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1 **Impact of Contrasting Poultry Exposures on Human, Poultry, and Wastewater**
2 **Antibiotic Resistomes in Bangladesh**

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26 **Abstract**

27 Interactions between humans, animals and the environment are considered critical foci for
28 addressing antimicrobial resistance (AMR). However, One Health data on AMR in low- and
29 middle-income countries are presently scarce. Using metagenomics, we investigated
30 whether and how the antibiotic resistomes of humans are influenced by exposure to
31 intensive and non-intensively reared poultry within contrasting settings of urban wet markets
32 and rural households in Bangladesh. We also consider poultry and wastewater resistomes in
33 these settings. We found that occupational poultry exposure did not significantly alter the
34 human faecal resistome. In contrast, macrolide-lincosamide-streptogramin and streptothricin
35 antibiotic resistance genes (ARGs) were enriched in poultry from urban wet markets relative
36 to rural household chickens. Wastewater had the highest ARG richness, though this was
37 only partially explained by poultry caecal and human faecal sources. Wastewater also
38 contained clinically significant carbapenem ARGs. This study therefore provides critical
39 insight into the distribution of ARGs in Bangladesh.

41 **Introduction**

42 The faecal carriage rate of extended-spectrum beta-lactamase producing Enterobacterales
43 (ESBL-E) has been increasing globally with the highest prevalence rates in South Asian
44 countries (Bezabih et al., 2021). Intestinal colonisation with antibiotic resistant organisms in
45 humans poses an elevated risk of subsequent infection with resistant organisms. Besides,
46 colonised humans and animals shed ESBL-E through faeces which are often disposed into
47 the environment due to poor sanitation infrastructure. Therefore, reduction of community
48 carriage of antimicrobial resistance (AMR) has been considered as a major step in
49 combating AMR (Maillard et al., 2020). There are many drivers for drug resistant infections in
50 low- and middle-income countries (LMICs), including unregulated sales of antibiotics; misuse
51 of antibiotics in clinical medicine and agriculture; poor sanitation and sewerage
52 infrastructure; and overall poor governance in health care.

53 Unregulated use of antibiotics in intensive farming of food-producing animals and in
54 aquaculture has become a common practice in many LMICs (Van Boeckel et al., 2015).
55 More alarmingly, antibiotics critical for human health are often used in animal and fish
56 farming, resulting in development of resistance to clinically important antibiotics among
57 bacterial pathogens of concern for human health (Myers et al., 2022). It has been suggested
58 that transmission from animals to humans of bacteria and/or mobile genetic elements
59 carrying ESBL-encoding genes may contribute to human infection with ESBL-producing
60 *Escherichia coli* (ESBL-Ec), however existing evidence suggests this occurs infrequently
61 (Madec et al., 2017; Nguyen et al., 2019). Nonetheless, in a community-based survey in
62 Bangladesh, we found that 67.5% of healthy adults and 68.0% of poultry were colonised with
63 ESBL-Ec whilst 92.5% of wastewater samples tested positive for ESBL-Ec with similar
64 prevalence rates in rural and urban settings (Rousham et al., 2021).

65 Small scale commercial poultry production in Bangladesh makes heavy use of antibiotics
66 which are added routinely to water or poultry feed (Masud et al., 2020). The close proximity
67 of humans and animals in these farms provide opportunities for bidirectional transmission of

68 antibiotic resistant bacteria (ARB) and antibiotic resistant genes (ARGs) between hosts.
69 Moreover, the lack of sanitation infrastructure, waste management and waste treatment in
70 both rural and urban areas has led to widespread environmental contamination by faecal
71 bacteria, resistance genes and antibiotic residues. Small-scale commercial poultry farmers
72 and the sellers of live poultry in urban markets in Bangladesh also face direct exposure to
73 animal tissues, waste products and associated ARB and ARGs because they use little or no
74 protective clothing, gloves or masks.

75 The contribution of different sources to AMR in the environment is considered an important
76 gap in current understanding (Larsson et al., 2018). Although ARB have been shown to be
77 prevalent in humans, animals and environmental samples in Bangladesh, the diversity and
78 abundance of ARGs and the extent of sharing between different hosts and environmental
79 compartments is less studied.

80 Only a few studies in LMICs have assessed all three domains (human, animal and
81 environment) of the One Health paradigm for AMR surveillance (Rousham et al., 2018b).
82 Importantly, there is a lack of data on sharing of microbiome and antibiotic resistomes
83 between humans, domestic/farmed animals and environments within similar ecological units
84 in LMIC settings, where humans and livestock often live in close proximity, and sanitation is
85 often inadequate. Metagenomics-based surveillance makes it possible to compare
86 resistomes and bacterial population structures within and between different ecological
87 settings. Organisms that are significant contributors of ARGs within a population, including
88 culturable, non-culturable and under-studied organisms, can be detected by metagenomic
89 analysis. Recently sewage and wastewater surveillance for AMR using metagenomics has
90 gained traction due to its advantages over traditional population-based surveillance, which is
91 resource intensive for many LMICs (Pruden et al., 2021; Prieto Riquelme et al., 2022).
92 Paradoxically, many LMIC settings that are considered as hotspots for AMR have the
93 sparsest data on metagenomic-based AMR surveillance.

94 In this study, we aimed to understand the dynamics of antimicrobial resistance transmission
95 in Bangladesh. Specifically, to determine whether the human gut resistome can be explained
96 by exposure to poultry and wastewater in both poultry intensive and non-intensive settings.
97 Additionally, we investigated the impact of intensive antibiotic exposure among poultry on
98 the composition of the poultry gut resistome and its contribution to downstream
99 environmental contamination through direct disposal of wastewater. We achieved this by
100 leveraging systematically collected samples from humans, poultry and the surrounding
101 wastewaters in urban wet markets and rural villages in Bangladesh as part of a purposefully-
102 designed One health AMR surveillance study (Rousham et al., 2018a) and analysed them
103 for whole metagenomic profiling.

104 **Methods**

105 *Sampling strategy*

106 The samples discussed in the present work were collected as part of a One Health
107 surveillance study on ESBL-Ec in Bangladesh (Rousham et al 2018; Rousham et al 2021).
108 Human faecal, poultry ceca and wastewater samples were collected between February and
109 October 2017, from urban wet markets in Dhaka city and from households in a rural area in
110 the Tangail district. The current work focusses on a subset of 40 samples which were
111 subsequently processed for metagenomic sequencing. Specifically, these include 20 human
112 faecal samples where we use the terms high and low exposure to refer to the extent of
113 exposure to poultry (rural householders with backyard poultry $n=2$; rural households without
114 backyard poultry $n=5$; poultry slaughterers in urban wet markets $n=6$, people with other
115 professions in urban wet markets $n=7$), 10 poultry caecal samples (backyard poultry from
116 rural households $n=4$; broilers from urban wet markets $n=6$) and 10 wastewater samples
117 (rural households $n=4$; urban wet markets $n=6$). Wastewater samples were collected from
118 the outlet of the main wastewater drain at each urban market and rural household.
119 Wastewater in urban market outlets typically contain waste from all parts of the market
120 including large and small animals, fish, fresh produce etc. More detailed metadata are

121 available in Supplementary Table 1. We aimed to analyse shared and distinct resistomes
122 according to sample origin (human, poultry or wastewater), occupational exposure to poultry
123 and setting (urban wet markets versus rural households).

