# **Antimicrobial resistance in dairy slurry tanks: a critical**

## 2 point for measurement and control

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### 44 Abstract

45 Waste from dairy production is one of the world's largest sources of contamination from 46 antimicrobial resistant bacteria (ARB) and genes (ARGs). However, studies to date do 47 not provide necessary evidence to inform antimicrobial resistance (AMR) 48 countermeasures. We undertook a detailed, interdisciplinary, longitudinal analysis of 49 dairy slurry waste. The slurry contained a population of ARB and ARGs, with resistances 50 to current, historical and never-used on-farm antibiotics; resistances were associated 51 with Gram-negative and Gram-positive bacteria and mobile elements (ISEcp1, Tn916, 52 Tn21-family transposons). Modelling and experimental work suggested that these 53 populations are in dynamic equilibrium, with microbial death balanced by fresh input. 54 Consequently, storing slurry without further waste input for at least 60 days was 55 predicted to reduce ARB spread onto land, with >99% reduction in cephalosporin 56 resistant Escherichia coli. The model also indicated that for farms with low antibiotic use, 57 further reductions are unlikely to reduce AMR further. We conclude that the slurry tank is 58 a critical point for prevalence and control of AMR, and that measures to limit the spread 59 of AMR from dairy waste should combine responsible antibiotic use, including low total 60 quantity, avoidance of human critical antibiotics, and choosing antibiotics with shorter 61 half-lives, coupled with appropriate slurry storage.

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### 63 Introduction

Antibiotics provided to food-producing animals account for 73% of global antibiotic sales (1), prompting concerns about the selection of antibiotic resistance bacteria (ARB) and genes (ARGs), and their migration from livestock and their environment to humans. ARB and ARGs associated with livestock can enter humans through consumption of animal products, e.g. contaminated meat (2, 3) and dairy (4, 5), or more indirectly, e.g. through land-application of animal waste, which may subsequently infiltrate crops (6, 7) and connected water resources (8, 9). 71 Cattle production comprises 50% of global Livestock Standard Units (10), so has considerable environmental impacts that need to be mitigated (11). There are 72 73 approximately 265 million dairy cows globally, producing high volumes of waste manure, 74 estimated at 3 billion tonnes per year (<u>www.faostat.org</u>). In the UK, the site of this 75 study, dairy farms are estimated to account for 80% (67 million tonnes) of total annual 76 livestock manure production (12), with more cattle waste material applied to soil in 77 England and Wales than swine and poultry combined (13). 78 Antibiotics are routinely administered to dairy cattle for treatment, and, in some cases, 79 prevention of common illnesses, including mastitis and respiratory disease (14-16). 80 Lameness, the most costly disease to UK dairy farms (17), is often prevented with 81 application of antimicrobial metals (copper, zinc) or other chemicals (formalin, 82 glutaraldehyde) in the form of footbaths (18), known to co-select for antibiotic resistance 83 (19, 20). Dairy waste can therefore contain selective and co-selective pressures in the 84 form of mixtures of antibiotics and assorted antimicrobials, as well as ARB, including 85 Extended Spectrum Cephalosporin-Resistant (ESC-R) E. coli (21, 22), and genetic 86 resistance determinants (23, 24). Thus, dairy waste may represent one of the world's 87 most substantial routes for AMR to enter the environment, including onto fields and

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90 To limit the risks of AMR, many countries have introduced responsible use policies, 91 including reducing overall agricultural use of antibiotics (25), or of human critical 92 antibiotics, including 3<sup>rd</sup>/4<sup>th</sup> generation cephalosporins (26). However, antibiotics and 93 other antimicrobials remain necessary for safeguarding animal health and welfare. Thus, 94 other countermeasures are also needed to reduce the transmission or prevalence of ARB 95 and ARGs from dairy waste into the environment. For example, current UK guidelines 96 suggest that storage of solid manure and slurry without fresh input for three months can 97 ameliorate AMR risk (27), but no evidence is provided. Slurry storage is essential in the 98 UK and other countries where dairy cows are housed indoors for large parts of the year,

grasslands used for food production and into water ways.

and where slurry cannot be spread onto land that is frozen or deemed nitrate vulnerable.
Two European studies have assessed storage effects on dairy manure, finding that
certain ARGs increased during storage (28, 29); however, this 'stored' effluent regularly
received fresh input. Contrastingly, a survey of several US dairy farms evaluating a
different set of ARGs did not detect clear storage effects on ARG abundancesHurst,
Oliver (30).

105 Other dairy waste studies took a 'snapshot in time' (31-34), which does not allow for 106 assessment of temporal stability of the resistome and the influence of storage. Factors 107 such as temperature also influence the prevalence of enteric pathogens, indicator 108 organisms and resistance phenotypes during manure storage (35-39). Meanwhile, 109 studies assessing how cattle faecal resistomes respond to contrasting antibiotic 110 management practices generally place emphasis on individual cattle (40-42), with 111 different microbiomes, rather than the collective faecal output of the herd. Liquid-solid 112 separation of manure may also influence the persistence of AMR (43). Therefore, there is 113 a need for detailed longitudinal studies of AMR in dairy slurry and potential mitigations.

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115 This study assessed three key questions about AMR in slurry and its relationship to 116 antibiotic use and slurry storage: (1) does the slurry tank select for or against AMR; (2) 117 how does the resistance content of the slurry tank relate to altered patterns of farm 118 antibiotic use; and (3) can slurry storage help reduce AMR in slurry before application to 119 land? Our interdisciplinary approach combined phenotypic, genomic, and metagenomic 120 microbiological analyses with chemical analyses, antibiotic use records and predictive 121 mathematical models, to provide a temporal evaluation of slurry tank content over six 122 months. This was supplemented by concurrent mini-slurry tank experiments which 123 facilitated the controlled study of isolated slurry. We designed the mathematical model 124 to enable us to study the impact of farm practices that would be impractical or unethical 125 to perform through purely empirical approaches. These included major changes to farm 126 slurry handling, antibiotic reduction to a level that would threaten animal welfare, or the reintroduction of use of human critical 3<sup>rd</sup> or 4<sup>th</sup> generation cephalosporins. Thus, this
study enables the identification of approaches to reduce the spread of AMR into the
environment from an important source of such contamination.

