in rate, while this study measured differences 422 423 conductivity between different compounds.

424 This study focused primarily on the catalytic influence of the compounds found within anaerobic culture medium on the breakdown of amoxicillin and has demonstrated that non-buffering materials 425 can still have a catalytic effect. The compounds used in this study were not exhaustive and other 426 427 media types may also result in the rapid decay of amoxicillin. For example, amino acids, particularly those containing sulphur, such as cystine, cysteine, and methionine, have the potential to aggressively 428 429 degrade the beta lactam ring. Vitamin B12 has been long known to be a strong nucleophile (Bailey, 430 1968). While it is produced by some bacteria and archaea, it is important for the synthesis of DNA 431 and therefore is often added to growth media. Metal ions break beta lactam rings. Though this wasn't 432 studied in this study, this could further impact the decay (Navarro et al., 2003; Alekseev et al., 2006). 433 Essential trace metals such as zinc and iron are often added to growth media (Shelton and Tiedje, 434 1984), which are not nucleophiles themselves, but are known to be catalysts for nucleophilic 435 substitution reactions (Murakami, Yorimitsu and Oshima, 2009; Li et al., 2018).

436 A review of emerging contaminants in municipal wastewater treatment plants reported that 437 amoxicillin and other beta lactams are rarely detected in the influent of water treatment plants (Tran, 438 Reinhard and Gin, 2018). The review believes that this is the result of rapid enzymatic decay by the 439 resident microbes within those systems. However, the work in this study suggests that chemical 440 decay may also be a significant part of that process. It also suggests that chemical pre-treatment of 441 hospital wastewater before its traditional treatment methods could be useful for the reduction in 442 exposure to pharmaceutical waste which is high in antibiotics and the proliferation of antibiotic 443 resistance. For example, the Fenton's reagent is a mixture of ferrous iron (usually in the form of 444 FeSO₄) and hydrogen peroxide (H_2O_2) and can catalyse the oxidation of organic contaminants in 445 wastewater. It has been proposed as a means of decontaminating the pharmaceutical residues and 446 resistant microbes within hospital waste (Berto et al., 2009) and anaerobic wastewater containing 447 amoxicillin (Su et al., 2019).

448 This study shows that the stability of amoxicillin is largely dependant on its surrounding matrix, 449 which implies that its toxicity does as well. Studies which aim to assess amoxicillin activity through 450 culturing should keep this chemical decay in mind. For example, due to the interaction of amoxicillin

451 and the oxygen scavenger l-cysteine, it is possible that high amoxicillin concentrations could 452 deactivate the l-cysteine in turn reducing its effectiveness in removing oxygen in media prepared for 453 obligate anaerobes. In this context it is not the amoxicillin directly inhibiting growth, but rather an 454 interaction between the amoxicillin and the growth media. Therefore, the decay of amoxicillin should 455 be assessed directly as part of the experiment or should be commented on as part of the interpretation 456 of the results if direct measurements are not possible.

457 7

459

Solute	Buffer Concentration	рН	Temperature (C°)	K	t 1/2 (d)	Source
Acetate	2.2 M	4.6	-9.3	0.0263	26.38	Vahdat et al 2007
Acetate	2.2 M	4.6	-7.3	0.0224	30.99	Vahdat et al 2007
Acetate	40 mM	5.0	-20	0.0019	8788.704	This study
Carbonate	40 mM	9.0	-20	0.0208	603.864	This study
Phosphate	40 mM	7.0	-20	0.0275	800.064	This study
Phosphate	1 M	7.0	-13.5	0.0073	94.44	Vahdat et al 2007
Phosphate	1 M	7.0	-9.3	0.0298	23.27	Vahdat et al 2007
Phosphate	1 M	7.0	-7.3	0.0342	20.28	Vahdat et al 2007
Water	N/A	7.0	-20	0.0016	10135.37	This study

460

461 Table 9-7 - Comparison of results between this study and similar studies with temperature ranges from 35 – 43°C.
462 All values from this study report the decay reported for the 10 mg/L conditions.

Solute	Buffer Concentration	рН	Temperature (C°)	K	t 1/2 (h)	Source
Acetate	2.2 M	4.6	35	0.0081	86.00	Vadhat et al 2009
Acetate	2.2 M	4.6	42	0.0146	47.48	Vadhat et al 2009
Acetate	40 mM	5.0	35	0.0018	380.90	This study
Acetate	40 mM	7.0	35	0.0018	406.56	This study
Carbonate	40 mM	7.0	35	0.0057	122.59	This study
Carbonate	40 mM	9.0	35	0.0237	29.92	This study
Glucose	1 M	7.0	35	0.0012	563.31	This study
NaCl	40 mM	7.0	35	0.0025	285.17	This study
Phosphate	N/A	5.0	37	0.0038	183.58	Chadha et al 2003
Phosphate	1 M	7.0	35	0.0154	45.01	Vadhat et al 2009
Phosphate	40 mM	7.0	35	0.0071	103.56	This study
Phosphate	N/A	7.0	37	0.0043	162.10	Chadha et al 2003
Phosphate	N/A	7.0	43	0.0085	81.62	Chadha et al 2003
Phosphate	N/A	9.0	37	0.0387	17.91	Chadha et al 2003
Phosphate	N/A	9.0	43	0.0665	10.42	Chadha et al 2003
Water	N/A	7.0	35	0.0015	464.13	This study

⁴⁵⁷ Table 9-6 - Comparison of results between this study and similar studies below freezing.
458 All values from this study report the decay reported for the 10 mg/L conditions.

1

² Chapter 9 ³ Conclusions and Future Work

4

5 **10.1 CONCLUSIONS**

6 The study of antibiotics within wastewater treatment has been widely reported in the context of large 7 wastewater treatment systems. However, much of the world is reliant on decentralised water treatment technologies which primarily rely on anaerobic communities for biological water 8 9 treatment. Despite being one of the most prescribed antibiotics globally, there is very little research 10 into the effects that amoxicillin has on anaerobic systems. In this research project I explored the 11 effects that an acute exposure to amoxicillin would have on an anaerobic community in terms of 12 methanogenic activity, treatment efficiency, change in the microbial ecology, and the development 13 of antimicrobial resistance genes. Furthermore, I explored the stability of amoxicillin structure within 14 anaerobic culture medium with the aim of measuring the rate of abiotic decay. The primary findings 15 of this dissertation were:

16 Does amoxicillin inhibit methane production and water treatment?

17 Chapter 6 established that for all feed types tested the presence of 0.5 - 100 mg/L amoxicillin resulted in a measurable change in the methanogenic activity, though the sensitivity of the microbes 18 19 depended on the type of substrate used. There was a strong relationship between the addition of 20 amoxicillin and a decrease in the rate of methane production for the cultures glucose and complex 21 broth (a mixture of proteins, lipids, and sugars). In both cases, the presence of 5ppm of amoxicillin 22 and above resulted in a measurable decrease in the rate of methane production. Additionally, an 23 increase in the concentration of amoxicillin was associated with an increase in the lag methane 24 production for both substrates. This suggests that the pathways responsible for degrading sugars and proteins were most sensitive to the presence of amoxicillin. However, a second experiment which exposed anaerobic granules to 1, 10, and 50 mg/L of amoxicillin with glucose as the feed, did not find any statistical differences in the rate of methane production for the different amoxicillin conditions. The differences between these runs were likely due to changes in the community structure resulting from different preparation conditions. Based on sCOD measurements the presence of amoxicillin did not have a strong effect on the ability for culture to treat waste.