124 *Sample collection*

125 Human faecal samples were provided by study participants using a sterile stool sample
126 container supplied by field staff. All faecal samples were stored on ice within 2 hours of
127 collection. For poultry caeca samples, chickens were slaughtered, and the skin removed on-
128 site by the owner following their usual procedures. The carcass was placed in a sterile bag,
129 sealed immediately and placed in a cool box on ice for transportation. Wastewater samples
130 were collected by taking approximately 150 mL of wastewater from three locations along the
131 runoff drain adjacent to the selected household or market, by dipping a sterile container into
132 the drain. Wastewater samples were then pooled by location in a sterile 500-mL plastic
133 bottle (Nalgene, New York USA) and placed on ice for transportation.

134 All samples were transported to the laboratory within 5 hours of collection maintaining the
135 cold chain, refrigerated on arrival, and processed within 18 hours of collection. In the
136 laboratory, caeca samples were taken from the chicken carcass aseptically by cutting the
137 keel bone, identifying, and excising the cecum with sterilized scissors and extracting the
138 caecum contents.

139 *Ethical considerations*

140 Written and verbal information about the study was provided, and participating volunteers
141 gave written informed consent. Ethical clearance was obtained from icddr,b, Bangladesh
142 (PR-16071) and Loughborough University, UK (R17-P037). Local authorities were informed
143 prior to wastewater sampling in markets. Broiler poultry and backyard chickens were
144 purchased on a commercial basis and slaughtered by the owner/vendor who had consented
145 to participate. Owners followed their normal procedure for slaughtering animals as for
146 domestic consumption or commercial sale.

147 *Metagenomic Sequencing and Sequence Cleaning*

148 DNA was extracted from wastewater samples using the MO Bio Power Water DNA isolation
149 kit (MO BIO Laboratories Inc, Carlsbad, CA USA) the QIAamp DNA Stool Mini Kit (Qiagen,
150 UK) was used for human faecal and poultry ceca samples. Short read metagenomic
151 sequencing (Illumina NovaSeq 6000, 150bp paired end libraries) of extracted DNA was
152 carried out by Novogene (Novogene Co. Ltd, Cambridge, UK). On average, sequencing
153 yielded approximately 11GB of data per sample.

154 Removal of sequencing adapters and quality trimming was carried out with *Trimmomatic*
155 v0.38 (Bolger et al., 2014) (settings: 2:30:10, leading:3, trailing:3, slidingwindow:4:15,
156 minlen:36). Host and other non-bacterial reads were removed by mapping with *bowtie2*
157 v2.3.5 (Langmead and Salzberg, 2012) using default settings. The read removal strategy
158 was based on the biological context of samples. For human stool samples, the human
159 genome (RefSeq GCF_000001405.40_GRCh38.p14) was used for reference mapping and
160 read exclusion. For poultry ceca, the broiler chicken (*Gallus gallus*) genome was used for
161 reference mapping (RefSeq GCF_016699485.2_bGalGal.mat.broiler.GRCg7b). Finally, for
162 wastewater samples, human, broiler chicken and cattle (*Bos taurus*; RefSeq
163 GCF_002263795.2_ARS-UCD1.3) genomes were used for reference mapping.

164 Two samples (DL_164_WW2 and DL_087_WW2) were shown to be heavily contaminated
165 (>80% reads mapped concordantly) with broiler chicken genetic material. Both samples
166 originated from urban wet markets; the high proportion of reads mapping to broiler chicken is
167 therefore likely to be a consequence of poultry slaughtering and waste disposal practice.

168 *Metagenome Assembly*

169 Following quality control and non-target sequence removal, samples were assembled
170 individually using *Megahit* v1.2.9 with 21, 29, 39, 59, 79, 99, 119, and 141 *k*-mer intervals (Li
171 et al., 2015b).

172 *Annotation of Antimicrobial Resistance Genes in Unassembled Data and Contigs*

173 ARGs were annotated with a locally installed copy of *ARG-OAP* v2.0 (Yin et al., 2018) using
174 an ARG sequence identity cut-off of 80% and minimum query alignment length >25 amino
175 acids, expect-value 1×10^{-7} (Feng et al., 2018; Murray et al., 2019; Qian et al., 2021).

176 Normalisation of ARGs by estimated genome number was carried out by *ARG-OAP* which
177 uses *diamond* v2.0.15. (Buchfink et al., 2021) to identify a suite of 30 universal single copy
178 genes.

179 Contigs were screened for ARGs based on protein homology using the RGI online platform
180 (resistance gene identifier, v6.0.0), combined with the CARD database (v3.2.5) (Alcock et
181 al., 2020). Only matches exceeding 90% identity and coverage of reference protein
182 sequences were considered in analyses.

183 *Taxonomic Classification of Unassembled Data and Resistance Gene-Bearing Contigs*

184 After quality checking, taxonomic classification of short reads was carried out with *Kaiju*
185 v1.7.1 (Menzel et al., 2016) in combination with the pre-built *nr_euk* protein database
186 (downloaded from *Kaiju* webserver March, 2019). The database contains non-redundant
187 protein sequences for Archaea, Bacteria, Viruses, fungi and microbial eukaryotes from the
188 NCBI-BLAST database (Sayers et al., 2019). *Kaiju* was run in 'greedy' mode, allowing three
189 mismatches.

190 For ARG-bearing contigs of particular interest, NCBI megaBLAST (Morgulis et al., 2008) was
191 performed to assign putative taxonomy.

192 *Data Exploration and Statistical Analysis*

193 To evaluate whether the predicted resistome composition varied between sample origin
194 (poultry caecal, human faecal and wastewater) and setting (rural households and urban wet
195 markets), Principal Coordinates Analysis (PCoA) was first used to generate unconstrained
196 ordinations of sample composition (Hellinger distances of bacterial cell normalised ARG
197 category and subtype abundance). As PCoA indicated clustering, permutational multivariate
198 analysis of variance (PERMANOVA) was used to establish if significant multivariate

199 differences were observed between groups, following testing for heterogeneity of
200 multivariate dispersion using the PERMDISP test. We performed discriminant analysis using
201 canonical analysis of principal coordinates (CAP) in Hellinger space. To avoid model over-
202 parameterisation, we identified the optimal PCoA axes to employ in CAP by maximising a
203 leave-one-out allocation success to treatments. Having established any clustering within the
204 multivariate ordination, we determined the likely ARG categories associated with sample
205 clustering using Pearson correlation coefficients to determine linear relationships between
206 ARGs and clusters. PCoA, PERMDISP, PERMANOVA and CAP were all conducted using
207 the *PERMANOVA+* add on to *PRIMER* version 7.0.20 (Clarke and Gorley, 2006; Anderson
208 et al., 2008). For all tests, probabilities were based upon 99,999 permutations (denoted
209 p_{perm}). In cases where the number of observations was insufficient to allow at least 999
210 permutations for post hoc pairwise tests, Monte Carlo probabilities (denoted p_{MC}) were
211 calculated based upon an asymptotic permutation distribution.