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### 131 Methods

#### 132 Sample site

133 We surveyed a mid-sized, high performance commercial dairy farm in England, housing 134  $\sim$ 200 milking Holstein Friesian cattle at the time of study. Practice at this farm is typical 135 of management methods at high-performance dairy farms, although all farms vary. 136 Milking cattle are housed indoors on concrete, and all excreta are regularly removed 137 from cattle yards by automatic scrapers into a drainage system terminating at the 138 3000m<sup>3</sup> slurry tank. The drainage system also receives used cleaning materials and 139 wash water, used footbath containing zinc and copper, waste milk from cows treated 140 with antibiotics, and rainwater runoff. An automated screw press (Bauer S655 slurry 141 separator with sieve size 0.75 mm; Bauer GmbH, Voitsberg, Austria) performs liquid-142 solid separation prior to the slurry tank. Liquids enter the slurry tank semi-continuously, 143 while solids are removed to a muck heap. Calves, dry cows, and heifers are housed 144 separately from the milking cows. Faeces and urine from calves drain into the common 145 drainage system, whilst dirty straw from calf housing is taken directly to the muck heap. 146 Excess slurry can be pumped to an 8000m<sup>3</sup> lagoon for long term storage. Slurry from 147 either the slurry tank or lagoon is used to fertilise grassland and arable fields.

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### 149 Microbiological sampling, strain isolation, antimicrobial susceptibility

#### 150 *testing and whole genome sequencing*

151 Liquid samples were collected from the slurry tank on 17 dates between May and

152 November 2017 (Table S1). *Escherichia coli* strains were isolated using Tryptone Bile X-

153 Glucuronide (TBX) or MacConkey agar or TBX/MacConkey supplemented with 16 µg ml<sup>-1</sup> 154 ampicillin (AMP), or 2 µg ml<sup>-1</sup> cefotaxime (CTX); or on CHROMagar ESBL<sup>™</sup> agar. Putative 155 *E. coli* isolates were subcultured onto TBX agar or TBX agar supplemented with 2  $\mu$ g ml<sup>-1</sup> 156 CTX. E. coli strains were confirmed using oxidase (22) and catalase tests. Antimicrobial 157 susceptibility testing (AST) using a range of antibiotic discs (Table S2) was carried out 158 on 811 E. coli isolates in accordance with CLSI (44) guidelines. ESC-R E. coli strains 159 were identified by phenotypic resistance profile as putatively *ampC* or CTX type, and 160 confirmed by PCR (22). Presence of Tn21-like mercury resistance transposons within the 161 E. coli isolates was initially screened for by growing isolates on LB agar containing 25 µg 162 ml<sup>-1</sup> HqCl<sub>2</sub>. Their presence was confirmed by PCR (45). Genome assembly of selected 163 ESC-R and mercury resistant E. coli strains using PacBio, was carried out by the Centre 164 for Genomic Research (CGR), University of Liverpool, with methods for library preparation and 165 sequencing as previously described (46) or by Illumina short read WGS by MicrobesNG 166 (University of Birmingham, UK). Genome sequence analysis and annotation was 167 conducted using Prokka (47), CSARweb (48), Snapgene viewer (Insightful Science; 168 snapgene.com), Res Finder (49) and Plasmid Finder (50). Genome sequences are 169 deposited with NCBI under BioProject PRJNA736866. 170

### 171 Metagenomics Sample collection and DNA extraction

- 172
- 173 Main tank Sample Collection

Samples were collected from the slurry tank monthly between June and October 2017, using a clean stainless steel bucket, and aliquoted into 2 large glass bottles with external PE protection. Three replicate extractions were performed on 250 µl of each sample using a PowerFecal Kit (Qiagen), according to manufacturer's instructions (15 extractions in total). DNA was quantified using a Qubit fluorometer (Invitrogen) while quality was assessed via Nanodrop 1000 (ThermoFisher). Extracted DNA was stored at 4°C pending sequencing.

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### 182 Mini-Tank Experiments

183 Miniaturised experimental slurry tanks were set up to assess the impact of storing slurry

- 184 (control tanks) and to measure antibiotic stability. Twelve mini-tanks were situated on
- 185 the farm from 23/4/2018 to 15/6/2018 at ambient temperature (mean 24 hour
- 186 temperature in liquid ranged between 7° to 17°), protected from rain and direct sunlight,
- 187 and containing 10L grab samples of slurry from the surface of the main slurry tank. Six
- 188 different conditions were tested in duplicate (all amounts per litre): control; + SSD
- 189 (0.2mL of slurry solids homogenised by stomacher, including 67 CFU of CTX-resistant E.
- 190 *coli*); + SSD +  $3\mu g$  cefquinome weekly addition; + SSD +  $40\mu g$  cefalexin weekly
- addition; + SSD + 16.8g of footbath mix (Cu + Zn); + SSD + footbath + cefquinome).
- 192 Mini-tanks were sampled four times (0, 2, 4 and 7 weeks after initial filling).
- 193 Experimental conditions were mainly used for model calibration (Supplementary Text 1).
- 194 E. coli were isolated and cultured as described above except MacConkey agar was not
- 195 used. DNA was extracted and processed for sequencing as above. Antibiotic
- 196 concentrations were measured as described previously Baena-Nogueras, Ortori (51) with

Metagenomic sequencing of DNA extracted from the main slurry tank was performed by

- 197 further methods described in Supplementary Text 3.
- 198

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#### 199 Metagenomic Sequencing, Assembly and Analysis

201 Liverpool Genomics using the Illumina HiSeq platform, and from the mini-slurry tanks by 202 Edinburgh Genomics using the Illumina NovaSeq platform (150 bp paired end libraries in 203 both cases). For the main tank, reads were trimmed with Cutadapt v1.2.1 (52) and 204 Sickle v1.2.0.0 (53), while mini-tank reads were trimmed with Fastp v0.19.07 (54). 205 Assembly was performed on trimmed reads using Megahit v1.1.3 (55). Main tank 206 technical replicates were pooled by date and assembled using the settings: k-step 10; k-207 range 27-87. Mini-tank metagenomes were assembled individually (k-step  $\sim$ 20, k-range: 208 21-99). Metagenome sequences are deposited with the ENA under Study Accession 209 PRJEB38990.

