

Metagenomic evidence for co-occurrence of antibiotic, biocide and metal resistance genes in pigs

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24 Highlights

- 25 • A comprehensive gut microbiome metagenomics analysis of 278 pigs
- 26 • Co-selection phenomena were investigated via co-occurrence patterns as a proxy
- 27 • Twenty-seven types of co-occurrences involving 131 resistance genes were detected
- 28 • Regardless of use of antibiotics, AMR can be maintained by co-occurrence with MRGs/BRGs
- 29 • Maintenance of AMR is not a random selection process but pertains to specific phylogenetic clades

31 Abstract

32 Antibiotic-resistant pathogens constitute an escalating public health concern. Hence a better
 33 understanding of the underlying processes responsible for this expansion is urgently needed. Co-selection
 34 of heavy metals/biocides and antibiotic resistance genes (ARGs) has been suggested as one potential
 35 mechanism promoting the proliferation of antimicrobial resistance (AMR). This paper aims to elucidate
 36 this interplay and exploit differences in antibiotic usage to infer patterns of co-selection by the non-
 37 antibiotic factors metals and biocides in the context of pig farming. We examined 278 gut metagenomes
 38 from pigs with continuous antibiotic exposure, only at weaning and at no exposure. Metals as growth
 39 promoters and biocides as disinfectants are currently used with little restrictions in stock farming. The
 40 pigs under continuous antibiotic exposure displayed the highest co-occurrence of ARGs and other
 41 genetic elements while the pigs under limited use of antibiotics still showed abundant co-occurrences.
 42 Pathogens belonging to *Enterobacteriaceae* displayed increased co-occurrence phenomena, suggesting that
 43 this maintenance is not a random selection process from a mobilized pool but pertains to specific
 44 phylogenetic clades. These results suggest that metals and biocides displayed strong selective pressures
 45 on ARGs exerted by intensive farming, regardless of the current use of antibiotics.

47 Keywords

Antibiotic resistance; co-selection; biocide; metal; mobile genetic element; Horizontal gene transfer

Introduction

Antimicrobial resistance in pathogenic bacteria constitutes a considerable public health concern (Baker-Austin et al., 2006). Despite making tremendous efforts to restrict usage of several key antibiotics worldwide, AMR can frequently be found in different environments (B. Li et al., 2015; Rodriguez-Mozaz et al., 2015). To control the potential threat from AMR spread, it is necessary to investigate the underlying processes responsible for this expansion.

Generally, the occurrence of AMR is largely driven by the selection pressure of antibiotics (Goossens et al., 2005). Besides that, other agents such as antibacterial biocides and heavy metals can help to maintain AMR via co-selection (Ashbolt et al., 2013; Hartmann et al., 2016; Mazhar et al., 2021). The co-selection of ARGs and metal resistance genes (MRGs) has been commonly observed in contaminated soil (Song et al., 2017), water (Icgen and Yilmaz, 2014), and industrial environments (Mazhar et al., 2021; Stepanauskas et al., 2005) with enriched metals. Some literature has demonstrated that ARGs had a stronger correlation to some metals than to their corresponding antibiotics (Ji et al., 2012). Unlike antibiotics, metals can constitute a long-term selection pressure due to metals not being degradable (Stepanauskas et al., 2005), which can bring stability to the ARG pool in both engineered and natural systems. Compared to metals, biocides have received less attention as a potential co-selection agent. Biocide substances have been employed for centuries as disinfectants for medical equipment and as preservatives in pharmaceuticals, cosmetics and food (SCENIHR, 2009), especially biocides with limited toxicity to animals such as quaternary ammonium compounds (QACs), biguanides and bisphenols (Grande Burgos et al., 2013). However, some research has suggested excessive use of QACs increased bacterial tolerance to antibiotics (Buffet-Bataillon et al., 2012; Tandukar et al., 2013). Metals and biocides can co-select for antibiotic-resistant bacteria by several mechanisms (Chapman, 2003): resistance genes

are physically located on the same genetic element (co-resistance); the same gene confers resistance to both antibiotics and biocide/metal (cross-resistance); biocide resistance genes (BRGs)/MRGs share the same regulatory mechanism with ARGs (co-regulation).