212 Based on PERMANOVA, univariate differential abundance of ARGs in rural and urban
213 settings were only assessed in poultry caecal samples. Only ARG categories associated
214 with poultry caeca having $r > 0.2$ were analysed, see vectors associated with CAP analysis
215 (Fig. 1). To account for non-normal data distribution and heterogeneity of variance between
216 sample data, Welch's unequal variances t -tests were combined with 99,999 Monte Carlo
217 permutations to determine probabilities associated with t . The Benjamini-Hochberg
218 procedure for false discovery rate adjustment was applied to the resulting probabilities
219 (denoted p_{adj}).

220 ARG subtype diversity was assessed using Hill numbers implemented by the *iNEXT* v3.0.0
221 (Hsieh et al., 2016) package in R with 40 knots and confidence intervals bootstrapped 399
222 times.

223 Other exploratory visualisations included boxplots of genome-normalised abundances of the
224 five most abundant ARG subtypes by sample origin and ARG categories associated with
225 sample origin according to CAP ($r > 0.2$). Abundant phyla and genera were summarised by

226 sample origin in compositional plots produced with the R package *microbiome v1.18.0* (Leo
227 Lahti, 2017) (Supplementary Figure 1).

228 Further investigation into associations between specific ARG subtypes, taxa and sample
229 origins were guided by contigs of interest and initial data exploration. In doing so we sought
230 to avoid data dredging. We focussed on the following objectives: i) identifying whether
231 specific taxa could be shown to drive dominant features of the resistome in the three sample
232 origins (human, poultry, wastewater), ii) identifying taxa associated with multi-drug
233 resistance determinants (three or more different antibiotic categories) iii) determining the
234 distribution of the WHO priority one antibiotic resistant bacteria of critical concern (WHO,
235 2017). Where appropriate, associations between the Centred Log Ratio (CLR) of taxa and
236 ARG count data were tested using major axis regressions in R using *lmodel2 v1.7-3*
237 (Legendre, 2018). Only contigs >1 kbp were used for assigning putative taxonomy.

238 Antibiotic resistome source contribution analysis was carried out with *FEAST* algorithm R
239 package (Shenhav et al., 2019; Chen et al., 2023) where poultry, human and wastewater
240 samples were collected from the same urban wet market site ($n = 3$). *FEAST* was supplied
241 count data for ARG subtype and was run using default settings (EM iterations= 1000,
242 coverage = minimal sequencing depth in sink and sources).

243 **Results**

244 **Sample origin has the greatest impact on resistome composition; setting has a** 245 **secondary influence on poultry and wastewater ARG carriage.**

246 Genes associated with tetracycline resistance were the most abundant ARG category on
247 average, regardless of sample origin (mean abundance, 1.2 copies per bacterial genome
248 [cpbg], ± 0.12 standard error of the mean). However, based on abundance data, poultry
249 ceca samples were the most enriched with tetracycline resistance genes (1.79 ± 0.18 cpbg,)
250 compared to both wastewater (1.07 ± 0.32 cpbg) and human faecal samples (1.05 ± 0.05
251 cpbg). The next most abundant ARG categories overall, were Macrolide-Lincosamide-

252 Streptogramin (MLS) resistance genes (0.63 ± 0.09 cpbg) and beta-lactam resistance genes
253 (0.55 ± 0.06 cpbg); these were most associated with poultry ceca (1.10 ± 0.22 cpbg) and
254 human faecal samples (0.80 ± 0.06 cpbg), respectively. These trends were evident in the
255 CAP linear ARG vector overlay (Figure 1.) and ARG category normalised abundance
256 boxplots (Figure 2b).

257 At the subgroup level, no single ARG was dominant across human, poultry or wastewater
258 samples. On average, the tetracycline ARG *tet(Q)* was the most prevalent subtype in poultry
259 samples, *tet(Q)* and the beta-lactam resistance subtype *cfxA6* in human faecal samples
260 (0.51 ± 0.05 and 0.61 ± 0.06 cpbg respectively), whereas wastewater samples were marked
261 by greater between-sample variability and therefore multiple similarly abundant subtypes
262 were present (Figure 2a).

263 Around 15% of core ARG subtypes (here defined as subtypes detected in at ≥ 2 samples in
264 one or more sample origin; total $n = 1047$) were shared by all sample origins (Supplementary
265 Table 2). However, analyses show that the overall resistome compositions associated with
266 each sample origin were distinct. This is evident in the separation of poultry ceca, human
267 faecal and wastewater samples by unconstrained PCoA and CAP of ARG subtype and
268 category data (Figure 1, see Supplementary Figure 2 and 3 for PCoA). ARG subtype
269 diversity was assessed using Hill numbers. There was a significant difference in median
270 ARG subtype richness (0D) between sample origin (human, poultry, wastewater) (Kruskal-
271 Wallis test; $\chi^2 = 23.3$, two-tailed $df = 3$, $p = 3 \times 10^{-5}$), but not setting (urban wet market versus
272 rural households) Kruskal-Wallis test; $\chi^2 = 1.5$, $df = 1$, $p = 0.214$. (Supplementary Figure 4).
273 Post hoc pair-wise comparisons of median 0D indicated significant differences between
274 wastewater (median ${}^0D = 526$) and all other sample origins (smallest difference; wastewater
275 and poultry caecum, median ${}^0D = 223$, Mann-Whitney $Z_U = 3.6$, $p_{adj.} = 0.00074$). No
276 significant differences in richness estimates were observed between poultry caecum or
277 human faecal samples (median 0D of human faeces from high or low poultry exposure were
278 137 and 171, respectively). However, a significant interaction was identified between setting

279 and origin; Kruskal-Wallis test; $\chi^2 = 27.5$, $df = 5$, $p = 4.573e^{-5}$. Despite this, group-wise
280 testing indicated no significant differences after correction for multiple testing (Kruskal-Wallis
281 tests: poultry caecal; $\chi^2 = 6.55$, $df = 1$, $p_{adj.} = 0.06312$; human faecal; $\chi^2 = 0.01$, $df = 1$, $p_{adj.} =$
282 1.00 ; wastewater; $\chi^2 = 0.73$, $df = 1$, $p_{adj.} = 1.00$).

283

284 For the multivariate test of ARG assemblages, no significant heterogeneity of multivariate
285 dispersion was observed for either sample origin or setting for ARG type. For ARG subtype
286 however, there was significant heterogeneity of multivariate dispersion associated with
287 sample origin (PERMDISP, pseudo- $F = 5.9$, $p_{perm} = 0.0058$). For both ARG type and
288 subtype, dispersion was greater for wastewater samples than either human faecal or poultry
289 caecal samples. Testing the effect of sample setting (rural households, urban wet markets)
290 and sample origin (human faecal, poultry caecal, wastewater) indicated both sample origin
291 (PERMANOVA, pseudo- $F = 24.8$, $p_{perm} = 1 \times 10^{-5}$) and setting (pseudo- $F = 3.7$, $p_{perm} = 0.0087$)
292 exerted a significant influence upon ARG subtype assemblages. Significant interaction
293 between the two factors was also identified (pseudo- $F = 3.0$, $p_{perm} = 0.001$). Post hoc
294 pairwise comparisons indicated that while there was no significant difference between ARG
295 subtype assemblages in faecal samples collected from human subjects with either high or
296 low exposure to poultry, assemblages in poultry caeca, human faeces and wastewater were
297 all significantly different; however, we cannot discount the fact that this may reflect
298 differences in dispersion between the different environments. This pattern was the same for
299 both rural households and urban wet markets. In addition, differences in ARG subtype
300 assemblages were also observed between wastewater ($t = 1.9$, Benjamini-Hochberg
301 adjusted $p_{MC} = 0.014$) and poultry caeca ($t = 3.3$, Benjamini-Hochberg adjusted $p_{MC} =$
302 0.0007) from rural households and urban wet markets. These trends were consistent with
303 the effect of factors upon ARG type.