31 Does amoxicillin change the microbial community structure of anaerobic systems?

32 This study showed that changes in the anaerobic microbial community structure can be detected 33 within hours after amoxicillin was added to the culture. A decrease in relative community robustness 34 between the samples was observed within 24 hours of exposure to amoxicillin for all tested 35 conditions indicating communities exposed to amoxicillin can be sensitive to further stresses. Changes to the overall community structure resulting from the presence of amoxicillin was 36 observable form the first point at 6 hours from inoculation. In the absence of amoxicillin, the 37 38 community composition and structure shifted with the presence of glucose substrate, which was not seen to the same extent when amoxicillin was present. This suggests, that even if methanogenic 39 40 activity occurs at the same rate with and without amoxicillin, small concentrations (1mg/L and less) 41 can have an effect on the community make up.

42 Does the presence of amoxicillin increase the abundance of antimicrobial resistance genes?

Changes in the resistome were detectable within the first 48 hours. A two-to-three-fold increase in the relative net presence of AMR resistance genes within effluent was detected even when amoxicillin in the influxes was as low as 1 mg/L. Despite net increases in the relative AMR gene counts for T5 (30h) and T6 (36h), the total AMR counts appeared to decrease for T7 (48h). This suggests that while AMR genes might be produced in response to the presence of amoxicillin, once the antibiotic is no longer active (through decay or non-reversable PBP bonding) these genes are no longer selected for.

50 How fast does amoxicillin degrade within a lab setting?

This study demonstrated that the decay of amoxicillin is driven more by the chemistry of the other solutes than by pH alone. The stability of amoxicillin is largely dependent on its surrounding matrix, which implies that its toxicity does as well. Studies which aim to assess amoxicillin activity through culturing should keep this chemical decay in mind. For example, due to the interaction of amoxicillin and the oxygen scavenger 1-cysteine, it is possible that high amoxicillin concentrations could deactivate the 1-cysteine in turn reducing its effectiveness in removing oxygen in media prepared for obligate anaerobes.

10.2 FUTURE WORK

59 This work expanded our understanding of the effects of beta-lactam antibiotics in anaerobic systems.
60 However, there are still many unanswered questions. This section explores some of the questions
61 posed by this research.

62 1. What metabolic pathways are most inhibited by the presence of amoxicillin?

This work demonstrated that the presence of amoxicillin affected both the rate and volume of 63 methane produced as well as changed the community structure of anaerobic granules. Both are 64 evidence that there is disruption to the metabolic pathways within the community. However, the 65 66 details of which pathways were disrupted and where this disruption took place was outwith the scope of this work. There are several potential methods for further studying metabolic pathways identified 67 68 here. One way to fill in the gaps of this work would be to incorporate additional modalities such as 69 metabolomics (targeted / untargeted) in an integrative framework. This is routinely done in human 70 microbiome studies where short chain fatty acids (SCFAs) are typically also analysed. Additionally, rather than taking short read amplicon-based approaches, shotgun metagenomics could be employed 71 72 to recover actual function of these microbial communities. Finally, while this work focused 73 exclusively on changes to the 16s rRNA genes. This method is not able to distinguish between the 74 active and inactive or dead parts of the community. Focusing on the RNA sequencing would only 75 measure parts of the community which hare active and therefore could measure changes in activity 76 in response to the presence of amoxicillin. This is particularly important when studying anaerobic 77 communities. Recent work by De Vrieze at all (2018) found that while there was some overlap

between the DNA and RNA community detected at OTU level for bacteria within the community, a greater difference for the archaea present in that community (De Vrieze et al., 2018). This suggest that when measuring changes to anaerobic communities, DNA alone is unlikely to capture all the changes in activity present. This paired with AMR data and over a longer time scale could be highly elucidating for our understudying of the impact of non-lethal antibiotics within anaerobic communities.

84 85

2. Will a community exposed to amoxicillin "recover" from the exposure and converge with the community distribution of the unexposed community?

86 The work in Chapter 7 suggested that the presence of amoxicillin inhibited community development such that the community makeup of the test conditions was more closely related to the control than 87 to those fed with glucose in the absence of amoxicillin. Given the short duration of the experiment, 88 89 it was not determined if the communities would eventually recover to be closer in composition and 90 robustness to the unexposed conditions once the pressure of amoxicillin exposure ceases, or if these changes would remain in the community make up even after the amoxicillin was no longer acting 91 92 within the community. This could be explored through repeating the experiment with a longer 93 window of observation. This could be done either through repeat batch feeding or through use of a 94 bioreactor. The results of this work could inform knowledge of long-term impacts of acute exposure 95 events.

96 3. How long are the presence of AMR genes within the effluent detectable after a shock 97 exposure to amoxicillin?

98 Results in Chapter 7 found that AMR genes were detectable in the effluent of anaerobic systems.
99 However, the results suggested that the presence of AMR genes was only temporary as the
100 concentrations decreased in the final samples of the experiment. Running a similar test, in which
101 anaerobic granules are exposed to amoxicillin, and then monitored for a longer period of time,
102 possibly with multiple rounds of feeding would be able to determine if the decrease in AMR gene
103 concentrations continues. This would determine of the effluent from anaerobic systems could act as
104 point sources for the spread of AMR genes into the environment.

4. Do other oxygen scavengers degrade amoxicillin at the same rate of l-cysteine? Would 105 other formulas for anaerobic medium change the observed inhibitory effects of 106 amoxicillin on anaerobic cultures? 107 108 The results from Chapter 8 found that the presence of l-cysteine has a strong catalytic effect on the 109 degradation of amoxicillin. Cysteine was used in this anaerobic culture as an oxygen scavenger. Other oxygen scavengers such as sodium sulphite and ascorbic acid are also used in anaerobic 110 culturing. Given that these both have differing molecular structures than cysteine, it is possible that 111 they would interact with amoxicillin in a differently. If so, could changing the composition of the 112 anaerobic medium change the measured toxicity of the amoxicillin? To test this, another decay test 113 114 like the ones conducted in Chapter 8 could determine if these oxygen scavengers have the same decay rate. If they do, running parallel studies in which, an array of amoxicillin concentrations and medium 115 116 comprising the different oxygen scavengers could determine how much the medium itself influenced 117 the measured methanogenic inhibition.