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211 Read-based searches for ARGs were performed with DeepARG v2 (56). ARGs were also 212 identified on contigs (>1.5 kb length) in order to investigate the wider genetic context of 213 the core resistome using ABRicate v1.0.1 (57), using MegaRes 2.0 for ARGs and metal 214 resistance genes (MRGs) (58) (including experimentally verified MRGs; genes requiring 215 SNP validation were excluded) and ACLAME 0.4 for MGEs (59). All data were analysed 216 with stringencies: >60% gene coverage, >80% identity(60). Lastly, the BacMet2 217 database (61) was screened against translated peptides (based on Prodigal (62) output) 218 from meta-assemblies of the main and mini-tanks (stringencies: >60% sequence 219 identity and match length >50% of peptide length). 220 Taxonomic assignment of reads was performed using Kaiju v1.6.2 (63), with default settings. The reference database used was a microbial subset of the NCBI database (64), 221 222 including additional fungal and other microbial eukaryotic peptide sequences. Contigs of 223 interest were assigned putative identities using NCBI-nucleotide BLAST (65)

224 (MegaBlast(66), highly similar sequences).

225 For both ARG and taxonomic assignments, statistical comparisons were carried out using

- the DirtyGenes likelihood ratio test (67), using randomized resampling (n=1000) from
- the null distribution to establish p-values.
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#### 229 Water Quality Analysis

230 Water quality analysis was performed on the same samples as for microbiological 231 culturing. For each sample, 2.5L was initially sampled. Probes were used to assess the 232 pH (Hach PHC201), dissolved oxygen (Hach LDO101) and NaCl (Hach). The probe tip 233 was rinsed in Milli-Q water (Merck), dabbed dry and submerged into the bottle 234 containing slurry and left to equilibrate. The sample was then homogenized by shaking 235 vigorously before decanting into a 250mL bottle for analysis using a Hach DR3900 236 Laboratory Spectrophotometer with cuvette test kits: sulphate (LCK153); ammonium 237 (LCK303); chloride (LCK311); copper (LCK329); LATON total nitrogen (LCK338); nitrate 238 (LCK340); nitrite (LCK342); phosphate (LCK348); zinc (LCK360); COD (LCK514); and 239 TOC (LCK381). Standard procedures are available from https://uk.hach.com.

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### 241 Mathematical Model

242 A mechanistic, multi-strain model of AMR in the slurry tank was constructed to simulate 243 a range of relevant farm management scenarios that would have been impractical or 244 unethical to carry out empirically. In brief, it is a coupled ordinary differential equation 245 model of bacterial populations including logistic growth, death (baseline and 246 antimicrobial induced), horizontal transfer and fitness cost of resistance, inflow and 247 outflow (68, 69). The model considered mobile resistance to penicillin, tetracycline, 248 cephalexin, cefquinome, copper, and zinc, and was simulated for a full year in order to 249 capture the recorded input of cephalexin and other antibiotics. The choice of resistances 250 reflects our interests in ESC-R E. coli strains, and the risk of environmental 251 contamination by mobile genes following slurry spreading. Full model description is 252 provided in Supplementary Text 1, equations in Supplementary Text 2 and parameter 253 values in Table S4. This model was deposited in BioModels (70) as MODEL1909100001. 254 The secondary storage model is derived from this model by duplicating equations for 255 each storage vessel (70) and also deposited as MODEL1909120002. A reduced model 256 was used for parameter inference from mini-tank data. Model simulations were carried 257 out in Matlab using the ode45 solver.

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259 Results

260 Resistance to antibiotics with historic, current and no documented farm use 261 The majority of antibiotics administered to milking cows during the sampling period were 262 aminocoumarins, aminoglycosides and beta-lactams delivered in combination, and beta-263 lactams and tetracyclines administered individually (Table S3). The last recorded use of 264 sulphonamides (sulfadoxine) was in June 2016; of first generation cephalosporins 265 (cephalexin) was in April 2017 (shortly before the start of the sampling period); of third 266 generation cephalosporins (ceftiofur) was in January 2016; and of fourth generation 267 cephalosporins (cefquinome) was in August 2015. Residual antibiotics or ARB associated 268 with historical use could potentially be present in sludge at the bottom of the tank that

269 cannot be piped for spreading. Smaller quantities of antibiotics are also given to

270 youngstock; their waste does not enter the slurry system.

The dominant resistance phenotypes of cultured *E. coli* isolates from the slurry tank (Figure 1a) were ampicillin (34.6%), cefpodoxime (39.3%), cefotaxime (29.6%) and streptomycin (26.5%); other common phenotypes included tetracycline (13.6%), chloramphenicol (10.7%) and nalidixic acid (9.6%). Multidrug resistant *E. coli* strains ( $\geq$ 3 different antibiotic classes, Magiorakos, Srinivasan (71)) represented 37% of the cultured isolates (Figure 1b), detected in strains isolated on both antibiotic-

277 supplemented and non-supplemented media. Of these isolates, 12 cefotaxime resistant

278 *E. coli* strains were sequenced to characterize the resistance genes and mobile elements

279 carrying them. Three carried IS*Ecp1* CTX-M-15, additionally carrying *qnrS* and *tetM* 

within the *ISEcp1* element. The other sequenced ESC-R strains were chromosomal *ampC*mutants.

282 In main slurry tank metagenomes, eight resistance classes account for 98% of the ARGs

identified in reads (Figure 1c): multidrug resistance genes (36.7%); tetracycline

resistance genes (21.6%); macrolide-lincosamide-streptogramin (MLS) resistance genes

285 (21.4%); aminoglycosides (7.3%); beta lactams (4.5%); peptides (4.0%); bacitracins

286 (1.6%) and glycopeptides (1.2%). MRGs were also identified (*mer*: mercury; *cop*, *cus*,

287 *pco/sil*: copper, copper/silver; *cad*, *czc*: cadmium, cadmium/zinc/cobalt; *ars*,

arsenic/antimony; *pbr* lead resistance). In equivalent metagenome read assemblies, MLS

and tetracycline ARGs were most frequently detected (70 and 46 contigs, respectively).

Few MRGs were detected in main tank metagenome assemblies, limited to TCR copperresistance genes (5 contigs).