304 After log₁₀ transformation, total normalised ARG abundance varied significantly between
305 settings (Welch $t = -4.17$, $df = 29.623$, $p = 0.00024$), with urban wet market samples
306 exhibiting higher mean total ARG abundance than rural households. Due to small sample

307 size, unequal variance, and non-normal distributions which could not be resolved by
308 transformation, we did not carry out further statistical tests on interaction effects between
309 sample origin and setting. However, we observed trends which suggest the difference
310 between urban wet market and rural households is primarily driven by wastewater samples
311 (Supplementary Figure 5).

312 *FEAST* source attribution where wastewater, human faecal and poultry caecal samples were
313 collected from the same urban wet markets showed that the majority of the wastewater
314 resistome could not be explained by either potential source (mean 75.87% \pm 4.50 unknown
315 source). In urban wet markets, the poultry resistome made a greater contribution to
316 corresponding wastewater streams (21.39% \pm 3.08) than human faecal sources (2.73% \pm
317 0.01%).

318 **MLS and streptothricin ARGs are enriched in broiler poultry ceca**

319 The multivariate differences between rural backyard chickens and urban wet market broiler
320 poultry resistomes indicated by PERMANOVA were further investigated. Of the three ARG
321 categories shown to associate with poultry resistomes according to CAP vector overlay
322 (Figure 1), MLS and streptothricin resistance genes were significantly enriched in urban
323 broilers (Welch $t = 3.8$, $p_{adj.} = 0.0145$; Welch $t = 6.3$, $p_{adj.} = 0.0245$, respectively; Figure 3).
324 However, tetracycline resistance genes were prevalent at similar levels in urban and rural
325 bird ceca (Welch $t = 0.51$, $p_{adj.} = 0.567$). Given the small sample size and numerous ARG
326 subtypes, statistical tests are not reported; nonetheless, heatmaps of MLS and streptothricin
327 ARG subtypes are provided for poultry samples in Supplementary Figure 6. Several ARG
328 categories were shown to be indicative of wastewater resistomes, however differences
329 between urban and rural samples were not tested due to low statistical power.

330 **Urban and rural wastewaters contain carbapenem resistance genes**

331 A total of 1662 contigs contained at least one ARG. Of these, 1553 were ≥ 1 kbp in length
332 and were selected for taxonomic profiling (poultry, $n = 363$; human faeces, $n = 640$;
333 wastewater, $n = 550$).

334 According to the WHO Global Priority Pathogens List, priority one 'critical concern' antibiotic
335 resistant pathogens include ESBL-E, carbapenem resistant *Enterobacteriaceae* (CRE),
336 *Acinetobacter baumannii* and *Pseudomonas aeruginosa*. While 21 contigs >1kbp contained
337 ESBL/carbapenem resistance genes, after cross-validation of taxonomy, only three contigs
338 unambiguously fulfilled the 'critical concern' criteria; all originated from wastewater samples
339 (Supplementary Table 3).

340 Human faecal (high and low exposure), poultry, and wastewater contig libraries contained
341 *Enterobacteriaceae* carrying TEM-family genes, however all were classified as broad-
342 spectrum beta lactamases (BSBLs) TEM-1. Similarly, SHV genes in wastewater (SHV-27,
343 SHV-110) and human faecal (SHV-11, SHV-27) contigs were determined to be non-ESBL
344 variants. PER and RSA group ESBL genes were detected, however, they could not be
345 confidently associated with *Enterobacteriaceae*.

346 OXA-group genes were the most frequently identified beta-lactamases on contigs across the
347 entire dataset ($n = 50$, $n = 49 >1\text{kb}$) and were found in all sample origins. However, OXA
348 ESBL/carbapenemase genes could not be definitively associated with *Enterobacteriaceae*.
349 For example, one rural wastewater contig carried an OXA-372 family carbapenemase gene
350 (Bonnin et al., 2021) (OXA-641-like); OXA-641 is known to be found in *Morganella* spp
351 (Supplementary Figure 7a). Additionally, one chromosomal *Acinetobacter baumannii* rural
352 wastewater contig encoded a variant (OXA-65) belonging to the OXA-51 family (Evans et al.,
353 2008) (Supplementary Figure 7b). Another *Acinetobacter* spp. contig derived from urban
354 wastewater contained OXA-58, another a potential carbapenemase (Poirel et al., 2005)
355 (Supplementary Figure 7c).

356 VEB group ESBLs were also identified; these were only recovered from wastewater contigs
357 and could not be categorically associated with *Enterobacteriaceae*. However, the longest
358 VEB-bearing contig (7.7kbp) contained a class 1 integron cassette with the potential to
359 confer multidrug resistance (Supplementary Figure 7d) across four different antibiotic
360 categories (beta-lactam, phenicol, aminoglycoside and fluoroquinolone). NCBI-BLAST

361 indicated this contig shared high nucleotide homology with enterobacterial plasmids
362 (*Klebsiella quasipneumoniae*, 99.69% identity, 100% query coverage [CP058135.1], and
363 *Escherichia coli*, 99.98%, 100% query coverage [LC745731.1]) and non-enterobacterial
364 chromosomes (*Aeromonas veronii*; 100% identity, 94% query coverage [CP054855.1]).
365 Finally, three wastewater contigs contained genes from the GES carbapenemase family.
366 Urban and rural wastewater contigs (1.3-1.5 kbp) containing *bla*GES-2 shared high
367 homology with *Pseudomonas aeruginosa* chromosomes and enterobacterial plasmids
368 (Supplementary figure 7e). Another urban wastewater contig (1.8 kbp) contained *bla*GES-5
369 with nucleotide homology among Gammaproteobacteria, including chromosomal sequences
370 of *Pseudomonas aeruginosa* (99.90% identity, 70% query coverage [KY860573.1]) and
371 *Klebsiella pneumoniae*-associated plasmids (100% identity, 70% query coverage
372 [MN436715.1]) (Supplementary figure 7f, Supplementary Table 3).

373 **Select ARGs are associated with specific genera and sample origin, while most are**
374 **widely dispersed.**

375 To investigate whether select ARGs are linked to specific bacterial hosts and sample origin,
376 contigs were used to direct correlation analyses between short-read abundance data of
377 ARGs and taxa. Since the prevalence of beta-lactam resistance genes were shown to
378 distinguish human faecal samples from those collected from poultry caeca and wastewater
379 (see Figure 1, 2b), dominant beta-lactam ARGs were identified. The principal beta-lactam
380 ARG in human samples was *cfxA6*. Contig data suggested CFXA family genes were
381 typically associated with *Bacteroides* and *Prevotella* spp. We found a significant positive
382 correlation between the CLR of *Prevotella* spp. and *cfxA6* counts across the entire dataset
383 ($R = 0.95$, $t = 18.954$, $p = 5.799111e^{-21}$, Supplementary Figure 8) and positive trends held
384 within each sample origin. However, correlations with *Bacteroides* spp. or phylum
385 Bacteroidetes were inconsistent within samples of different origin. Although contig analyses
386 show *cfxA* genes are present in both *Bacteroides* spp. and *Prevotella* spp., these findings
387 indicate *Prevotella* spp. were the most consistent carrier of *cfxA6* in these data.