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² Chapter 10

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617	

Appendix A – Gene Targets Used in Screening Chip

	Gene	Target antibiotic s (major)	Forward Primer	Reverse Primer
1	16S rRNA	16S rRNA	GGGTTGCGCTCGTTGC	ATGGYTGTCGTCAGCTCGTG
2	16S rRNA2	16S rRNA	CCTACGGGAGGCAGCAG	ATTACCGCGGCTGCTGGC
3	cfiA	Beta Lactam	GCAGCGTTGCTGGACACA	GTTCGGGATAAACGTGGTGACT
4	blaMOX/blaCMY	Beta Lactam	CTATGTCAATGTGCCGAAGCA	GGCTTGTCCTCTTTCGAATAGC
5	blaOCH	Beta Lactam	GGCGACTTGCGCCGTAT	TTTTCTGCTCGGCCATGAG
6	blaPAO	Beta Lactam	CGCCGTACAACCGGTGAT	GAAGTAATGCGGTTCTCCTTTCA
7	blaVEB	Beta Lactam	CCCGATGCAAAGCGTTATG	GAAAGATTCCCTTTATCTATCTCAGACA A
8	blaROB	Beta Lactam	GCAAAGGCATGACGATTGC	CGCGCTGTTGTCGCTAAA
9	blaOXY	Beta Lactam	CGTTCAGGCGGCAGGTT	GCCGCGATATAAGATTTGAGAATT
10	blaPSE	Beta Lactam	TTGTGACCTATTCCCCTGTAATAGAA	TGCGAAGCACGCATCATC
11	cphA 1	Beta Lactam	GCGAGCTGCACAAGCTGAT	CGGCCCAGTCGCTCTTC
12	bla-L1	Beta Lactam	CACCGGGTTACCAGCTGAAG	GCGAAGCTGCGCTTGTAGTC
13	cfxA	Beta	TCATTCCTCGTTCAAGTTTTCAGA	TGCAGCACCAAGAGGAGATGT
14	conA	Beta		
15	empC/bloDUA	Beta		
15		Beta		
16	blaGES	Lactam Beta	GCAATGTGCTCAACGTTCAAG	GTGCCTGAGTCAATTCTTTCAAAG
17	blaSFO	Lactam Beta	CCGCCGCCATCCAGTA	GGGCCGCCAAGATGCT
18	blaTLA	Lactam Beta	ACACTTTGCCATTGCTGTTTATGT GGAGATAAAGTAACAAATCCAGTTAGAT	TGCAAATTTCGGCAATAATCTTT
19	blaZ	Lactam Beta	ATGA	TGCTTAATTTTCCATTTGCGATAAG
20	blaVIM	Lactam	GCACTTCTCGCGGAGATTG	CGACGGTGATGCGTACGTT
21	pbp5	Lactam	GGCGAACTTCTAATTAATCCTATCCA	CGCCGATGACATTCTTCTTATCTT
22	pbp	Lactam	CCGGTGCCATTGGTTTAGA	AAAATAGCCGCCCCAAGATT
23	mecA	Beta Lactam	GGTTACGGACAAGGTGAAATACTGAT	TGTCTTTTTAATAAGTGAGGTGCGTTAA TA
24	blaCTX-M_5	Lactam	GCGATAACGTGGCGATGAAT	GTCGAGACGGAACGTTTCGT
25	penA	Beta Lactam	AGACGGTAACGTATAACTTTTTGAAAGA	GCGTGTAGCCGGCAATG
26	blaCTX-M_8	Beta Lactam	CGTCACGCTGTTGTTAGGAA	CGCTCATCAGCACGATAAAG
27	blaNDM	Beta Lactam	GGCCACACCAGTGACAATATCA	CAGGCAGCCACCAAAAGC
28	blaACC	Beta Lactam	CACACAGCTGATGGCTTATCTAAAA	AATAAACGCGATGGGTTCCA
29	bla1	Beta Lactam	GCAAGTTGAAGCGAAAGAAAAGA	TACCAGTATCAATCGCATATACACCTAA
30	blaCMY_2	Beta Lactam	AAAGCCTCAT GGGTGCATAAA	ATAGCTTTTGTTTGCCAGCATCA
31	blaCMY_3	Beta Lactam	CTGGCGCATACCTGGATTAC	GCCAGTTCAGCATCTCCCA
32	ampC_cefa	Beta Lactam	CAGGATCTGATGTGGGAGAACTA	TCGGGAACCATTTGTTGGC
33	blaSME	Beta Lactam	GAGGAAGACTTTGATGGGAGGATTG	CGCTATATTGCAATGCAGCAGAAG

34	blaCTX-M	Beta Lactam	CGTACCGAGCCGACGTTAA	CAACCCAGGAAGCAGGCA
35	blaFOX	Beta Lactam	CCTACGGCTATTCGAAGGAAGATAAG	CCGGATTGGCCTGGAAGC
36	blaIMIR	Beta Lactam	AGCCGGACTAGAGCTTCATG	GGCAGAACTCATCATCTGCAAA
37	blaOXA51	Beta Lactam	CGACCGAGTATGTACCTGCTTC	TCAAGTCCAATACGACGAGCTA
38	bla0XY1	Beta Lactam	AAAGGTGACCGCATTCGC	CCAGCGTCAGCTTGCG
39	blaPER	Beta Lactam	GCAAATGAAGCGCAGATGC	GACCACAGTACCAGCTGGTA
40	blaSHV11	Beta Lactam	TTGACCGCTGGGAAACGG	TCCGGTCTTATCGGCGATAAAC
41	blaTEM	Beta Lactam	CGCCGCATACACTATTCTCAG	GCTTCATTCAGCTCCGGTTC
42	blaKPC	Beta Lactam	GCCGCCAATTTGTTGCTGAA	GCCGGTCGTGTTTCCCCTTT
43	beta ccra	Beta Lactam	CACTGGCACGGCGATTGTA	CGGCAGCCAAACCACGATA
44	bl1acc	Beta Lactam	TGTTATCCGTGATTACCTGTCTGG	CTCAGCGAGCCAACTTCAAATA
45	heta B2	Beta	GTAACGCCTACTGGAAGTCCA	CAGCTTCTCCTTGAGAATGCAG
46	blaACT	Beta	AAGCCGCTCAAGCTGGA	GCCATATCCTGCACGTTGG
47	blaB	Beta		GGGATAGTAAACCTGAAACTCGGA
4.9		Beta		
40		Beta		
50		Beta		
50	blaIMI	Beta		
51		Beta		
52		Beta		
53	blaLEN	Lactam Beta	TGTTCGCCTGTGTGTTATCTCC	GCAGCACTITTAAAGGTGCTCAC
54	blaMIR blaBEL-	Lactam Beta	CGGTCTGCCGTTACAGGTG	AAAGACCCGCGTCGTCATG
55	nonmobile blaADC-	Lactam Beta	ATGTCCATGGCACAGACTGTG	CCTGTCTTGTCACCCGTTACC
56	nonmobile	Lactam Beta	GGTATGGCTGTGGGTGTTATTCA	AGGCAAGGTTACCACTTGTATACG
57	ampC_1	Lactam Beta	AACAAAAGATCCCCGGTATGG	ACGCCCGTAAATGTTTTGCT
58	blaCMY_1	Lactam Beta	CCGCGGCGAAATTAAGC	GCCACTGTTTGCCTGTCAGTT
59	ampC_2	Lactam Beta	TCCGGTGACGCGACAGA	CAGCACGCCGGTGAAAGT
60	blaSHV_1	Lactam Beta	TCCCATGATGAGCACCTTTAAA	TTCGTCACCGGCATCCA
61	blaOKP	Lactam Beta	GCCGCCATCACCATGAG	GGTGACGTTGTCACCGATCTG
62	blaOXA10_1	Lactam Beta	CGCAATTATCGGCCTAGAAACT	TTGGCTTTCCGTCCCATTT
63	cphA_2	Lactam Beta	GTGCTGATGGCGAGTTTCTG	GGTGTGGTAGTTGGTGTTGATCAC
64	blaCMY2	Lactam Beta	GCGAGCAGCCTGAAGCA	CGGATGGGCTTGTCCTCTT
65	fox5	Lactam Beta	GGTTTGCCGCTGCAGTTC	GCGGCCAGGTGACCAA
66	ampC_3 blaOXA1/blaOXA	Lactam	CCGCCCAGAGCAAGGACTA	GCTCGACTTCACGCCGTAAG
67	30	Lactam Beta	CGGATGGTTTGAAGGGTTTATTAT	TCTTGGCTTTTATGCTTGATGTTAA
68	blaCTX-M_1	Lactam	GGAGGCGTGACGGCTTTT	TTCAGTGCGATCCAGACGAA ATCGGATTATAGTTAACCAGGTCAGATT
69	blaCTX-M_2	Lactam	GCCGCGGTGCTGAAGA	Т
70	blaCTX-M_3	Lactam	CGATACCACCACGCCGTTA	GCATTGCCCAACGTCAGATT
71	blaCTX-M_4	Lactam	CTTGGCGTTGCGCTGAT	CGTTCATCGGCACGGTAGA