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293 Overall, the identification of aminoglycoside, beta-lactam (excepting 3<sup>rd</sup>/4<sup>th</sup> generation 294 cephalosporins) and tetracycline resistance genes and phenotypes reflect current or 295 recent farm antibiotic use, while the presence of zinc and copper resistance genes reflect 296 transition metal use. The presence of sulphonamide and cephalosporin resistance genes 297 and phenotypes may be due to historical use, or reflect widespread environmental

- 298 occurrence (72). The prevalence of MLS resistance genes is unlikely to be associated
- with antibiotic use, as there is no recorded MLS use for milking cows. 299
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#### 301 Slurry tank properties and AMR remained stable due to frequent inputs

302 Water quality measures were largely stable (Figure 2a), with some fluctuations in July 303 and August likely to be associated with mixing of slurry in the tank prior to spreading on 304 fields. The relative contribution of the dominant drug-resistance categories listed above 305 remained unchanged throughout the sampling period (Figure 1c; p=0.172, DirtyGenes 306 test). Likewise, taxonomic analyses of read data showed the time-stable dominance of six bacterial phyla with at least 1% prevalence (Figure 2b; p=0.254, DirtyGenes test): 307 308 Bacteroidetes (13.8%), Firmicutes (13.7%), Proteobacteria (4.7%), Spirochaetes 309 (2.9%), Euryarcheaota (1.9%) and Tenericutes (1.4%). These phyla only account for 310 38% of the microbial community: there is considerable diversity in the tank with 178 311 phyla identified (Table S4). 312 The overall numbers of *E. coli* identified through culture-based enumeration also 313 remained stable (Figure 2c), with concentrations of  $4.23\pm0.40$  (Log<sub>10</sub> CFU mL<sup>-1</sup>) on TBX

plates and 4.29±0.46 (Log<sub>10</sub> CFU mL<sup>-1</sup>) on MacConkey media. *E. coli* strains resistant to

315 ampicillin (TBX/Amp 16  $\mu$ g mL<sup>-1</sup>) were stable at concentrations of 3.99±0.43 (Log<sub>10</sub> CFU

316 mL<sup>-1</sup>), i.e. ~58% of cultured *E. coli* strains. *E. coli* that could be cultured on cefotaxime

- 317 selective plates (TBX/CTX 2  $\mu$ g L<sup>-1</sup>) were detected on only five of 17 sampling dates, with
- 318 counts below 10 colonies per plate on all but one day (22<sup>nd</sup> August). Thus, cefotaxime
- 319 resistant E. coli were present at low levels, but could not be reliably quantified. The full
- 320 AST profiles of the 811 isolates also show consistency over time, with some random
- 321 variation, both on antibiotic-free and antibiotic-supplemented media (Figure 3).
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#### 323 Model predictions are consistent with microbial data

324 In the mathematical model, predicted resistance to penicillins fluctuated between 0.4%

- 325 and 6.4% and cephalosporins between 0.5% and 7.9% (Figure 4a), i.e. both present but
- 326 low, despite frequent inflow of antibiotics into the tank (Figure 4b). Resistance to

327 tetracycline increased from low initial levels to fluctuate around ~25% of the E. coli 328 population (Figure 4a), before slowly declining over the longer term, reflecting the 329 decline in tetracycline use later in the year. These predicted levels of tetracycline and 330 cephalosporin resistances are concordant with the empirical phenotype above. Penicillin 331 resistance in the model is lower than observed empirically, probably because resistance 332 in the model is plasmid-borne, while many strains have chromosomal mutations of *ampC* 333 or chromosomally located resistance genes that could be mobilised (e.g. ISEcp1CTX-M-334 15 elements). The model predicts that zinc resistance is highly prevalent, rising to 335 fluctuate around 80%, with co-occurrence of tetracycline and zinc resistance, typically 336 fluctuating between 10 and 15%, consistent with predictions that the metal 337 concentrations in the tank are co-selective (69).

#### 338 Associations of ARGs with other ARGs, integrons and Gram-positive taxa

339 Several metagenome contigs contained two or more ARGs, MRGs or associations with 340 MGE markers in both the main tank (37 contigs) and mini-tank metagenome assemblies 341 (101 contigs) (Figures S1 and S2). These include ARGs belonging to the same resistance 342 gene group, e.g. aph3 and aph6 (both aminoglycoside resistance genes; Figure S3a) 343 which were co-localised on five main-tank and eight mini-tank contigs; as well as genes 344 associated with entirely different antibiotic resistance classes, e.g. ant6 and tet44 345 (aminoglycoside and tetracycline resistance, respectively; Figure S3b) were co-localised 346 on two main-tank and eight mini-tank contigs. In other mini-tank contigs, aph3-aph6 347 were additionally co-resident with either a sulphonamide (sul2, 1 contig) or tetracycline 348 (tetY, 1 contig) resistance gene. tetM was embedded within the widely documented 349 Tn916 transposon (18 tetM contigs in total, nine of which were linked with Tn916 350 elements). The two largest Tn916-like contigs (18.3-18.9 kb) appear to be carried within 351 Gram-positive bacteria, possibly Streptococcus spp. or Enterococcus spp. (NCBI-BLAST, 352 ~99.96% identity, ~91% query coverage; Figure S3c). Furthermore, 21.4% (n= 6/28) 353 of main and mini-tank contigs containing *cfxA* (class-A beta-lactamase) were co-localised 354 with mobile elements.

Further identification of mobile resistance cassettes was through a screen of all *E. coli* strains for phenotypic mercury resistance as a marker for Tn21 carriage. Sequence analysis of mercury resistant *E. coli* strains showed that three carried Tn21 variants carrying the integron intI2 conferring co-occurrence of combinations of penicillin, sulphonamide, aminoglycoside and quaternary ammonium compound resistances.