388 The abundance of several ARG categories was shown to distinguish wastewater samples
389 from other environments, including multidrug, sulfonamide, trimethoprim and phenicol ARGs.
390 Contig data suggests chloramphenicol (*cmIA5*) and trimethoprim (*dfrA14*) resistance genes
391 may co-occur on either chromosomes or plasmids (Supplementary figure 7d). The co-
392 occurrence of multidrug (*qacEΔ1*) and sulfonamide (*sul1*) resistance genes may similarly
393 occur on both plasmids and chromosomes, and this configuration can be found in all sample
394 origins, not only wastewater. These two genes form the 3'-Conserved Segment of class 1
395 integrons, which are distributed across a broad taxonomic range. Considering this,
396 correlations between wastewater marker taxa and ARGs were not carried out.

397 Lastly, dominant tetracycline ARG subtypes in poultry did not consistently occur on contigs
398 belonging to specific taxa below the order level. Abundance correlations between taxa and
399 ARGs were therefore not tested.

400 **Discussion**

401 This study aimed to explore the sharing of antimicrobial resistomes between humans, poultry
402 and wastewater, with and without exposure to intensive poultry production. We also
403 assessed the influence of antibiotics fed to poultry on the composition of the poultry gut
404 resistome and its contribution to environmental contamination via wastewater outlets.

405 We found that antimicrobial resistomes were largely characterised by sample origin (human
406 faeces, poultry ceca or wastewater), although those of poultry ceca and wastewater were
407 additionally influenced by setting: namely, wet market broiler versus backyard poultry, and
408 urban wet market versus rural wastewater.

409 The primary separation of samples by source is not surprising. Existing resistome surveys
410 illustrate that different sample origins harbour distinct ARG assemblages (Li et al., 2015a;
411 Gatica et al., 2019). In certain cases, these differences may reflect the microbial taxa which
412 are adapted to specific environments and their intrinsic or commonly acquired resistance
413 genes. For example, the *cfxA* beta-lactamase gene family has been identified as a dominant

414 member of mammalian gut and faecal resistomes, including healthy humans (Hu et al.,
415 2013) cattle (Zhou et al., 2016; Baker et al., 2022) and pigs (Li et al., 2015a; Lim et al.,
416 2020). Gatica et al. (2019) also showed robust association of *cfxA* genes with bovine and
417 human faecal samples, while demonstrating their comparative scarcity in environmental
418 samples.

419 The *cfxA* genes are well-documented within the phylum Bacteroidetes, of which genera such
420 as *Prevotella* spp. and *Bacteriodes* spp. are abundant within mammalian anaerobic niches:
421 the gastrointestinal tract and oral cavity. Furthermore, Suriyaphol et al. (2021) found a
422 positive correlation between *Prevotella* spp. and *cfxA6* in pig gut microbiomes. We found a
423 positive correlation between *cfxA6* and *Prevotella* spp. across all samples in the current
424 work.

425 However, in many cases it was not possible to define such clear associations between highly
426 abundant ARGs and bacterial genera in wastewater and poultry samples. For example,
427 contigs harbouring tetracycline genes such as *tetQ* and *tetW* could not be linked to specific
428 taxa below the rank of order (Bacteroidales) or phylum (Firmicutes) in any sample origin.
429 This likely corresponds to the decoupling of these genes from strict phylogenetic constraints
430 by virtue of horizontal gene transfer. Indeed, *tetW* is known to have an extensive host range
431 encompassing both Gram-positive and -negative bacteria, and can be integrated on conjugal
432 transposons (Roberts, 2005). The mobility of *tetQ* is less well documented, although it has
433 been found on plasmids in *Bifidobacterium* strains (Ma et al., 2015).

434 Wastewater samples had the highest ARG subtype richness (0D) compared to human and
435 poultry samples consistent with having the greatest multivariate dispersion. Furthermore, Hill
436 number extrapolations indicate the true richness of some wastewater samples may be
437 underestimated due to under-sequencing (Supplementary Figure 8). These findings are
438 consistent with wastewater receiving varied inflow material, including washing detergents,
439 cooking residues, human and animal waste, and residues from animal slaughter. The
440 wastewater resistome is therefore not simply a combination of human and poultry ARGs.

441 *FEAST* source attribution also supports this assertion, since less than half of the wastewater
442 resistome in urban wet markets could be explained by corresponding human and poultry
443 samples.

444 The complexity and variability of the wastewater resistome is also likely to explain why
445 several ARG categories discriminate wastewater from human and poultry samples (shown in
446 Fig. 1). Another contributing factor could be co-localisation of genes conferring resistance to
447 different categories of antibiotic compounds. Several contigs assembled from wastewater
448 samples support this notion: aminoglycoside resistance genes were frequently identified
449 alongside beta-lactam, phenicol, trimethoprim and fluoroquinolone resistance genes. Our
450 results show examples of these contigs share homology with both the chromosomes of
451 *Aeromonas* spp. and plasmids associated with *Enterobacteriaceae* (Supplementary Figure
452 7d).

453 The overall predominance of tetracycline resistance genes in the present study parallels a
454 previous study showing these genes are prevalent in humans, goats and chicken faeces in
455 Bangladesh (Swarthout et al., 2022). Tetracycline resistance genes have also been shown
456 to dominate the resistomes of human, pig and poultry faeces in Chinese wet markets (Wang
457 et al., 2019). Metagenomic studies on wastewater outlets in Bangladesh are limited, with
458 most wastewater surveillance studies relying on cultivating select target organisms
459 (Asaduzzaman et al., 2022). To our knowledge, existing metagenomic studies on water
460 resistomes in Bangladesh only include surface water (McInnes et al., 2021). Since the
461 wastewater outlets surveyed in the present work discharge directly into surface water, we
462 discuss whether similar resistance genes were recovered in these compartments. McInnes
463 et al. (2021) found differences in the resistome of surface water and sediment obtained from
464 rural (Mymensingh, Shariatpur) and urban (Dhaka) sites in Bangladesh. Specifically, the
465 authors showed that urban surface waters were most enriched with macrolide, sulfonamide,
466 aminoglycoside and multidrug efflux resistance genes, relative to rural settings. In the
467 current work, sulfonamide, aminoglycoside and multidrug resistance genes were also among

468 those prevalent in wastewater samples from urban wet markets (Figure 1). McInnes et al.
469 (2021) suggested that human gut bacteria drive antibiotic resistance genes in surface water.
470 Given surface water bodies in urban areas receive wastewater run-off from wet markets and
471 mainstream sewage, both human and animal waste should be considered potential sources
472 of these genes in the environment. Lastly, McInnes et al. (2021) found tetracycline
473 resistance genes made a comparatively minor contribution to surface water bodies in urban
474 areas, indicating that while they dominate wastewater in our study, bacteria carrying these
475 genes (though apparently diverse) may be less capable of competing in the wider aquatic
476 environment.