The concurrent transfer of ARGs and BRGs/MRGs via mobile genetic elements (MGEs) is another challenging public health problem, which has aggravated the persistence and spread of ARGs (Benveniste and Davies, 1973). So far, research on the co-transfer of these three genetic determinants has been limited, although there are plenty of opportunities for this to occur. Stock farming is one such potential hot spot for the dissemination of combinations of ARGs, BRGs/MRGs, and MGEs due to the widespread use of antibiotics, heavy metals and biocides in food feed (Clark, 2004). Antibiotics and metals (especially zinc and copper) are commonly used as growth promoters in stock farming (Gaskins et al., 2002). However, the use of antibiotics as growth promoter has been banned in the European Union since 2006 (Lekagul et al., 2019) and therefore farmers have given more attention to alternatives such as copper and zinc (Jensen et al., 2016). This has resulted in the accumulation of these metals in soils where pig slurry was applied. In contrast, China does not prohibit the use of antibiotics as growth promoter (Lekagul et al., 2019) in stock farming and at the same time, the extensive use of heavy metals in feed in China has greatly increased over the last couple of decades (Wang et al., 2013). Many of the plants used for livestock are dependent on biocides to control weeds, insects and diseases (Clark, 2004) and biocides are also used for disinfection of animal housing facilities (Montfoort JA, Poel P van der, n.d.). However, investigations on the co-selection of heavy metals/biocides and antibiotics used in animal farms, which are closely connected to human health, are still scarce (Seiler and Berendonk, 2012), especially under various exposure levels of antibiotics.

Genome sequencing has often complemented analysis of phenotypic outcomes (Song et al., 2017; Stepanauskas et al., 2006) to clarify the genetic determinants of resistance, and indeed large-scale efforts

have investigated repositories of publicly available genomic data and established the co-presence of both metal- and antibiotic-resistance (Li et al., 2017; Pal et al., 2015). These repositories are comprehensive, but the information about the origins of the sequenced isolates varies, making it difficult to infer that genes of interest are present in the underlying populations as a whole. In this study, we attempted to make use of publicly available metagenomics data of 278 pigs in the different farms from three countries (Xiao et al., 2016). The antibiotic usage patterns in these pigs have been well-characterized. The pigs in French farms were fed organically and pigs in Danish farms received antibiotics only at weaning, while pigs in the Chinese farm were continuously fed on antibiotics. Metals and biocides were used in all farms. We hypothesized that the use of metals and biocides would exert strong selective pressures on ARGs in intensive farming regardless of antibiotic use.

Methods

Sample information, data collection and pre-processing

Table S1 illustrates the antibiotic exposure on 278 pigs in different farms (Xiao et al., 2016). Seventy-eight pigs in the Chinese farm were continuously exposed to various antibiotics, while one hundred pigs in Danish farms were only exposed at weaning. One hundred pigs in French farms were raised completely organic, although a subset of them was raised on farms with previous use of antibiotics.

The clean Illumina Hiseq paired reads of 278 pig gut metagenomes, which have removed adaptor contamination, low-quality reads and pig genomic DNA, were retrieved at <https://www.ebi.ac.uk/ena/data/view/PRJEB11755> from European Bioinformatics Institute (EBI). Assembly of contigs was done with megahit (version 1.1.3) with default options (D. Li et al., 2015). Any contigs shorter than 500bp were filtered out. The prediction of open reading frame (ORF) in the contigs was performed with prodigal (version 2.6.2) in META mode (Hyatt et al., 2010).