#### **360** Waste management for AMR reduction

361 We investigated the use of slurry storage to ameliorate resistance through a combination 362 of empirical and modelling work. In the mini-tanks, we found that storage of slurry 363 without inflow rapidly decreased the total concentration of cultured E. coli cells (Figure 364 S6a), as well as Escherichia, Pseudomonas and Klebsiella spp. sequences identified by 365 metagenomics (Figure 5). Reads assigned to gut-associated anaerobes belonging to 366 Bacteroidetes including Bacteroides spp., Alistipes spp. and Prevotella spp. declined in 367 steps. In contrast, the relative abundance of *Acinetobacter* spp. gradually increased until 368 week four, before declining again by the end of the experiment (Figure 5). 369 The prevalence of beta-lactam resistance genes declined considerably in <2 weeks 370 (Figure 6a). The overall relative abundance of tetracycline resistance genes declined 371 marginally over 7-weeks of storage (Figure 6b); however, different patterns were 372 observed with different gene groups: tetY (Figure 6c) and tet40 (Figure 6d) declined 373 sharply within two weeks, while others, e.g. tetM (Figure 6e) were maintained in stored 374 slurry. According to BLAST analysis against the NCBI database, mini-tank contigs 375 containing tetY (2 contigs) were likely associated with Gamma-Proteobacteria, while 376 tet40 (6 contigs) was consistently linked to Firmicutes. Similarly, tetM was typically 377 associated with Firmicutes (7 of 16 contigs; >89% sequence coverage, >99% sequence 378 identity), more specifically Bacilli. The proportion of MLS ARGs remained comparatively 379 stable throughout (Figure 6f), consistent with their presence not connected with patterns

380 of MLS use on the farm.

We implemented a two-stage in series storage mathematical model to consider whether the storage of slurry in the main tank, without fresh inputs, would reduce AMR in slurry prior to land application. The model predicted that after only four days of storage, 50% of the amoxicillin- and cefalexin-resistant *E. coli* are removed, and after 60 days of
storage, only 0.29% of cefalexin-resistant and 0.00001% of amoxicillin resistant *E. coli*remained (Figure 7a). However, the model predicts that tetracycline resistant bacteria
will increase over this time by 25% due to ongoing selective pressure and low fitness
cost. Importantly, multidrug resistant *E. coli* become undetectable.

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#### 390 Simulations of altered antibiotic use support criteria for responsible use

Simulations of on-farm antibiotic use (~9.7 mg/Population Correction Unit (PCU) in 2017) result in low levels of penicillin and cephalosporin resistance, consistent with the empirical data. We simulated further reductions in antibiotics entering the tank to either 50% or 10% of current use. Neither reduction had a material impact on either resistance (Figure 7b) but there is a small reduction in tetracycline resistance (33% reduction in resistance at 10% usage) because of the reduced selective pressure for tetracycline resistance.

398 Very few cephalosporin resistant E. coli were detected in the farm samples (detailed 399 above). Thus, we also simulated a return to use of the critically important 4<sup>th</sup> generation 400 cephalosporin (cefquinome) in place of cefalexin (1<sup>st</sup> generation), assuming that 401 cefquinome resistance also confers resistance to cefalexin. After accounting for the lower 402 recommended dosage of cefquinome relative to cefalexin, we found cefquinome use 403 increased resistance to both cefquinome and cefalexin of only 0.65% and 0.35% 404 increase respectively (Figure 7c). To represent high antibiotic use following an outbreak 405 of disease, we simulated 50 mg/PCU of cefquinome used in place of cefalexin. Such a 406 scenario was predicted to select an increase of cefquinome resistance of only 3.55%. 407

408 Discussion

409

#### 410 The slurry tank is a critical measurement and control point for AMR

411 The bacterial community and ARGs in the slurry tank appear to be maintained in a state 412 of dynamic equilibrium, with a balance between input of fresh microorganisms from the 413 cattle, and decline, as observed in the mini slurry-tank experiments. This equilibrium is 414 also evident in the observed stability of the virome of the same tank over the same 415 sampling period (73). The slurry tank maintains an array of ARGs, many of which have 416 been found in other animal wastes. These include MLS genes such as mefA (24, 29, 74) 417 and the *cfxA* group of beta-lactamase genes (24, 74, 75). The association of *cfxA* with 418 Gram-positive organisms suggests that AMR phenotyping should routinely include a 419 Gram-positive sentinel; Enterococcus spp. may be suitable because of their use in water 420 quality analysis (76) and the inclusion of *E. faecium* in the ESKAPE pathogens list (77). 421 Tetracycline resistance genes such as *tetW* and *tetM* have also been frequently found in 422 cattle and swine waste (29, 78, 79). Although present in low quantities relative to other 423 ARGs, tetM has the potential for selection and possible mobilisation (e.g., ISEcp1 or 424 Tn916-like elements). Consequently, the tank appears to be a critical sampling location, 425 representative of the AMR status of the farm as a whole, reflecting current and previous 426 antibiotic use. The presence of resistance genes to antibiotics with no recorded use (e.g. 427 quinolone resistance, MLS genes) are likely to reflect broader environmental, and 428 possibly human, input into the farm microbiome.

429 At a superficial level, the slurry tank appears to meet many criteria presumed to define a 430 'hotspot' for AMR, which cite a high abundance of bacterial populations and the routine 431 presence of antimicrobial residue (80). However, the concept of an AMR 'hotspot', where 432 bacterial and antimicrobial abundance are assumed to lead to increases in AMR 433 prevalence, alongside the related concept of 'reservoir', assumed to represent the 434 nascent AMR genes circulating in the environment poised to be mobilised through 435 antimicrobial exposure, are open to critique (81). Our findings suggest that the tank, 436 rather than generating resistance, can ameliorate resistance, depending on the waste 437 management practice, and that slurry be stored for at least two months without fresh 438 slurry inputs to the system/tank. Thus, the tank is neither a hotspot nor a reservoir, but, 439 if managed appropriately, can be a critical control point for reducing the transmission of 440 ARGs and ARB from livestock into the wider environment.

441

#### 442

#### 443 Agricultural AMR policy should combine responsible antibiotic use with

#### 444 effective waste management

445 Policy and industry guidance to reduce AMR focus on reduced or responsible agricultural 446 antimicrobial use (25, 82, 83), including the cessation of use of human critical 447 antibiotics. Our findings provide evidence in support of responsible use. Simulations of 448 reductions below the already low level of 9 mg/PCU did not predict reductions in 449 penicillin and cephalosporin resistance below current levels. However, reduced 450 tetracycline use led to reduced tetracycline resistance, associated with the environmental 451 stability of this antibiotic, suggesting that prudent antibiotic use could also include 452 antibiotic choice encouraging use for those with shorter half-lives where medically 453 appropriate. While our findings suggested that use of 3<sup>rd</sup> and 4<sup>th</sup> generation 454 cephalosporins did not lead to substantial increases associated resistances, once such 455 resistances are established, relevant genes, e.g. CTX-M, can be selected for by 1<sup>st</sup> 456 generation use. Although UK policy initiatives have greatly reduced the use of 3<sup>rd</sup>/4<sup>th</sup> 457 generation cephalosporins on UK dairy farms, globally their use remains prevalent, e.g. Ceftiofur (3<sup>rd</sup> generation cephalosporin) is routinely used in the US to treat metritis (84, 458 459 85) and mastitis (86). Eliminating the use of these antibiotics in agricultural production 460 should still be an important goal of national and global policies to mitigate the 461 environmental dissemination of AMR (87).