477 Our findings suggest there was no significant difference between the abundance of
478 tetracycline resistance genes in broiler poultry and backyard chicken ceca (Figure 3). This
479 was unexpected given that tetracycline is frequently used in commercial poultry farming
480 (Hasan et al., 2011; Rousham et al., 2021). Our household survey indicated that none of the
481 rural poultry had received tetracycline antibiotics in the 4 weeks prior to sampling (Rousham
482 et al., 2021). Though vendors reported that they do not give antibiotics to broiler poultry at
483 the point of sale, the small-scale commercial farms which supply bird markets regularly use
484 antibiotics throughout the production cycle. In a survey of small-scale commercial broiler
485 farmers, 65% reported using tetracyclines (Rousham et al., 2021) with no withdrawal period
486 before supplying birds to retail markets (Masud et al., 2020). It has been shown that
487 tetracyclines can have long half-lives in the environment compared to other commonly used
488 antibiotics (Baker et al., 2022), although many factors influence their persistence (Cycoń et
489 al., 2019). The high prevalence of tetracycline resistance in both rural backyard chicken and
490 broiler poultry may therefore indicate widespread contamination of the terrestrial
491 environment with tetracycline residues and the proliferation of resistant bacteria.

492 A salient finding of the present work is that the resistomes of rural backyard chicken ceca
493 are distinct from those of commercially reared poultry sold in urban wet markets. This
494 appears to be driven in part by MLS (macrolide-lincosamide-streptogramin) and

495 streptothricin (nucleoside) resistance genes, which were significantly enriched in wet market
496 broilers relative to rural backyard chickens. It is likely that the higher stocking density, range,
497 and frequency of antibiotic exposure in commercial broiler farms contribute to these
498 differences. Although the wet market environment itself may play a role in shaping the
499 resistome of broiler poultry, it is important to note that the surveyed birds spent less than 24
500 hours in the urban wet market prior to slaughter. Consequently, we contend that the rearing
501 environment exerts a greater influence on the resistome of these animals. According to
502 ordinations, the caecal resistome of rural backyard chickens were less distant from human
503 resistomes than urban wet market broiler resistome compositions (Supplementary Figure 2
504 and 3). This finding likely corresponds with microbiome sharing facilitated through routine
505 direct contact with owners and indirect exchange through mutual exposure within household
506 settings.

507 MLS-bearing contigs indicated multiple MLS resistance gene families were responsible for
508 category level enrichment in broiler poultry, with putative hosts among Firmicutes,
509 Bacteroidetes and Proteobacteria. MLS genes were occasionally co-localised with
510 tetracycline, aminoglycoside and beta-lactam resistance genes in various configurations.
511 These included association with transposable elements, highlighting their mobility. Macrolide
512 antibiotics are known to be widely marketed in Bangladesh (Orubu et al., 2021) for both
513 human and animal use; they have also been used by farms supplying urban wet markets
514 surveyed in this study. Streptothricin, on the other hand, has only been used outside human
515 clinical practice for prophylactic growth promotion in animals (Webb et al., 2017). We found
516 no evidence of its use within poultry operations in Bangladesh. Broiler poultry contigs
517 containing streptothricin resistance *sat*-family genes were commonly co-localised with
518 aminoglycoside resistance genes and distributed across phyla (Firmicutes and
519 Proteobacteria).

520 To date, few studies have applied metagenomic techniques to examine caecal resistomes in
521 poultry samples, in Bangladesh or elsewhere. One study used a qPCR array to compare the

522 faecal resistome of small-scale broiler chickens with backyard poultry in Ecuador (Guo et al.,
523 2018). The authors found ARG richness in production chickens was significantly higher than
524 that of household chickens; our metagenomic data indicated only a trend in Bangladesh
525 (Kruskal-Wallis test; $\chi^2 = 6.5$, $df = 1$, $p_{adj.} = 0.06312$).

526 A recent publication by Swarthout et al. (2022) used long-read sequencing to compare
527 faecal resistomes of humans, goats and chickens in urban and rural households in
528 Bangladesh. There are several experimental and methodological differences between the
529 aforementioned study and our current work. Firstly, only household backyard poultry (rural
530 and urban) were sampled, whereas the current study sampled commercially reared broilers
531 sold in urban wet markets and backyard poultry in rural households. Secondly, Swarthout et
532 al. (2022) separated urban and rural samples and then pooled DNA extractions, limiting the
533 ability to discern the level of variability between individuals within these two groups. Thirdly,
534 the current work obtained samples from poultry ceca whereas Swarthout et al. (2022)
535 sampled poultry faeces collected from the environment.

536 Despite these methodological differences, some high-level findings correspond across both
537 studies. For example, Swarthout et al. (2022) did not report a significant difference between
538 human faecal samples derived from urban and rural locations, which parallels our findings.
539 However, Swarthout et al. (2022) did not specifically survey urban wet market workers. This
540 is significant, since our work extends previous findings, implying that the resistome of urban
541 wet market workers is not dramatically altered by regular occupational exposure to broiler
542 poultry viscera. Another study collected faecal samples from broiler farm chickens, live
543 poultry market workers and humans with low exposure to poultry in China (Wang et al.,
544 2021). Their findings indicated that humans with low exposure to live poultry markets had
545 significantly lower ARG diversity than live poultry market workers. Although we did not find
546 significant differences between low and high exposure human faecal samples; this may
547 relate to the widespread practice of keeping backyard chickens and/or general
548 environmental contamination with antibiotic resistant organisms in Bangladesh.

549 Although the total load and richness of ARGs is an important consideration when identifying
550 potential areas to focus mitigation measures for AMR, specific combinations of ARGs and
551 bacteria can present an immediate threat to human health, reflected in the WHO prioritised
552 surveillance list (WHO, 2017).

553 We found GES carbapenemases in both rural and urban wastewater. These genes are of
554 particular interest as the contigs suggested possible carriage on *Enterobacteriaceae*
555 plasmids and *Pseudomonas aeruginosa* chromosomes, both 'critical concern' organisms. A
556 recent meta-analysis indicates that GES-2 genes are among the most widely distributed
557 carbapenemase genes in aquatic environments, including wastewater, freshwater and
558 sediment (Lin et al., 2022). Likewise, GES-5 is carried by bacteria abundant in aquatic
559 environments (Manageiro et al., 2014). Although Lin et al. (2022) shows limited association
560 of GES-2 with humans, culture-based analyses have previously identified this variant in
561 nosocomial outbreaks of ESBL-producing *Pseudomonas aeruginosa* (Poirel et al., 2002).

562 More broadly, the presence of carbapenemases (GES and OXA) indicates that while the
563 total load and richness of ARGs may be greater in urban wastewater, rural wastewater
564 should not be discounted as a source of ARGs critical to One Health initiatives.

565 It is noteworthy that previous real-time PCR studies have shown considerable prevalence of
566 CTX-M-1 ESBLs, and to a lesser extent NDM-1 CRE genes, in wastewater from urban
567 markets, poultry farms and rural households (Asaduzzaman et al., 2022). Detection of these
568 genes was limited across our metagenomic dataset. This may be a consequence of
569 insufficient sequencing depth and the lack of targeted amplification for these lower
570 abundance genes. Focussing on contigs allows more confident definition of taxon-ARG
571 associations and potential ESBL/carbapenemase activity (since the complete genes can be
572 screened against variant databases). However, the use of short reads can lead to
573 incomplete assembly and many fragmented contigs with partial genes present at contig
574 ends, which can lead to underestimated variant prevalence. Alternatively, future studies may

575 consider hybrid sequencing, or employing culture-enriched metagenomics to better study
576 these clinically relevant genes (Zhang et al., 2022).