ARG, BRG, MRG, and MGE prediction

The antibacterial biocide and metal resistance genes database (BacMet, version 2.0) was used for the predictions of BRG and MRG (Pal et al., 2014): the amino acid sequences of predicted ORFs were subjected to similarity searches against the BacMet database with diamond search in the more sensitive mode and k1 option (Buchfink et al., 2014). Only BRGs and MRGs with at least 90% identity and a maximum e-value of 1×10^{-3} were retained. The comprehensive Antibiotic Resistance Database (CARD database, version 3.0.0) was used to detect ARGs: the amino acid sequences of predicted ORFs were aligned with the CARD database through RGI identifier with diamond as aligner (Jia et al., 2017). Only ARGs with “Strict” or “Perfect” significance cut-off were preserved for further analysis. MGE homologs were characterized using PFAM (Finn et al., 2016) and TnpPred (Riadi et al., 2012) databases through HMMER v3.1b2, with an e-value of 1×10^{-5} as a threshold (Sáenz et al., 2019). If the predicted ORF had more than one hit for MGE homologs, the hit with the lowest e-value was retained. In the present study, MGEs were classified into 9 categories according to functions: conjugative transposon, integrase/integrase-related, phage integrase, recombinase, resolvase, RteC (related to tetracycline conjugative transposon), transposase/transposase-related, transposition related, transposon breakage related.

Gene coverage/abundance calculation

Clean reads were aligned against the reference ORF nucleotides with BWA aligner (Li, 2013). The alignment summary statistics were calculated using Samtools idxstats (Li et al., 2009). For the comparisons of gene coverage/abundance between samples, GCPM (Gene Coverage Per Million) value was used to normalize the mapped reads for sequencing depth and gene length. The formula is GCPM

$$(t) = \frac{(\text{counts}(t)/\text{gene length}(t)) \times 10^6}{\sum_i \text{counts}_i / \text{gene length}}$$

where GCPM (t) is the GCPM value of gene t, counts (t) is the number of

mapped reads to the gene (t), gene length (t) is the length of gene (t), n is the number of all the predicted ORFs.

Statistical analysis and R application

All statistical analysis and data sorting in the study were done in R (R Development Core Team, 2011). Between-individual diversity (β -diversity) of gene coverage/abundance was evaluated by Bray-Curtis dissimilarity matrices through R function “vegdist” in the package “vegan” (Oksanen et al., 2008). β -diversity matrices were ordinated by PCoA plot (R function “plot_ordination” in R package “phyloseq”) (McMurdie and Holmes, 2013). Permutation multivariate analysis of variance (PERMANOVA) was used to compare the differences in β -diversity between groups when FDR correction was required (R function “pairwise.adonis” in R) (Arbizu, 2017). Within-individual diversity (α -diversity) was measured by observed richness of gene coverage/abundance. One-way ANOVA (R package “stats”) was used to compare the differences in α -diversity among groups. TukeyHSD test (R package “stats”) was used to do pairwise comparisons of α -diversity. The pairwise comparison of gene abundance between three groups was done using the Wilcoxon rank sum test (R function “pairwise.wilcox.test”). Venn diagram was plotted using R function “draw.triple.venn” in R package “VennDiagram” (Chen and Boutros, 2011). Fisher’s exact test was used to test the statistical significance of the number of contigs between groups (R function “fisher.test” in R package “stats”).

Co-occurrence analysis

The resistance genes located in the same assembled contig were considered as co-occurring. Co-occurrence was considered trustable if the same association was present in at least two different contigs.

Contig source screening

Fifteen million contigs from 138,683 bacterial genomes in the NCBI Reference Sequence Database (RefSeq) were downloaded (19.12.2018) and the downloaded bacterial genomes were indexed to generate a BLAST database with makeblastdb. Blastn (version 2.6.0) was used to search for the contigs carrying ARGs, MGEs, BRGs/MRGs detected in the study from the reference genomes. A minimal blast similarity of 95%, a shortest alignment length of 10kb and a biggest e-value of 10e-10 were used to filter blast hits. R function “heat_tree” in “metacoder” package (Foster et al., 2017) was used to plot the taxonomic tree to visualize the bacterial contig source. The tree branch represents the affiliation relation between taxa.