72	blaIMP_1	Beta Lactam	AACACGGTTTGGTGGTTCTTGTA	GCGCTCCACAAACCAATTG
73	blaSHV 2	Beta Lactam	CTTTCCCATGATGAGCACCTTT	TCCTGCTGGCGATAGTGGAT
74	blaTEM 1	Beta Lactam	AGCATCTTACGGATGGCATGA	TCCTCCGATCGTTGTCAGAAGT
75	blaCTX-M 6	Beta Lactam	CACAGTTGGTGACGTGGCTTAA	CTCCGCTGCCGGTTTTATC
76	pbp2x	Beta Lactam	TTTCATAAGTATCTGGACATGGAAGAA	CCAAAGGAAACTTGCTTGAGATTAG
77	blaPER 1	Beta Lactam	TGCTGGTTGCTGTTTTTGTGA	CCTGCGCAATGATAGCTTCAT
78	blaIMP 2	Beta Lactam	ΔΔGCCΔCCΔTTTCCTCTCΔTTTT	GGATAGATCGAGAATTAAGCCACTCT
70		Beta		
79	ampt_4	Beta		
80	ampC_5	Beta		CAGUGAGUULAUTTUGA
81	blaOXA58	Lactam Beta	GCAATTGCCTTTTAAACCTGA	CTGCCTTTTCAACAAAACCC
82	blaKPC_3	Lactam	CAGCTCATTCAAGGGCTTTC	GGCGGCGTTATCACTGTATT
83	blaSHV_3	Lactam	GCGTTATTTTCGCCTGTGTA	AGGTGCTCATCATGGGAAAG
84	blaCTX-M_7	Beta Lactam	CGATGTGCAGTACCAGTAA	GCAATGGGATTGTAGTTAA
85	blaIMP_3	Beta Lactam	GGAATAGAGTGGCTTAATTC	GGTTTAACAAAACAACCACC
86	pbp2b	Beta Lactam	AGACGGTAACGTATAACTTTTTGAAAGA	GCGTGTAGCCGGCAATG
87	blaKPC_2	Beta Lactam	GCCGCCGTGCAATACAGT	GCCGCCCAACTCCTTCA
88	bla-SME	Beta Lactam	AACGGCTTCATTTTGTTTAG	GCTTCCGCAATAGTTTTATCA
89	ampC 6	Beta Lactam	GCAGCACGCCCCGTAA	TGTACCCATGATGCGCGTACT
90	blaACT 1	Beta Lactam		CACTATCTCCTCACCCCATCCT
01	blaACT 2	Beta		
91	juint DidACI_2	Beta		
92		Beta		
93	blaUXA48	Beta	TGTTTTTGGTGGCATCGAT	GTAAMKATGUTTGGTTUGU
94	blaSFC	Lactam Beta	GGCTTACTGAACCAGCGAATTC	GGTCCAGACGAAATACATTGTCAC
95	blaGIM	Lactam	GACGACGAATTCACACTGGGAA	GAACTTCCAACTTTGCCATGCC
96	blaKHM	Lactam	GACCTACGCATCGACCCA	GCCGAGATTGCCCAAGC
97	blaSIM	Lactam	TCCAGGCCCAGGACACA	GAGTTTCAATAGTGATGCGTCTCC
98	blaSPM-45	Beta Lactam	ATGAAGCCGAAGAAAGTAGTAGCC	CAGGATGGGAACTCAGAATCCTTC
99	blaTMB	Beta Lactam	AAGTCATACAGCCGGGTGGA	GCTTCCCGCCAGCTCATAC
10 0	blaSFH	Beta Lactam	GGTATCTTTGTGTACTTCCCAGCA	GCAATGATCGAGTCGACTTTAAGC
10 1	blaAIM	Beta Lactam	GAGATCGCCACATGAAACGTC	GCGCATGTTGGCCAGGA
10		Beta		
10	blaUXA23	Beta	GUUGUGUAAATAUAGAATATG	
3 10	blaOXA24	Lactam Beta	GGCATTGTCAGCAGTTCCAGTATA	AGGTAATCGGTTATGTGCAAGGTC
4 10	blaOXA48_2	Lactam Beta	AATAGCTTGATCGCCCTCGA	TTTGGCGGGCAAATTCTTGA
5	blaOXA51_2	Lactam	AATGATCTTGCTCGTGCTTCGA	CATAGCATCGCCTAGGGTCATG
6	blaOXA54	Lactam	GTGATAGTGCTTTGGAACGAGAAC	GCCGCGATATCCCGAGTC
10 7	blaOXA55	Beta Lactam	GCCTTATTGCGTTGGAAACCG	GCCAACTGCTGATATACAGGCA
10 8	blaOXA58_2	Beta Lactam	GTCGTATTGGTCCAAGCTTAATGC	TCTGCGCTCTACATACAACATCTC
10 9	intI1_2	Integrons	CGAAGTCGAGGCATTTCTGTC	GCCTTCCAGAAAACCGAGGA