462 A policy focus on antibiotic use is limited because of the need to use antibiotics to treat 463 sick livestock. We also showed that waste management practice provides an additional 464 mechanism to control AMR, by reducing the prevalence of resistance genes and key 465 microbial phyla in slurry prior to soil amendment. Specifically, secondary storage of 466 slurry for a period of 60 days, without fresh inflow, would significantly reduce the levels 467 of ARB within the tank, representing an opportunity for rational farm design and practice 468 to minimize AMR outcomes. This result is also concordant with other practices for 469 mitigating AMR on farms, including the use of anaerobic digestion (79, 88),

470 vermicomposting and solid-liquid separation (43).

471 Two qPCR-based studies surveying Finnish swine and dairy farms reported that storage 472 of animal manure slurry coincided with significant increases in select tetracycline, 473 sulphonamide and aminoglycoside resistance genes when compared to fresh manure 474 (28, 29). However, the farms involved in these studies used storage systems which 475 received regular fresh inflow during the sampling period. Our metagenomic analyses of mini-tanks indicate that in the absence of fresh input a range of ARG classes decline 476 477 (e.g. aminoglycoside and beta-lactam ARGs) or remain relatively stable (e.g. MLS 478 ARGs). Moreover, culture-based results confirm an overall reduction in antibiotic 479 resistant E. coli in slurry stored without inflow. Collectively, this provides empirical 480 evidence supporting existing UK guidelines regarding the storage of slurry without 481 further input as a means of reducing environmental exposure to AMR determinants.

482

#### 483 **Evaluation of co-selection needs alternative approaches**

484 Aminoglycoside, tetracycline and sulphonamide resistance genes were found on the 485 same contigs. The result is consistent with sulphonamide resistance being co-selected by 486 concurrent use of multiple antimicrobials because aminoglycosides and tetracycline were 487 the two antibiotic classes used most during the sampling period. We anticipated finding 488 evidence of co-occurrence of ARGs and MRGs in assembled metagenomic data, in 489 accordance with other studies (19, 24, 89). However, apart from antibiotic resistance 490 associated with Tn21-like elements carrying integrons, we found no evidence for such 491 linkage in the slurry metagenomes or sequenced *E. coli* strains. This lack of evidence 492 might not be evidence of absence of ARG-MRG co-occurrence, as these genes may not 493 necessarily be genetically linked on a chromosome or on plasmids, and yet still be 494 subject to co-selection if they reside in the same cell. Accordingly, the use of long-read 495 or hybrid genome sequencing of strains selected for zinc or copper resistance may be 496 more appropriate for detecting the co-occurrence of ARGs and MRGs (90).

497

### 498 Conclusions

499

500 We have conducted a longitudinal, interdisciplinary study of the dynamics of AMR in a 501 dairy slurry tank. The microbiota was in a state of dynamic equilibrium, with fresh input 502 of bacteria from the animals balanced by natural decay. Antibiotic resistance was 503 maintained, reflecting current and previous veterinary practice, as well as interaction 504 with the broader environment. The slurry tank is therefore both a natural measurement 505 point for on-farm resistance, as well as a control countermeasure point for resistance 506 being released into the wider environment (land and water). The spread of antibiotic 507 resistance into the wider environment through slurry application can be mitigated by a 508 combination of responsible antibiotic use, including low total quantity, avoidance of 509 human critical antibiotics, and antibiotic choice with shorter half-lives, with slurry storage. These approaches can mitigate spread of AMR into the environment from one of 510 511 the world's largest sources of AMR pollution. 512 513 Author Contributions 514 515 Michelle Baker: Methodology, Formal Analysis, Investigation, Writing – Original Draft, 516 Visualization 517 Alexander D Williams: Formal Analysis, Investigation, Data Curation, Writing - Original 518 Draft, Visualization

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- 552 Reviewing and Editing
- 553 Dov J Stekel: Conceptualization, Formal Analysis, Supervision, Writing Reviewing and
- 554 Editing, Project Administration, Acquisition of Funding

555

556 Data Availability

557

- 558 Genome sequences are deposited with NCBI under BioProject PRJNA736866.
- 559 Metagenome sequences are deposited with the ENA under Study Accession PRJEB38990.
- 560 Mathematical models are deposited in BioModels as MODEL1909100001 and
- 561 MODEL1909120002. All other data, including all details of accession numbers of genome
- and metagenome sequences, are on Figshare (<u>https://figshare.com</u>) under project
- 563 number 133176.

564

## 565 Acknowledgments

566

567 This work was supported by Antimicrobial Resistance Cross Council Initiative supported 568 by the seven United Kingdom research councils (NE/N019881/1). ADW was funded by a 569 NERC STARS PhD scholarship (NE/M009106/1). CJGH and ACWP were funded by the 570 BBSRC Nottingham-Rothamsted Doctoral Training Partnership (BB/M008770/1). RC is 571 supported by a scholarship from the Medical Research Foundation National PhD Training 572 Programme in Antimicrobial Resistance Research (MRF-145-0004-TPG-AVISO). 573 Bioinformatic analysis was made possible via the use of MRC-CLIMB (MR/L015080/1.) 574 and CLIMB-BIG DATA (MR/T030062/1). We thank Chris Thomas, Emma Allaway and 575 David Allaway for support with grant development. We thank Nigel Armstrong and the 576 farm staff for their time, patience and support. We thank the external advisory board 577 members for support, critique and feedback of our research: Nigel Brown, Brian Dalby,