Results

Gut microbiome of pigs in the Chinese farm contained the highest abundance of resistance genes

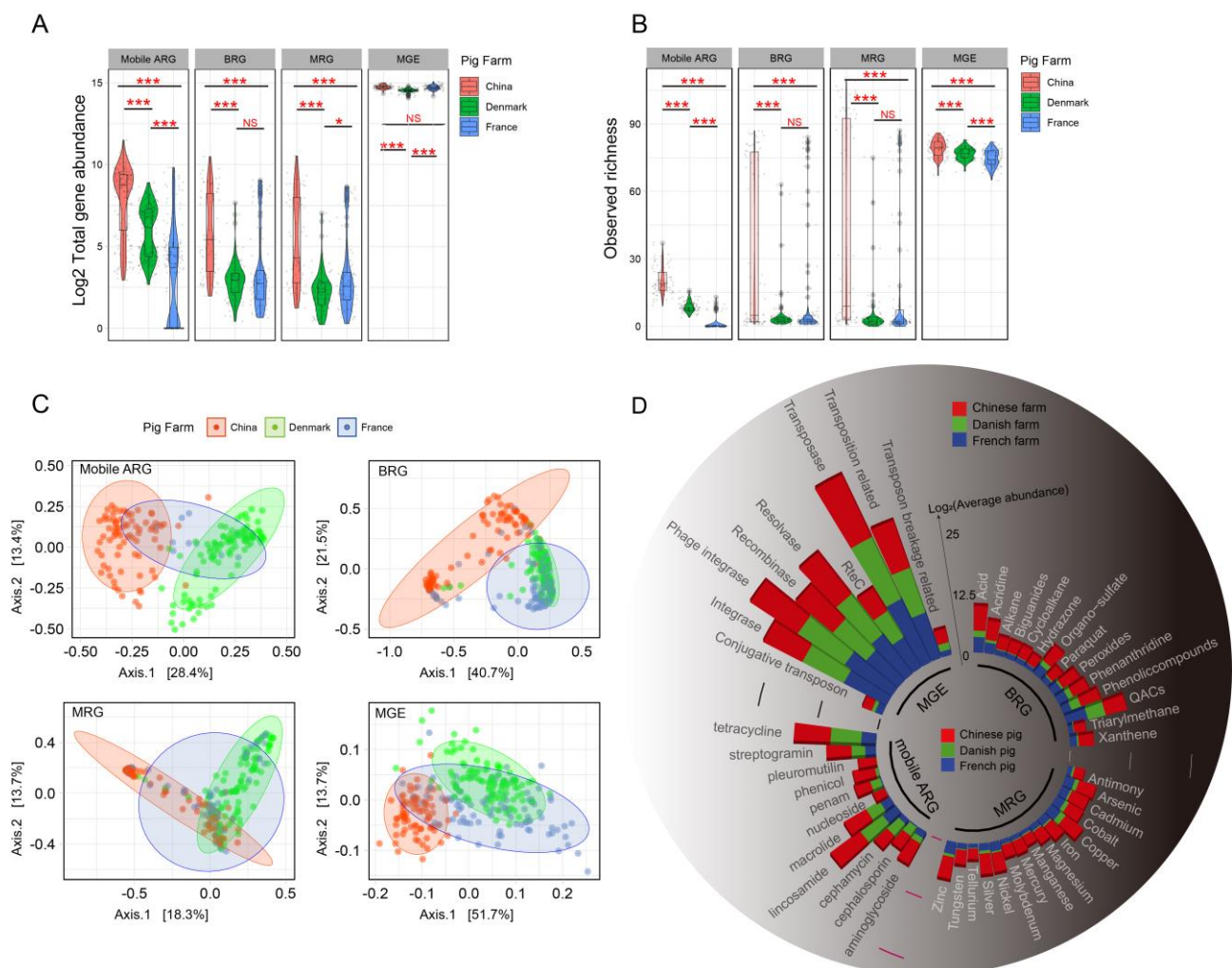
The total loads of BRGs, MRGs, and potentially mobile ARGs (ARGs found together with MGEs in one contig) varied significantly between different farms (Fig. 1A). When looking at all resistance determinants, the gut microbiome of pigs in the Chinese farm carried the highest load (Pairwise Wilcoxon rank-sum test, FDR-adjusted; Pigs in Danish, French and Chinese farms were represented by D, F and C, respectively. Mobile ARG: C vs D, $P < 2e-16$; C vs F, $P < 2e-16$; D vs F, $P < 2e-16$. BRG: C vs D, $P = 4.2e-14$; C vs F, $P = 4.5e-10$; D vs F, $P = 0.53$. MRG: C vs D, $P = 5.2e-14$; C vs F, $P = 3.0e-07$; D vs F, $P = 0.027$. MGE: C vs D, $P < 2e-16$; C vs F, $P = 0.073$; D vs F, $P < 2e-16$). The pigs in Danish farms had the second-highest abundance of mobile ARGs and similar abundance of BRG to the pigs in French farms. The pigs in French farms had the second-highest abundance of MRGs. Notably, only 10 of 100 pigs in French farms had mobile ARGs in their gut microbiome, though the total abundance of MGEs in French pigs was similar to Chinese pigs.