11 0	intI1_1	Integrons	CGAACGAGTGGCGGAGGGTG	TACCCGAGAGCTTGGCACCCA
11 1	intl3	Integrons	CAGGTGCTGGGCATGGA	CCTGGGCAGCATCACCA
11 2	intI1_3	Integrons	GCCTTGATGTTACCCGAGAG	GATCGGTCGAATGCGTGT
11 3	intI3_1	Integrons	GCCACCACTTGTTTGAGGA	GGATGTCTGTGCCTGCTTG
11 4	intI2_2	Integrons	TGCTTTTCCCACCCTTACC	GACGGCTACCCTCTGTTATCTC
11 5	intI3_2	Integrons	GCCACCACTTGTTTGAGGA	GGATGTCTGTGCCTGCTTG
11 6	intI1_4	Integrons	CTGGATTTCGATCACGGCACG	ACATGCGTGTAAATCATCGTCG
11 7	pmrA	MDR	TTTGCAGGTTTTGTTCCTAATGC	GCAGAGCCTGATTTCTCCTTTG
11 8	acrB_1	MDR	AGTCGGTGTTCGCCGTTAAC	CAAGGAAACGAACGCAATACC
11 9	acrF	MDR	GCGGCCAGGCACAAAA	TACGCTCTTCCCACGGTTTC
12 0	adeA	MDR	CAGTTCGAGCGCCTATTTCTG	CGCCCTGACCGACCAAT
12 1	cmr	MDR	CGGCATCGTCAGTGGAATT	CGGTTCCGAAAAAGATGGAA
12 2	acrA_1	MDR	GGTCTATCACCCTACGCGCTATC	GCGCGCACGAACATACC
12 3	emrD_1	MDR	CTCAGCAGTATGGTGGTAAGCATT	ACCAGGCGCCGAAGAAC
12 4	mdtE	MDR	CGTCGGCGCACTCGTT	TCCAGACGTTGTACGGTAACCA
12 5	mexA	MDR	AGGACAACGCTATGCAACGAA	CCGGAAAGGGCCGAAAT
12 6	emrB/qacA_1	MDR	CTTTTCTCTAACCGTACATTATCTACGAT AAA	AGAACGTAGCGACTGATAAAATGCT
12 7	mtrE	MDR	CGATGTGTCGTTTTGGAAGGT	CCTGCACCATGATTCCTCAATA
12 8	oprD	MDR	ATGAAGTGGAGCGCCATTG	GGCCACGGCGAACTGA
12 9	ttgA	MDR	ACGCCAATGCCAAACGATT	GTCACGGCGCAGCTTGA
13 0	mepA	MDR	ATCGGTCGCTCTTCGTTCAC	ATAAATAGGATCGAGCTGCTGGAT
13 1	mexE	MDR	GGTCAGCACCGACAAGGTCTAC	AGCTCGACGTACTTGAGGAACAC
13 2	cfr	MDR	GCAAAATTCAGAGCAAGTTACGAA	AAAATGACTCCCAACCTGCTTTAT
13 3	mexB	MDR	CTGGAGATCGACGACGAGAAG	GAAATCGTTGACGTAGCTGGAA
13 4	mdsA	MDR	CGGAGTCCATCGACCATTTG	ATCGTCGGCAAGGAGAATCA
13 5	tolC_2	MDR	CAGGCAGAGAACCTGATGCA	CGCAATTCCGGGTTGCT
13 6	acrR_1	MDR	GCGCTGGAGACACGACAAC	GCCTTGCTGCGAGAACAAA
13 7	marR_3	MDR	GCTGTTGATGACATTGCTCACA	CGGCGTACTGGTGAAGCTAAC
13 8	oqxA	MDR	GAGTCAACCTACCTCCACTATCA	GCTGCGAGTTATCCAGCAG
13 9	adel	MDR	CAGTCTGGTTTGCAGTAACCA	CACTCCTACAACAACAGGCAA
14 0	bexA/norM	MDR	TCGGGCATCCCGTTTATGATC	GTAGGCTGCGCATAATACCCA
14 1	mdtA	MDR	ACAAGCCCAGGGCCAAC	CCTTAATGGTGCCTTCGGTTTC
14 2	mdtH	MDR	ATGCTGGCTGTACAAGTGATG	CACTCCAGCGGGCGATA
14 3	cefa_qacelta	MDR	TAGTTGGCGAAGTAATCGCAAC	TGCGATGCCATAACCGATTATG
14 4	qacA/B	MDR	AAGGGCCACTGCATTAGCTG	CCAGTCCAATCATGCCTGCA
14 5	qacF/H	MDR	CTGAAGTCTAGCCATGGATTCACTAG	CAAGCAATAGCTGCCACAAGC
14 6	arsA	MDR	CAGGTCAGCCGCATCAACC	GCCTGAAACACGGCAATTTCTTC
14				

14 8	copA	MDR	TGCACCTGACVGGSCAYAT	GVACTTCRCGGAACATRCC
14 9	czcA	MDR	GCCTTGTTCATCGGCGAAC	GGCAATGTCGCCTTCGTTC
15 0	pbrT	MDR	GATGCGCACTGGGCTTG	TCGGAATATGCGGAAATGCG
15 1	рсоА	MDR	TGGCGTATGGAGTTTCAATGC	GAATAATGCCGTGCCAGTGAA
15 2	sugE	MDR	CTTAGTTATTGCTGGTCTGCTGGA	GCATCGGGTTAGCGGACTC
15 3	tcrB	MDR	GTGCCGGAACTCAAGTAGCA	GCACCGACTGCTGGACTTAA
15 4	terW	MDR	TCAAAGAGCTACGCGAGTCATA	CCTTCCCTGTGGACTCACC
15 5	acrB_2	MDR	TGGTAGTGGGCGTCATTAACAC	GGCAACGTAATCCGAAATATCC
15 6	bexA	MDR	GCGGATCTCTGGTCAGCAA	TGATTGATGGTTCCCCGTACA
15 7	cmeA	MDR	GCAGCAAAGAAGAAGCACCAA	AGCAGGGTAAGTAAAACTAAGTGGTAA ATCT
15 8	mdtA_1	MDR	CCTAACGGGCGTGACTTCA	TTCACCTGTTTCAAGGGTCAAA
15 9	mdetl1	MDR	ATACAGCAGTGGATATTGGTTTAATTGT	TGCATAAGGTGAATGTTCCATGA
16 0	mdtG_1	MDR	TGGCACAAAATATCTGGCAGTT	TTGTGTGGCGATAAGAGCATTAG
16	yceL/mdtH	MDR	TCGGGATGGTGGGCAAT	CGATAACCGAGCCGATGTAGA
16 2	mdtH_1	MDR	CTGCCGTTAAATGGATGTATGC	ACTCCAGCGGGCGATAGG
16 3	mexD	MDR	TTGCCACTGGCTTTCATGAG	CACTGCGGAGAACTGTCTGTAGA
16 4	qасН	MDR	CATCGTGCTTGTGGCAGCTA	TGAACGCCCAGAAGTCTAGTTTT
16 5	sdeB_1	MDR	CACTACCGCTTCCGCACTTAA	TGAAAAAACGGGAAAAGTCCAT
16 6	mtrD_1	MDR	CCGCCAAGCCGATATAGACA	GGCCGGGTTGCCAAA
16 7	oprJ	MDR	ACGAGAGTGGCGTCGACAA	AAGGCGATCTCGTTGAGGAA
16 8	mexF	MDR	CCGCGAGAAGGCCAAGA	TTGAGTTCGGCGGTGATGA
16 9	mtrC	MDR	GGACGGGAAGATGGTCCAA	CGTAGCGTTCCGGTTCGAT
17 0	acrA_2	MDR	TACTTTGCGCGCCATCTTC	CGTGCGCGAACGAACAT
17 1	emrD_2	MDR	TTTAGGCAGCCTCGCTTCA	CCGAATCCAAATAAAACCCAATAA
17	qacA/qacB	MDR	TGGCAATAGGAGCTATGGTGTTT	AAGGTAACACTATTTTCGGTCCAAATC
17	tolC_1	MDR	GCCAGGCAGAGAACCTGATG	CGCAATTCCGGGTTGCT
17	emrB/qacA_2	MDR	GCAGTAGAAGGAACGATTGTTAGTACAG	TGCGTAAACCCAGCTAACAAGTT
17	qac	MDR	GGAGATTTAGCTCATGTAGCTGAAGAA	AAGCTGTTTTATCCCCGTAGCTTTA
17 6	mdtF	MDR	CCACCATCGGGCTTTCC	CCCTTCTTTCTGCATCATCTCA
17	acrA_3	MDR	CAACGATCGGACGGGTTTC	TGGCGATGCCACCGTACT
17 8	mdtG_2	MDR	TTATCTGTTTTCTGCTCACCTTCTTTT	GCGTGGTGACAAACAGGCTTA
17 9	mdtH_2	MDR	CGCGTGAAACCTTAAGTGCTT	AGACGGCTAAACCCCATATAGCT
18 0	mtrD_2	MDR	TGCGCGTAGTCGTTCATCTC	CGTTCCAATTTCCTGATGATTG
18 1	mtrD_3	MDR	GGTCGGCACGCTCTTGTC	TGAAGAATTTGCGCACCACTAC
18 2	acrA_4	MDR	CAGACCCGCATCGCATATT	CGACAATTTCGCGCTCATG
18 3	acrA_5	MDR	CGTGCGCGAACGAACA	ACTTTGCGCGCCATCTTC
18 4	qacG	MDR	CAATAATAACCGAAATAATAGGGACAAG TT	AATAAGTGTTCCTAGTGTTGGCCATAG
18 5	sdeB_2	MDR	GGCATGCAGAAAGTGTTTATGC	TTAAGTGCGGAAGCGGTAGTG