578 Gareth Hateley, Derek Armstrong, Katherine Grace, Marion Bos, Stacey Brown, Milen 579 Georgiev, Javier Dominquez, Martin Rigley, Karen Heaton, Rupert Hough, Josh Onyango, 580 Amreesh Mishra, Paul Wilson and Phil O'Neil. DJS thanks W Levine Stekel and CF Levine 581 Stekel for useful conversation and advice. We thank Emma Hooley for her support 582 throughout the entire research process. 583 584 585 586 587 588 589 590 References 591 Van Boeckel TP, Glennon EE, Chen D, Gilbert M, Robinson TP, Grenfell BT, et al. 1. 592 Reducing antimicrobial use in food animals. Science. 2017;357(6358):1350-2. 593 2. Vogt D, Overesch G, Endimiani A, Collaud A, Thomann A, Perreten V. Occurrence 594 and genetic characteristics of third-generation cephalosporin-resistant Escherichia coli in 595 Swiss retail meat. Microbial drug resistance. 2014;20(5):485-94. 596 Lammie SL, Hughes JM. Antimicrobial resistance, food safety, and one health: the 3. 597 need for convergence. Annual review of food science and technology. 2016;7:287-312. 598 4. Gundogan N, Avci E. Occurrence and antibiotic resistance of E scherichia coli, S 599 taphylococcus aureus and B acillus cereus in raw milk and dairy products in T urkey. 600 International journal of dairy technology. 2014;67(4):562-9. 601 5. Silveira-Filho VM, Luz IS, Campos APF, Silva WM, Barros MPS, Medeiros ES, et al. 602 Antibiotic resistance and molecular analysis of Staphylococcus aureus isolated from

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849 Figure Legends

#### 850 Figure 1: Antimicrobial resistance phenotypes and reads in the slurry tank (a)

851 Resistances to a panel of 16 antibiotics (Supplementary Table 2) largely do not depend

on the type of supplemented media used. (b) The number of resistances per isolates;

853 37% of cultured isolates resistant to three or more antibiotic classes. These resistances

are seen on all types of media. (c) Proportion of ARGs mapped to different antibiotic

resistance classes (% reads). The metagenomic resistance profile is largely stable over

time. There appears to be a gradual increase in the proportion of aminoglycoside and
beta lactam resistance genes, which could be seen as consistent with antibiotic use
during that period, but there is no statistical significance to the changes in proportions.
ARGs are also reasonably consistent with observed phenotype data.
Figure 2: Stability of microbial ecosystem, *E. coli* counts and water quality
measures. (a) water quality analysis from samples taken from the slurry tank over a

five month period concurrent with microbial counts. Water quality measures are

- 863 generally stable, with some fluctuations concordant with slurry use. (b) Six taxonomic 864 groups accounting for at least 1% each of microbial reads show stable abundance in 865 time. There is considerable diversity; these groups only account for 38% of reads, with 866 all reads mapped to 178 different microbial phyla. (c) Counts of *E. coli* concentrations 867 showing E. coli on TBX and MaConkey plates (all E. coli), TBX and AMP plates (E. coli 868 resistant to ampicillin) and on CTX plates (ESC-R E. coli). Overall E. coli abundance is 869 stable throughout the sampling period, as are counts of ampicillin resistant strains. CTX-870 resistant *E. coli* are only observed on five sampling days, and on all of those occasions 871 at levels too low to be reliably quantified. The data for the other days are below the limit 872 of detection of the method used and are plotted at 0 for ease of display.
- 873

862

874 Figure 3: Antibiotic susceptibility testing of *E. coli* isolates shows diverse but 875 stable range of phenotypic resistances. In each panel, the heatmap shows the 876 proportion of strains resistant to each of 16 different antibiotics on each of the sampling 877 dates. Grey bars indicate no use of those plate types on those dates. (a) plates without 878 antibiotic supplement; (b) plates supplemented with ampicillin; (c) plates supplemented 879 with cephalosporins. In all cases, the patterns of resistances are stable in time. 880 Cephalosporin supplemented plates identify more resistant strains than other plates, 881 including to other antibiotic classes, including tetracyclines and quinolones. 882

Figure 4: Model simulations of antimicrobials and antimicrobial resistance in the
slurry tank. (a) Model prediction of resistant *E. coli* populations in the slurry tank over a

885 year's period, given (b) antibiotic usage on farm in 2017. Resistance groups are not 886 mutually exclusive. The resistances are reasonably stable once the model simulation 887 reaches its steady state, with fluctuations resulting from periodic removal of slurry for 888 use as fertilizer. (b) Mass (in mg) of oxytetracycline, cefalexin and amoxicillin given 889 during 2017 together with model simulation predicting concentrations (in mg  $L^{-1}$ ) of 890 these antibiotics in the slurry tank over the same period. Observe that tetracycline is 891 present in the tank, despite intermittent use, due to its high environmental stability. This 892 explains the consistent proportion of tetracycline resistance. The two beta lactam 893 antibiotics decay more rapidly after use.

894

895 Figure 5: Storage without further waste-addition leads to a decline in select 896 **bacteria.** Relative abundances of *Escherichia* spp., *Pseudomonas* spp., *Klebsiella* spp., 897 Bacteroides spp., Prevotella spp. and Alistipes spp. in stored slurry based on 898 metagenomic short-read data. Escherichia reads from metagenomics are concordant 899 with culturing data (viable *E. coli* counts in CFU/ml over time are given in Figure S6a), 900 both showing a stepwise decline. Pseudomonas and Klebsiella also show a stepwise 901 decline. Bacteroides, Prevotella, and Alistipes show a gradual decline. Acinetobacter 902 increase over the first four weeks before declining.

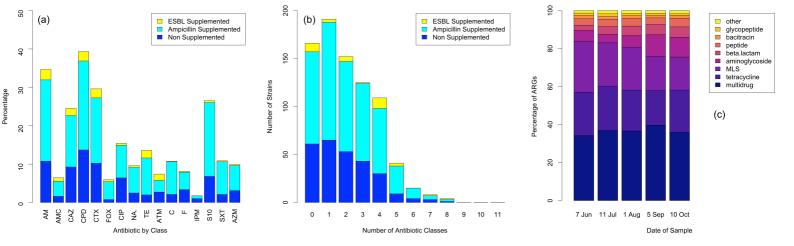
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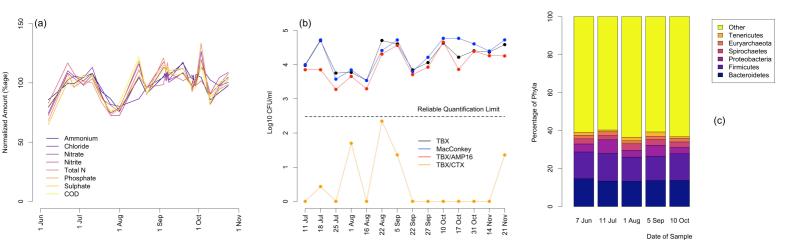
Figure 6: Impact of mini tank storage on selected ARGs based on DeepARG
analysis. Relative abundance (percentage of reads) of (a) beta lactam ARGs; (b)
tetracycline ARGs; (c) *tetY*; (d) *tet40*; (e) *tetM*; (f) MLS ARGs. The decline in beta
lactam reads is consistent with other data. Tetracycline ARGs show different patterns for
different genes. The persistence of MLS ARGs is consistent with their presence not
related to lack of MLS use on the farm.