194 We calculated observed richness for each resistance determinant in different farms (Fig. 1B). For each
 195 gene type, pigs in the Chinese farm were found to have a higher diversity followed by the pigs in Danish
 196 farms (TukeyHSD test; Mobile ARG: C vs D, $P < 1e-07$; C vs F, $P < 1e-07$; D vs F, $P < 1e-07$. BRG: C
 197 vs D, $P < 1e-07$; C vs F, $P < 1e-07$; D vs F, $P = 0.11$. MRG: C vs D, $P < 1e-07$; C vs F, $P < 1e-07$; D vs
 198 F, $P = 0.08$. MGEs: C vs D, $P = 2.34e-05$; C vs F, $P < 1e-07$; D vs F, $P = 8.28e-05$). However, pigs in
 199 French farms had similar BRG and MRG diversity compared to pigs in Danish farms.

200

201 In general, mobile ARGs, BRGs, MRGs and MGEs had a pronounced separation in composition based
 202 on farms (Fig. 1C) (pairwise Adonis test, FDR-adjusted; Mobile ARG/BRG/MRG/MGE: C vs D, $P =$
 203 0.003 ; C vs F, $P = 0.003$; D vs F, $P = 0.003$). All detected ORFs conferred resistance to 20 metals, 36
 204 biocides, 34 antibiotics, and encoded 9 types of mobile genetic elements. Fig. 1D shows the most
 205 abundant genes conferring resistance to 13 metals, 14 biocides, and 10 antibiotics. Overall, of the MRGs
 206 detected, genes conferring Zn and Cu resistance were the most abundant. Among BRGs, genes encoding
 207 resistance to QACs were the most abundant. When we investigated the potentially mobile fraction of the
 208 ARG pool, lincosamide and tetracycline resistance genes were found to be the most abundant and the
 209 transposases to be the most common trait related to mobility mode.

Fig. 1.



separately counted in the individual function group; for example, the gene with AMR and biocide resistance (BR) was included both in AMR and BR abundance plots. Labels on the stacked bars describe the metals and biocides associated with BRGs and MRGs, different drug classes associated with mobile ARGs and 9 categories of MGEs. Heights of the stacked bars represent the mean value of the total abundance in log-scale.