18 6	acrR_2	MDR	GATGATACCCCCTGCTGTGAGA	ACCAAACAAGAAGCGCAAGAA			
18 7	marR_1	MDR	GCGGCGTACTGGTGAAGCTA	TGCCCTGGTCGTTGATGA			
18 8	marR_2	MDR	TCTGGCGTTAGCTTCACCAGTAC	GTGCAAAGGCTGGATCGAA			
18 9	acrR_3	MDR	TGCAACACGCGCTTTCTC	ACGATTGCGGGCAGGTT			
19 0	Tp614	MGE	GGAAATCAACGGCATCCAGTT	CATCCATGCGCTTTTGTCTCT			
19 1	IS613	MGE	AGGTTCGGACTCAATGCAACA	TTCAGCACATACCGCCTTGAT			
19 2	tnpA_1	MGE	GCCGCACTGTCGATTTTTATC	GCGGGATCTGCCACTTCTT			
19 3	tnpA_2	MGE	CCGATCACGGAAAGCTCAAG	GGCTCGCATGACTTCGAATC			
19 4	tnpA_3	MGE	GGGCGGGTCGATTGAAA	GTGGGCGGGATCTGCTT			
19 5	tnpA_4	MGE	CATCATCGGACGGACAGAATT	GTCGGAGATGTGGGTGTAGAAAGT			
19 6	tnpA_5	MGE	GAAACCGATGCTACAATATCCAATTT	CAGCACCGTTTGCAGTGTAAG			
19 7	tnpA_6	MGE	TGCAGATGGTTTAACCTTGGATATTT	TCGGTTCATCAAACTGCTTCAC			
19 8	tnpA_7	MGE	AATTGATGCGGACGGCTTAA	TCACCAAACTGTTTATGGAGTCGTT			
19 9	trfA	MGE	ACGAAGAAATGGTTGTCCTGTTC	CGTCAGCTTGCGGTACTTCTC			
20 0	orf37-IS26	MGE	GCCGGGTTGTGCAAATAGAC	TGGCAATCTGTCGCTGCTG			
20 1	ISPps	MGE	CACACTGCAAAAACGCATCCT	TGTCTTTGGCGTCACAGTTCTC			
20 2	IS1247_2	MGE	TGGATCGACCGGTTCCAT	GCTGACCGAGCTGTCCATGT			
20 3	ISAba3	MGE	TCAGAGGCAGCGGTATACGA	GGTTGATTCAGTTAAAGTACGTAAAAC TTT			
20 4	ISEfm1	MGE	AGGTGTCCATGACGTGAAAGTG	TCCTTTGTCCCCTAGGATATTGG			
20 5	IS1111	MGE	GTCTTAAGGTGGGCTGCGTG	CCCCGAATCTCATTGATCAGC			
20 6	IS1133	MGE	GCAGCGTCGGGTTGGA	ACGCGTTCGAACAACTGTAATG			
20 7	Tn5	MGE	TCAGAGGCAGCGGTATACGA	GGTTGATTCAGTTAAAGTACGTAAAAC TTT			
20 8	IncN_rep	MGE	AGTTCACCACCTACTCGCTCCG	CAAGTTCTTCTGTTGGGATTCCG			
20 9	IncN_oriT	MGE	TTGGGCTTCATAGTACCC	GTGTGATAGCGTGATTTATGC			
21 0	IncP_oriT	MGE	CAGCCTCGCAGAGCAGGAT	CAGCCGGGCAGGATAGGTGAAGT			
21 1	IncQ_oriT	MGE	TTCGCGCTCGTTGTTCTTCGAGC	GCCGTTAGGCCAGTTTCTCG			
21 2	IncW_trwAB	MGE	AGCGTATGAAGCCCGTGAAGGG	AAAGATAAGCGGCAGGACAATAACG			
21 3	pAMBL	MGE	CAGGCTCTTAATGTGATA	TTATGCTCAATACTCGTG			
21 4	pAKD1	MGE	GGTAAGATTACCGATAAACT	GTTCGTGAAGAAGATGTA			
21 5	cro	MGE	AGATGTTATCGACCACTTCGGA	CCGCTTGGCGATAAGCG			
21 6	EAE_05855	MGE	CCCATCACCGCTGAACTGG	TGGGCGCTGCCATCTAAAC			
21 7	IncHI2-smr0018	MGE	ATAATGATTCACCGGGGTAG	CTTCAGGCTATCGTTTCG			
21 8	Incl1_repl1	MGE	CGAAAGCCGGACGGCAGAA	TCGTCGTTCCGCCAAGTTCGT			
21 9	IncN_korA	MGE	GGAACGTTTGTAYCTTGTATTG	ACTCACTATCTTCTGTTGATTG			
22 0	IS1247_1	MGE	CGGCCGTCACTGACCAA	TCGGCAGGTTGGTGACG			
22 1	IS26_2	MGE	CAATACCTTTGATGGTGGCGTAAG	CTTACGCCACCATCAAAGGTATTG			
22 2	IS200_1	MGE	CCAAATACCGAAGACAAGCGTTC	CCAAACTGCTCGTAAAGCATCAG			
22 3	IS200_2	MGE	GCACACCCGATGGAACTGTAAA	TCGGCGGGATCTCCAGAAG			