577 *Conclusion*

578 We provide in-depth contextualisation of resistomes associated with human faecal, poultry
579 caecal and wastewater samples in Bangladesh. We demonstrate that the impact of
580 environmental setting on the resistome can differ depending on sample origin. The resistome
581 of faecal samples originating from humans with and without routine occupational exposure to
582 poultry are not significantly different. However, broiler poultry from urban wet markets have a
583 significantly higher abundance of MLS and streptothricin ARGs compared to rural backyard
584 chickens. The ARG compositions of human faecal and poultry caecal samples are also
585 distinct. Overall, wastewater samples have the highest ARG richness and were under-
586 sampled in our campaign. Nonetheless, rural wastewater was identified as a source of
587 'priority one' antibiotic resistant organisms selected by the WHO, highlighting wastewater in
588 both urban and rural settings are a concern for human and animal health in Bangladesh.
589 Wastewater is an important, but poorly understood component of One Health studies on
590 AMR in Bangladesh. Further studies using long-read/hybrid or culture-enriched sequencing
591 of rural backyard and broiler poultry in farms and in urban wet markets would generate a
592 more complete understanding of how the poultry-rearing practices of medium- and small-
593 scale farms in Bangladesh determine the resistome as opposed to the contaminated
594 environments where free range chickens roam. Finally, deeper sequencing is likely to reveal
595 that wastewater in Bangladesh contains an even greater variety of ARGs than identified in
596 the present work.

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802 **Data availability:** Genomic data are available from the European Nucleotide Archive; study
803 accession PRJEB48068. Other data relating to the wider project are openly accessible at
804 <https://doi.org/10.5285/0239cdaf-deab-4151-8f68-715063eaea45> and
805 <https://doi.org/10.5285/dda6dd55-f955-4dd5-bc03-b07cc8548a3d>.

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818 **Author contributions:** ADW led the metagenomic and statistical analysis and manuscript
819 writing; EKR contributed design of the study, manuscript writing and review; MBA
820 contributed microbiological and laboratory analyses; ALN contributed metagenomic and
821 statistical analyses; JH and DS contributed microbiological and statistical analytical
822 expertise; MAI contributed design of the study, microbiological and laboratory analyses,
823 manuscript writing and review. All authors reviewed and gave critical input into the draft and
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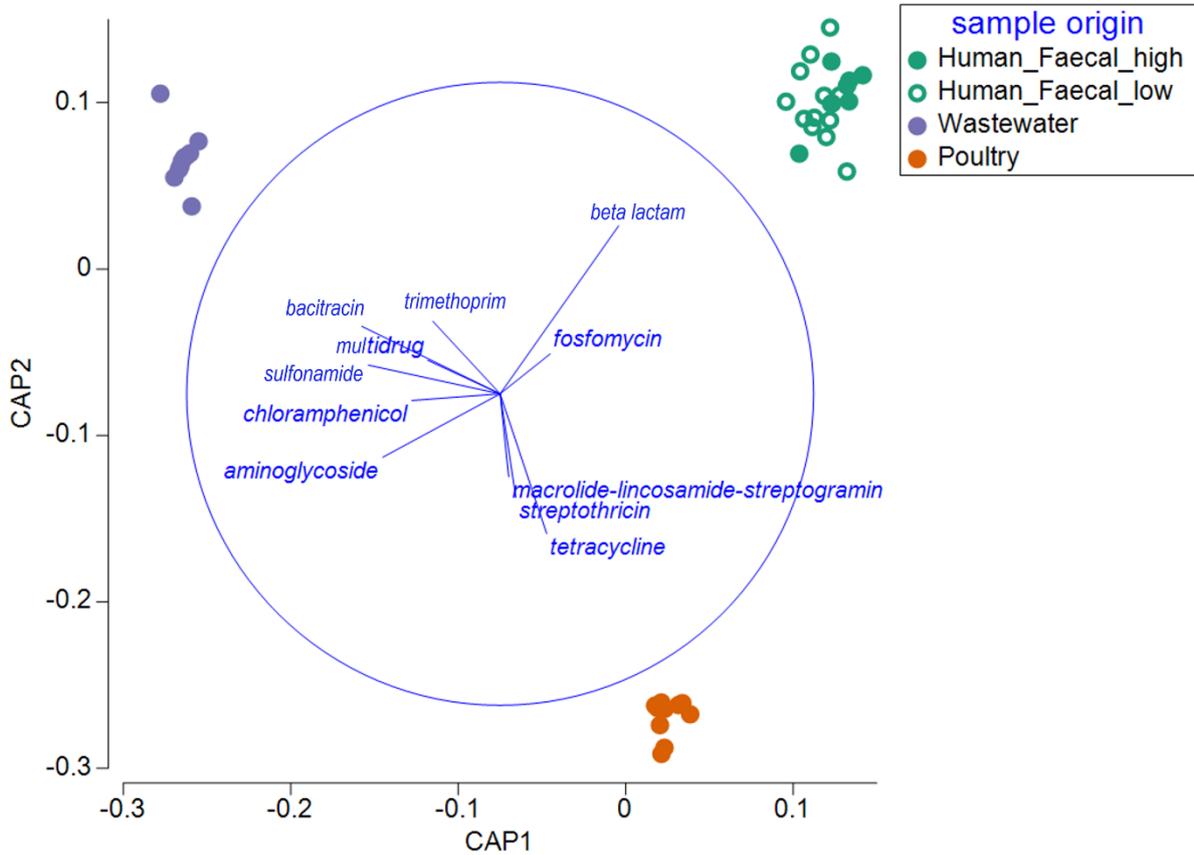
826 **Materials and correspondence:** Alexander Williams and Mohammad Aminul Islam