Surveying assembled contigs carrying co-occurrence of ARGs and BRGs/MRGs/MGEs

We then investigated the co-occurrence of resistance genes on contigs, in the categories of ARGs, BRGs/MRGs, and MGEs to evaluate the potential for co-selection (Fig. 2A). Evidently, the pigs in the Chinese farm had the largest amount of overlap from different categories, with 61 contigs harboring all 3 different categories of genes, while this was not observed in any of the pigs in French and Danish farms. MGEs were the largest category and had ample overlap with other categories. The pigs in the Chinese farm had the highest proportion of contigs carrying multiple co-occurrences (Fig. 2B) (Fisher's exact test; ARGs+MGEs: C vs D, $P < 2.2e-16$, odds ratio = 2.69; C vs F, $P < 2.2e-16$, odds ratio = 18.5; D vs F: $P < 2.2e-16$, odds ratio = 6.87; ARGs+BRGs/MRGs: C vs D, $P < 2.2e-16$, odds ratio = 11.8; C vs F, $P < 2.2e-16$, odds ratio = 2.04; D vs F: $P < 2.2e-16$, odds ratio = 0.17; ARGs+BRGs/MRGs+MGEs: C vs D, $P < 2.2e-16$, odds ratio = NA; C vs F, $P < 2.2e-16$, odds ratio = NA; D vs F: $P = 1$, odds ratio = 0; BRGs/MRGs+MGEs: C vs D, $P < 2.2e-16$, odds ratio = 9.48; C vs F, $P < 2.2e-16$, odds ratio = 36.6; D vs F: $P = 0.0007$, odds ratio = 3.86). The pigs in French farms had a significantly higher proportion of contigs with ARGs and BRGs/MRGs than the pigs in Danish farms. In contrast, the pigs in Danish farms had a higher proportion of contigs carrying resistance genes together with MGEs than the pigs in French farms. Compared to BRGs/MRGs, a larger proportion of ARGs generally tended to sit together with MGEs in the same contigs (Fisher's exact test; $P < 2.2e-16$, odds ratio: 2.9) (Fig. 2C).

Fig. 2.

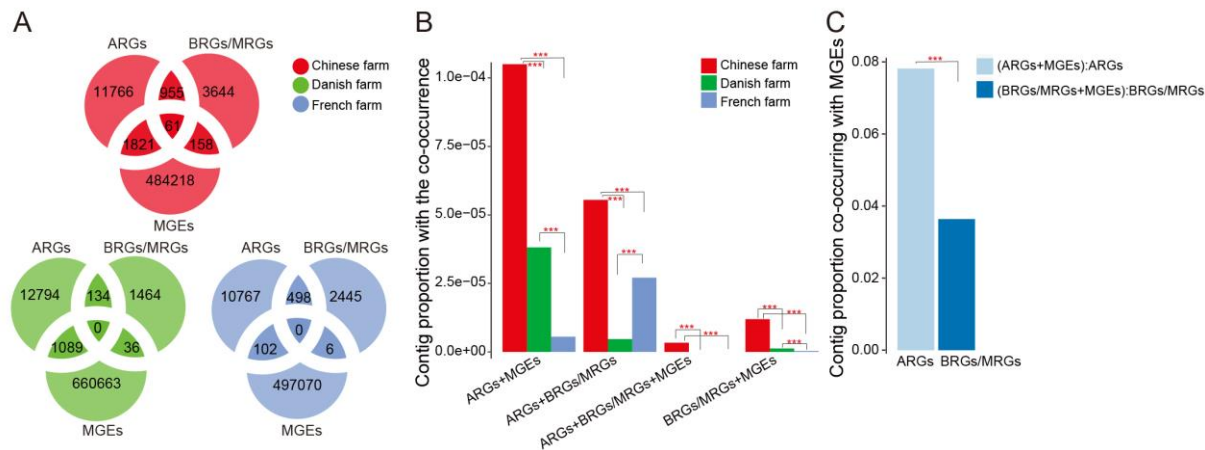


Fig. 2. Overview of co-occurrences of ARGs, BRGs, MRGs and MGEs in assembled contigs. A).

Venn diagram showing the number of contigs carrying ARGs, BRGs/MRGs, MGEs and their combinations in pigs. **B).** The proportion of contigs with co-occurrences of ARGs and BRGs/MRGs,

ARGs and MGEs, BRGs/MRGs and MGEs, BRGs/MRGs, ARGs and MGEs in pigs. Asterisks stand for significant statistical difference between groups (Fisher's exact test; $P < 0.05^*$, $P < 0.01^{**}$, $P < 0.001^{***}$).