22	1621 164-20	MCE		
4 22	1521-15AS29	MGE	GILLGILAGGLALAAGIL	GGGATCGTATCGGCAAGCC
5	IS256	MGE	CTTGCGCATCATTGGATGATGG	AAGAACGGCTCCAATTAAGCGA
22	1026 1	MCE		
22	1526_1	MGE	AIGGAIGAAACCIACGIGAAGGIC	CGGIACITAATCIGICGGIGIICA
7	IS3	MGE	CGGTCTGAGCTTCGGGAA	AGAACTGTCACTCCGGTCTG
22	ISE /IS1102	MCE	TTCTCCAACAATCCCCATCCC	
22	135/131162	MGE	TTCTCGAAGAATCGCCATGGC	GETTTGGATCGCTCCAATCGA
9	IS6/257	MGE	ATATCGTGCCATTGATGCAGAG	ACCATTGCTACCTTCGTTGAAG
23	156100	MCF	CGCACCGGCTTGATCAGTA	CTGCCACGCTCAATACCGA
23	130100	MUL		
1	IS630	MGE	CCGCCACCAGTGTGATGG	TTGGCGCTGACTGGATGC
23	1591	MGE	GGATGCCACTGCTGGTCA	ACAGTGGATACAGTATCTGCTGAG
23	1071	Mul		A chief de l'anti-
3	ISCR1	MGE	ATGGTTTCATGCGGGTT	CTGAGGGTGTGAGCGAG
23 4	ISEcp1	MGE	CATGCTCTGCGGTCACTTC	GACGCACCTTCTTGATGACC
23	- · F			
5	lncF_FIC	MGE	GTGAACTGGCAGATGAGGAAGG	TTCTCCTCGTCGCCAAACTAGAT
23 6	mobA	MGE	GCTTCCCGTAACGAGGTAGT	CCTTGAACGGTATCAGCACG
23				
7	Tn3	MGE	GCTGAGGTGTTCAGCTACATCC	GCTGAGGTAGTCACAGGCATTC
23 8	Tn5403	MGE	AAGCGAATGGCGCGAAC	CGCGCAGGGTAAACTGC
23				
9	traN	MGE	GCTTGGCGGTCAGCAATT	TTAGGAATAACAATCGCTACACCTTTA
0	trbC	MGE	CGGYATWCCGSCSACRCTGCG	GCCACCTGYSBGCAGTCMCC
24		MOR		6 1 6 6 6 1 m 6 6 1 6 6 6 6 6 6 6 6 6 6
1 24	repA	MGE	CCCCCAGGACTTGCGAGCG	GAGGCATGCACGCCGACCA
21	pNI105	MGE	CGCTAAGGATGTTTACAC	CTCAACCGTTCTAGGATT
24		Tetracycli		
24	tetA_2	ne Totracycli	CTCACCAGCCTGACCTCGAT	CACGTTGTTATAGAAGCCGCATAG
4	tetA/B_1	ne	AGTGCGCTTTGGATGCTGTA	AGCCCCAGTAGCTCCTGTGA
24		Tetracycli		
5 24	tetD	ne Totracycli	AATTGCACTGCCTGCATTGC	GACAGATTGCCAGCAGCAGA
6	tetA_1	ne	GCTGTTTGTTCTGCCGGAAA	GGTTAAGTTCCTTGAACGCAAACT
24		Tetracycli		
7	tetD_1	ne Totra such	TGCCGCGTTTGATTACACA	CACCAGTGATCCCGGAGATAA
24 8	tetA/B 2	ne	GCCCAGTGCTGTTGTTGTCAT	TGAAAGCAAACGGCCTAAATACA

Appendix B – Gene Targets Used in Experiment

Forward Primer

Target antibiotics Gene (major) 1 16S rRNA 16S rRNA 2 penA Beta Lactam 3 fox5 Beta Lactam 4 cfxA Beta Lactam 5 blaTMB Beta Lactam 6 blaSHV_1 Beta Lactam blaSFO Beta Lactam 8 blaOXA48 2 Beta Lactam 9 intI1_2 Integrons 10 intI1_1 Integrons intl3 11 Integrons acrA_5 12 MDR 13 acrR_3 MDR 14 arsA MDR 15 IS1111 MGE 16 IS1247_1 MGE IS6100 MGE 17 18 IS630 MGE 19 ISEcp1 MGE 20 orf37-IS26 MGE 21 repA MGE 22 Tn5403 MGE 23 tnpA_2 MGE 24 tnpA_3 MGE 25 tnpA_5 MGE tetQ 26 Tetracycline 27 blaMIR Beta Lactam 28 blaAIM Beta Lactam 29 mepA MDR 30 mexF MDR 31 oprD MDR 32 oprJ MDR 33 tolC_1 MDR 34 trbC MGE 16S rRNA2 16S rRNA 35 36 blaACT Beta Lactam

GGGTTGCGCTCGTTGC AGACGGTAACGTATAACTTTTTGAAAGA GGTTTGCCGCTGCAGTTC TCATTCCTCGTTCAAGTTTTCAGA AAGTCATACAGCCGGGTGGA TCCCATGATGAGCACCTTTAAA CCGCCGCCATCCAGTA AATAGCTTGATCGCCCTCGA CGAAGTCGAGGCATTTCTGTC CGAACGAGTGGCGGAGGGTG CAGGTGCTGGGCATGGA CGTGCGCGAACGAACA TGCAACACGCGCTTTCTC CAGGTCAGCCGCATCAACC GTCTTAAGGTGGGCTGCGTG CGGCCGTCACTGACCAA CGCACCGGCTTGATCAGTA CCGCCACCAGTGTGATGG CATGCTCTGCGGTCACTTC GCCGGGTTGTGCAAATAGAC CCCCCAGGACTTGCGAGCG AAGCGAATGGCGCGAAC CCGATCACGGAAAGCTCAAG GGGCGGGTCGATTGAAA GAAACCGATGCTACAATATCCAATTT CGCCTCAGAAGTAAGTTCATACACTAAG CGGTCTGCCGTTACAGGTG GAGATCGCCACATGAAACGTC ATCGGTCGCTCTTCGTTCAC CCGCGAGAAGGCCAAGA ATGAAGTGGAGCGCCATTG ACGAGAGTGGCGTCGACAA GCCAGGCAGAGAACCTGATG CGGYATWCCGSCSACRCTGCG CCTACGGGAGGCAGCAG AAGCCGCTCAAGCTGGA