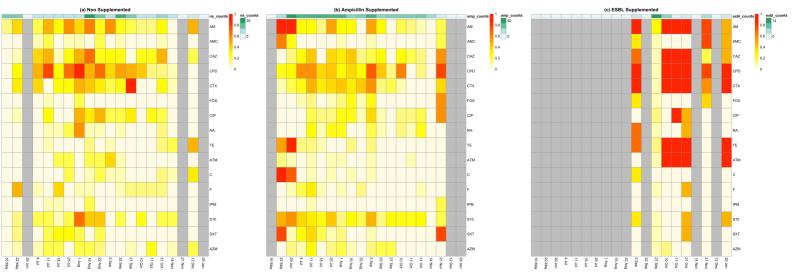
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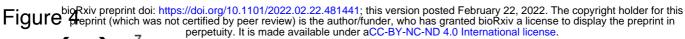
- 911 Figure 7: Model simulations of altered farm practise or antibiotic use (a) Storing
- 912 slurry without fresh inflow for 60 days is predicted to reduce resistance. Cephalexin
- 913 resistance is reduced by more than 99.99% while amoxicillin resistance is reduced by
- 914 more than 99%. (b) Model predictions of current antibiotic usage (9.7 mg/PCU)
- 915 compared to a 50% reduction (4.85mg/PCU) and 90% reduction (0.97mg/PCU) show
- 916 negligible impact on slurry tank resistance levels. (c) Model predictions of the change in
- 917 resistant *E. coli* in the tank when using a 4<sup>th</sup> generation cephalosporin instead of a 1<sup>st</sup>
- 918 generation cephalosporin on low, medium and high antibiotic usage farms showing
- 919 increased resistance to all relevant antibiotics.
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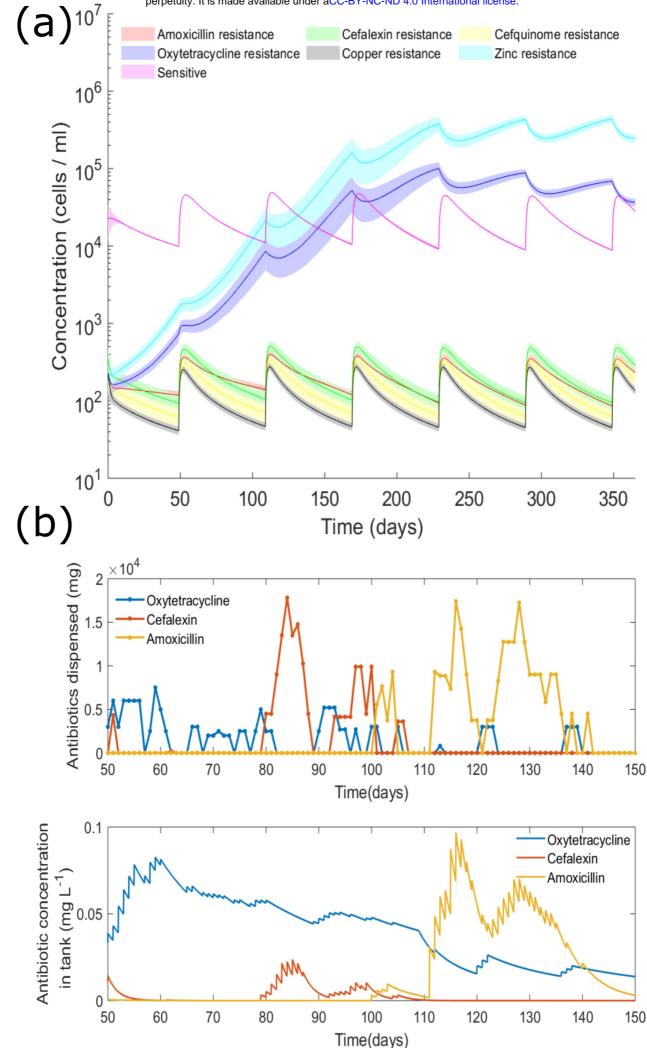
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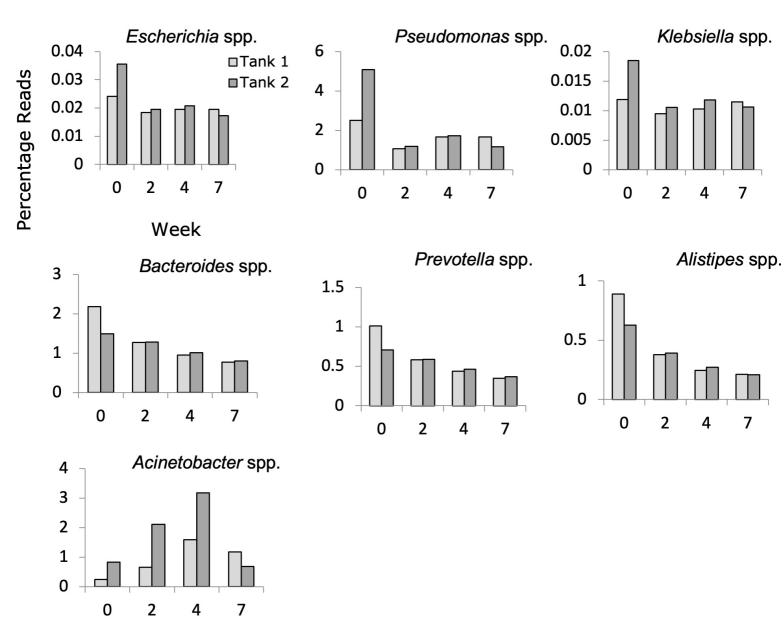








# Figure 5



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