C). The proportion of contigs carrying ARGs and BRGs/MRGs with MGEs. Asterisks stand for significant statistical difference between groups (Fisher's exact test; $P < 0.05^*$, $P < 0.01^{**}$, $P < 0.001^{***}$).

Co-localization of ARGs and BRGs/MRGs/MGEs in assembled contigs

To further verify the co-occurrence of ARGs and BRGs/MRGs/MGEs on the contigs, we visualized the gene organization on representative contigs (Fig. 3A). According to resistances carried in each representative contig, there were six major co-occurrence combinations and 27 co-occurrence subtypes. We selected the contig carrying the most complete gene distribution in each of 27 subgroups as the representative contig. All the co-occurrences involved 131 resistance genes to 17 metals, 28 biocides and 25 antibiotics. Of these, Cu, Ni and Zn resistance genes, fluoroquinolone, penam and tetracycline resistance genes, acid and QAC resistance genes were the most co-occurring genes. The information about these genes has been summarized in Table S2.

263

264 Co-occurrences in the pigs from the Chinese farm were detected to be most common, in terms of
 265 both the number of contigs carrying co-occurrences and the type of co-occurrence (Fig. 3A). We
 266 identified all 27 co-occurrence subtypes in the Chinese farm, 20 subtypes in French farms, and 9 subtypes
 267 in Danish farms. Some co-occurrences affiliated with the same subtype were only partly present in the
 268 presentative contig. Therefore, to further investigate the abundance of co-occurrences, we plotted a
 269 network to show the frequency of co-occurrences between resistance genes in all the contigs (Fig. S1).
 270 The detailed information for the network has been summarized in Table S3. As shown in Fig. S1,
 271 ARGs in the pigs from Danish farms mostly co-occurred with BRGs, which mostly tended to be
 272 functionally associated, such as multidrug efflux transporter genes *acrAB* and its regulator gene *acrR*.
 273 Notably, we detected an assembled contig carrying one colistin resistance gene *MCR-4* and Cd/TBT
 274 resistance gene *ygjW* in the pigs from Danish farms.

275

276 Some co-occurrences were only detected in the pigs from the Chinese farm (Fig. 3A) – for example, the
 277 co-occurrence of fluoroquinolone ARG *CRP* with the Ni resistance operon *nikA/B/C/D/E/R* and Zn
 278 resistance gene *zntA*; the co-occurrence of aminoglycoside ARG *kdpE* with nine Cu resistance genes
 279 *cusA/B/C/F/R/S*, *cutE/F*, *corC*; and the co-occurrence of streptogramin ARG *vgaC* with seven Cu
 280 resistance genes *pcoA/B/C/D/E/R/S*. Two archetypal Class I clinical integrons carrying 3' conserved
 281 segment (CS) of an antiseptic-resistance gene *qacEΔ1*, a sulfonamide resistance gene *sul1*, 5' CS of the
 282 integrase, as well as gene cassettes *aadA2-dfrA17* (Integron A) or an IS6 transposase gene (Integron B)
 283 were only detected in the pigs from the Chinese farm. An analogous structure consisted of *qacF*, *sul3*, an
 284 integrase, and gene cassette *dfrA12-aadA2-cmlA6-aadA* was only detected in the pigs from the Chinese
 285 farm as well (Integron C). To verify whether these three integrons are located on plasmids, we mapped
 286 the three representative contigs against the PLSDb plasmid database (Galata et al., 2019). It turned out
 287 that Integron C and Integron B were very similar to DNA fragments in 180 and 73 plasmids, respectively