Reverse Primer

ATGGYTGTCGTCAGCTCGTG GCGTGTAGCCGGCAATG GCGGCCAGGTGACCAA TGCAGCACCAAGAGGAGATGT GCTTCCCGCCAGCTCATAC TTCGTCACCGGCATCCA GGGCCGCCAAGATGCT TTTGGCGGGCAAATTCTTGA GCCTTCCAGAAAACCGAGGA TACCCGAGAGCTTGGCACCCA CCTGGGCAGCATCACCA ACTTTGCGCGCCATCTTC ACGATTGCGGGCAGGTT GCCTGAAACACGGCAATTTCTTC CCCCGAATCTCATTGATCAGC TCGGCAGGTTGGTGACG CTGCCACGCTCAATACCGA TTGGCGCTGACTGGATGC GACGCACCTTCTTGATGACC TGGCAATCTGTCGCTGCTG GAGGCATGCACGCCGACCA CGCGCAGGGTAAACTGC GGCTCGCATGACTTCGAATC GTGGGCGGGATCTGCTT CAGCACCGTTTGCAGTGTAAG TCGTTCATGCGGATATTATCAGAAT AAAGACCCGCGTCGTCATG GCGGATGTTGGCCAGGA ATAAATAGGATCGAGCTGCTGGAT TTGAGTTCGGCGGTGATGA GGCCACGGCGAACTGA AAGGCGATCTCGTTGAGGAA CGCAATTCCGGGTTGCT GCCACCTGYSBGCAGTCMCC ATTACCGCGGCTGCTGGC GCCATATCCTGCACGTTGG

Appendix C – Statistical Analysis of Methanogenesis and Amoxicillin

ANOVA – Rate ~ Feed type + Amoxicillin

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 '' 1

		Df	Sum Sq	Mean Sq	f value	Pr(>F)	
Run 1	Feed	7	673659	96237	101.808	< 0.001	* * *
	Amox	1	2536	2536	2.682	0.104	
Run 2	Feed	7	1848152	264022	70.012	< 0.001	***
	Amox	1	35363	35363	9.377	0.00276	**

Aov (Rate ~ Amoxicillin) - With amoxicillin as a continuous variable

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 '' 1

Rat e	at								0.5 ppm		1 ppm		2.5 ppm		5 ppm	
	Feed	D f	Sum Sq	Mean Sq	f valu e	Pr(>F)		% of contro I	STDE V	% of contro l	STDE V	% of contro I	STDE V	% of contro l	STDE V	
	Acetate	1	517	517	4.83	0.046 7	*	77%	0.09	55%	0.09	64%	0.04	59%	0.16	
	Ethanol	1	1314	1314. 3	2.84	0.116		123%	0.38	96%	0.72	221%	1.16	193%	1.00	
	Butyrate	1	83.6 6	83.66	3.61 7	0.079 6		181%	0.36	134%	0.15	144%	0.52	194%	0.16	
Run 1	Propionat e	1	212. 3	212.3	3.78 4	0.073 7		50%	0.06	72%	0.50	51%	0.19	38%	0.14	
	Glucose	1	867	867.1	2.31 6	0.152		117%	0.31	129%	0.28	116%	0.05	78%	0.15	
	Broth	1	7904	7904	8.13 7	0.013 6	*	122%	0.25	136%	0.08	95%	0.04	75%	0.08	
	H2/CO2	1	975	975	0.18 2	0.677		78%	0.02	53%	0.37	74%	0.02	77%	0.04	

								10 ppm		25 ppm		50 ppm		100 ppm	
	Feed	D f	Sum Sq	Mean Sq	f valu e	Pr(>F)		% of control	STD EV	% of contr ol	STDE V	% of contr ol	STDE V	% of contr ol	STDE V
	Acetate	1	193	192.6	0.67 8	0.425		66%	0.4 9	32%	0.05	44%	0.17	52%	0.18
	Ethanol	1	819	818.6	1.48 9	0.244		73%	0.4 1	54%	0.12	174%	1.56	137%	1.02
	Butyrate	1	296.3	296.3 3	5.02 1	0.0431	*	150%	0.8 4	129%	0.13	198%	0.94	58%	0.13
Ru n 2	Propionat e	1	7.6	7.585	0.3	0.593		124%	0.8 0	17%	0.05	30%	0.08	26%	0.07
	Glucose	1	8672	8672	11.4 3	0.0049 3	*	127%	0.4 3	83%	0.19	65%	0.26	27%	0.04
	Broth	1	12705	12705	14.1	0.0024 1	*	103%	0.2 8	88%	0.06	48%	0.06	20%	0.05
	H2/CO2	1	10150 8	10150 8	4.51 2	0.0534		92%	0.1 7	80%	0.11	56%	0.42	57%	0.09

Appendix D – Statistical Analysis of Total Methane and Amoxicillin

ANOVA - Volume ~ Feed type + Amoxicillin

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 '' 1

		Df	Sum Sq	Mean Sq	f value	Pr(>F)	
Run 1	Feed	7	2313.2	330.5	153.488	<2e-16	***
	Amox	1	1.5	1.5	0.683	0.41	
Run 2	Feed	7	1639.4	234.19	70.67	< 2e-16	***
	Amox	1	43.6	43.6	13.16	0.00043	***

Aov (Volume ~ Amoxicillin)

Signif. codes:	0 '***'	0.001	'**' 0.01	'*' 0.05	<i>'.'</i> 0.1	. ' ' 1
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			1					05,	nm	1 nnm		2.5 nnm		5 nnm	
								0.5	0.5 ppm		тррп		2.5 ppm		pin
	Feed	D f	Su m Sq	Mea n Sq	f valu e	Pr(> F)		% of contr ol	STDE V						
Ru n 1	Acetate	1	0.2 2	0.22	5.63	0.03	*	95%	11%	87%	17%	79%	20%	58%	17%
	Ethanol	1	0.5 8	0.58	2.79	0.12		107%	47%	74%	16%	135%	13%	136%	20%
	Butyrate	1	0.0 4	0.04	1.50	0.24		126%	32%	120%	16%	98%	36%	146%	9%
	Propiona te	1	0.1 0	0.10	2.94	0.11		110%	18%	100%	61%	49%	10%	65%	37%
	Glucose	1	0.0 5	0.05	0.79	0.39		121%	21%	118%	14%	121%	16%	93%	23%
	Broth	1	0.0 1	0.01	0.07	0.80		107%	22%	108%	16%	106%	10%	106%	6%
	H2/CO2	1	8.9 6	8.96	0.52	0.48		106%	6%	70%	51%	111%	1%	119%	7%

								10 ppm		25 ppm		50 ppm		100 ppm	
	Feed	D f	Sum Sq	Mea n Sq	f valu e	Pr(> F)		% of contr ol	STDE V						
Ru n 2	Acetate	1	0.06	0.06	2.37	0.15		106%	44%	47%	14%	63%	37%	56%	33%
	Ethanol	1	0.11	0.11	0.22	0.65		79%	29%	49%	13%	132%	110 %	99%	57%
	Butyrate	1	0.18	0.18	7.53	0.02	*	100%	20%	63%	11%	74%	33%	48%	10%
	Propiona te	1	0.03	0.03	5.65	0.03	*	125%	15%	63%	16%	95%	9%	63%	10%
	Glucose	1	1.57	1.57	7.23	0.02	*	87%	40%	64%	24%	56%	16%	39%	7%
	Broth	1	4.73	4.73	9.76	0.01	*	163%	7%	161%	3%	89%	28%	50%	2%
	H2/CO2	1	204.4 0	204.4 1	14.5 5	0.00	*	92%	4%	86%	7%	64%	35%	51%	16%