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829 **Figures**

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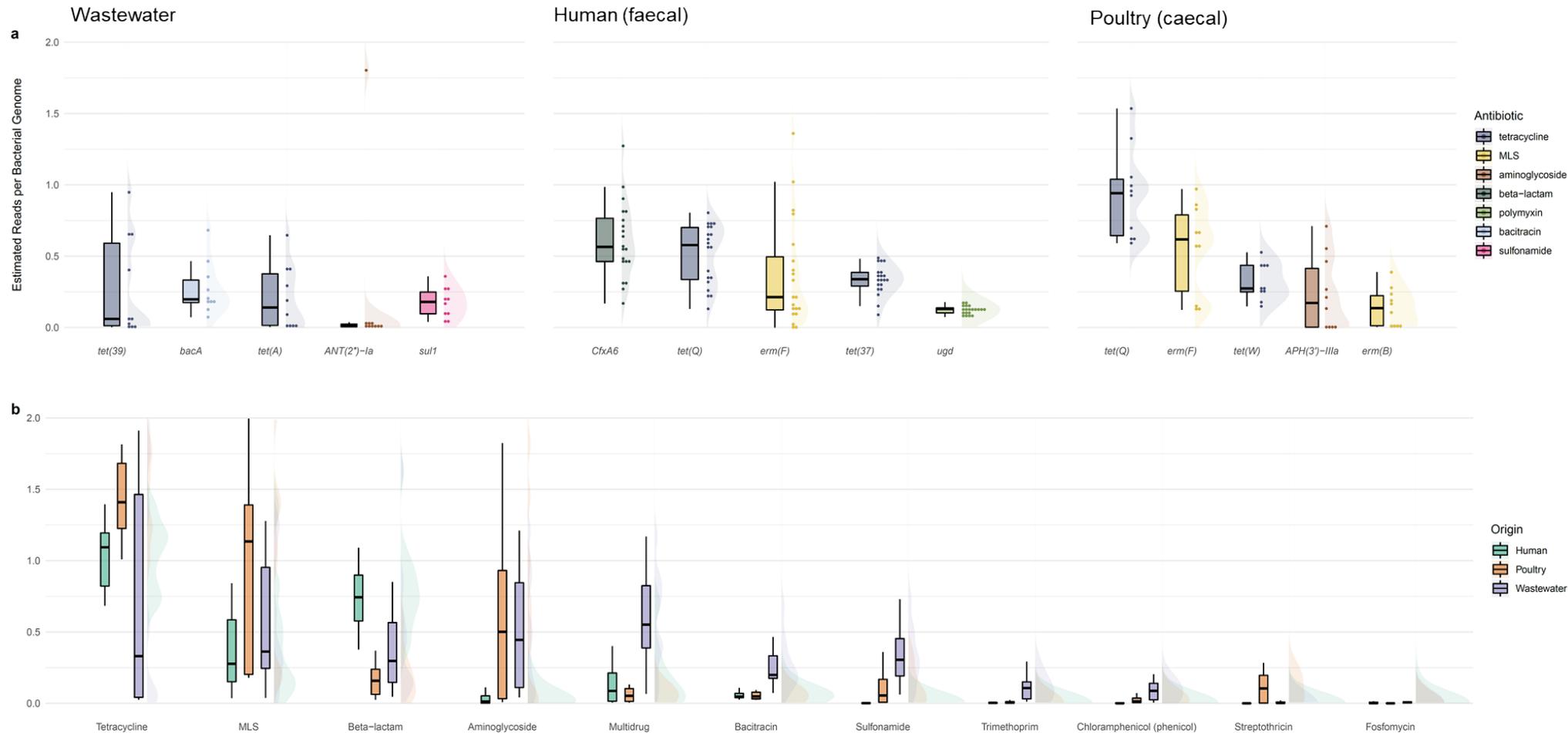
832 **Figure 1. ARG category-contingent Canonical Analysis of Principal Coordinates (CAP)**
833 **identifies significant separation of ARG assemblages between sample origins** (trace
834 statistic = 2.5, $p_{perm} = 1 \times 10^{-5}$). CAP1 squared canonical correlation (η^2) = 0.997, CAP2 η^2 =
835 0.988. Cross validation based upon a leave-one-out allocation of observations to groups
836 was associated with a 17.5% mis-classification error rate, resulting from mis-classifying
837 human faecal samples with high and low poultry exposure. Vector overlays represent
838 multiple partial correlations between CAP axes and ARG categories ($r > 0.2$). The circle has
839 a radius of $r = 1$; the relative size and position of origin is arbitrary with respect to the
840 underlying plot. Vector length and direction indicates the strength and sign, respectively, of
841 association between each ARG category and the CAP axes.

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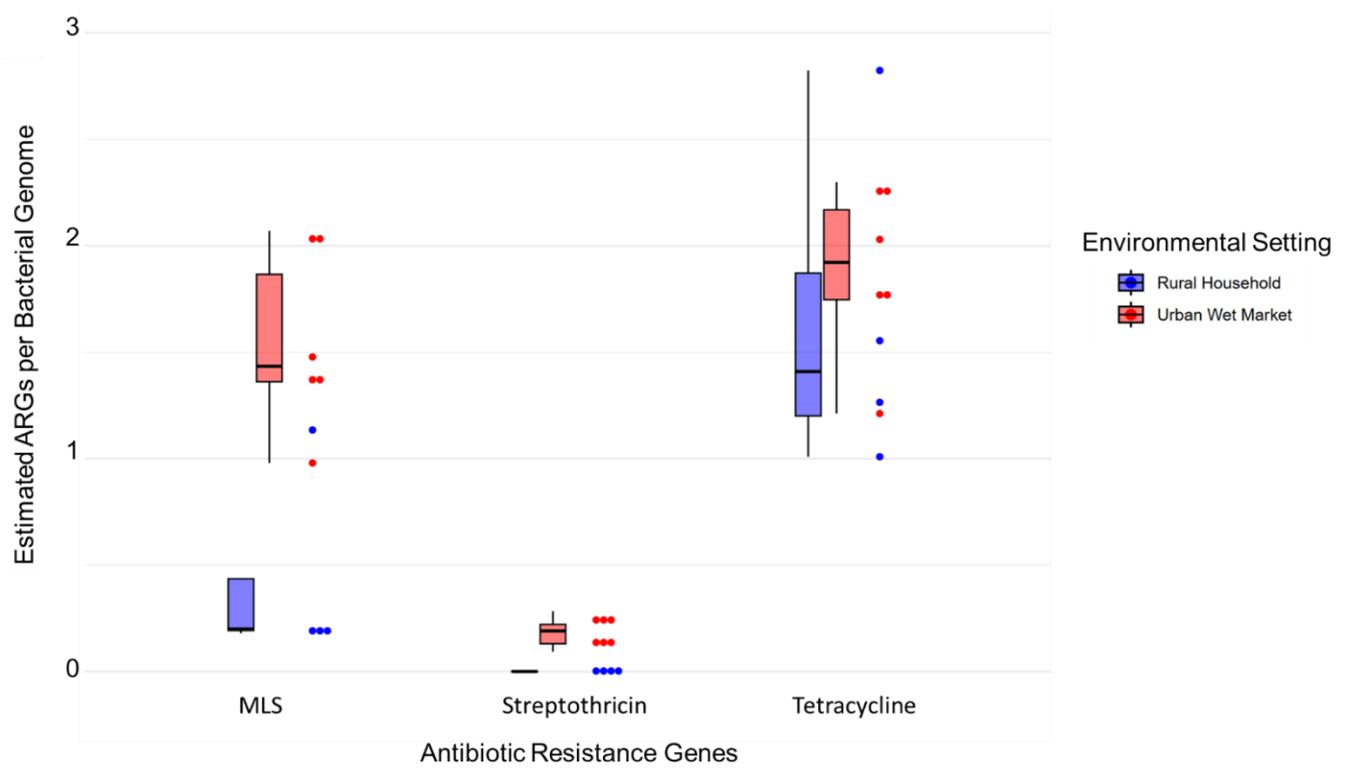
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Figure 2. Summary of abundant antibiotic resistance genes in each sample origin a) Boxplots and half-violin plots with stacked dots showing the 5 most abundant ARG subtypes by mean in each sample source (wastewater, human faecal and poultry caecal) ordered by mean estimated reads per bacterial genome; **b)** Boxplots and half-violin plots of estimated ARG copies per genome for ARG categories highlighted by canonical analysis of principal coordinates ($r > 0.2$) (compared by sample source).

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853 **Figure 3. Differences in the abundance of select antibiotic resistance gene categories**
854 **in poultry ceca samples collected from urban wet market and rural household**
855 **chickens. Boxplots and stacked dot plots of MLS (macrolide-streptogramin-lincosamide),**
856 **streptothricin and tetracycline resistance gene abundance are shown.**

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