288 (Mash search, Max distance=0.1). Integron C was frequently found in *E. coli*, *Klebsiella pneumonia* and
289 *Salmonella enterica* (Fig. S2-a and Table S4). The bacteria harboring Integron B covered a wide range of
290 species such as *E. coli*, *Pseudomonas aeruginosa*, *Klebsiella pneumonia* and *Enterobacter cloacae* (Fig. S2-c and Table
291 S5). The gut microbiome in the pigs from the Chinese farm harbored some co-occurrences related to
292 polymyxin including colistin resistance genes. For example, the polymyxin ARG *ugd* was located in the
293 same contig with Co-Ni-Fe related resistance genes *rtnA/B/R*, multidrug efflux system *MdtA/B/C*,
294 *BaeS/R* two-component system; Polymyxin resistance genes *pmrA/B/C*, colistin heteroresistance related
295 *soxR/S*, co-occurred with Cu resistance gene *cutA*, acid resistance gene *actP*, Cd-Hg-TBT-H₂O₂-HCl
296 resistance gene *PA0320* and a phage integrase.
297
298 Some co-occurrences were detected in pigs from both French and Chinese farms. For example, the
299 polymyxin resistance gene *pmrF*, fosfomycin resistance gene *GlpT*, microcin J25 resistance gene *YojI* were
300 found on the same contig with a QACs- H₂O₂ resistance gene *kpnO* and a Fe resistance gene *pmrG*; The
301 genes encoding the two-component regulatory system *CpxAR* co-occurred with Fe-Zn resistance gene
302 *fieF*, Se-H₂O₂ resistance gene *sodA* and Sb-As-glycerol uptake and resistance gene *glpF*; Cu resistance gene
303 *copA*, *cueR* and Fe-H₂O₂ resistance genes *fetA*, *fetB* were frequently detected to co-occur with genes
304 encoding the *AcrR/A/B* multidrug efflux pump operon; And acid resistance genes including *gadE*, *hdeA*,
305 *hdeB*, *mdtE*, *mdtF*, *gadW*, *gadX*, *gadA* were located in the same contig with Zn-Te resistance gene *pitA* and
306 As resistance gene *arsA/B/R*; Beta-lactam resistance gene *PBP3* was found to co-occur with two Cu
307 resistance gene *corD*, *cueO*, Fe resistance gene *ygiH* (only in Chinese pigs), one BRG *ostA* and one
308 transposase gene; A Cd-H₂O₂-HCl resistance gene *ybcN* co-occurred with the efflux pump system *AcrE/F*
309 and its repressor *AcrS*; A broad spectrum MDR ABRG *mdfA* was detected in the same contig with Cr-
310 nitrofurantoin resistance gene *nfxA*.
311

312 In this study, we plotted a network to demonstrate the co-occurrences between ARGs and MGEs in
313 the same contigs (Fig. 3B). In total, 282/1261 ARGs were co-located with MGEs on the same contigs,
314 referred to as mobile ARGs. Mobile ARGs are abundant in the pigs from Chinese and Danish farms,
315 especially transposase. The mobile lincosamide resistance gene *lnuC* was most detected in all three
316 farms. The mobile cephamycin resistance genes *CfxA2/CfxA6* and penam resistance gene *ACI-1* were
317 also abundant in the pigs from Danish and Chinese farms. Some tetracycline resistance genes
318 (*tet(40)/(42)/(C)/(D)/44/W/O/Q/X/adeF*), macrolide resistance genes (*ermA/B/F/G/Q*) and
319 aminoglycoside resistance genes (*ANT(6)-Ia/ANT(6)-Ib/APH(2)-IIa/APH(3)-Ib/APH(3)-*
320 *IIIa/APH(6)-Id/aadA/aadA2*) were also frequently detected to co-locate with MGEs in the same contigs.
321 The polymyxin resistance gene *pmrF* was found to be in the vicinity of genes encoding transposases
322 in the same contig. Colistin resistance-related genes *mexB* and *oprM*, and polymyxin B resistance gene
323 *eptA* were also detected to co-occur with MGEs in Chinese and Danish pigs. The detailed information
324 for the network has been summarized in Table S6.

Fig. 3.

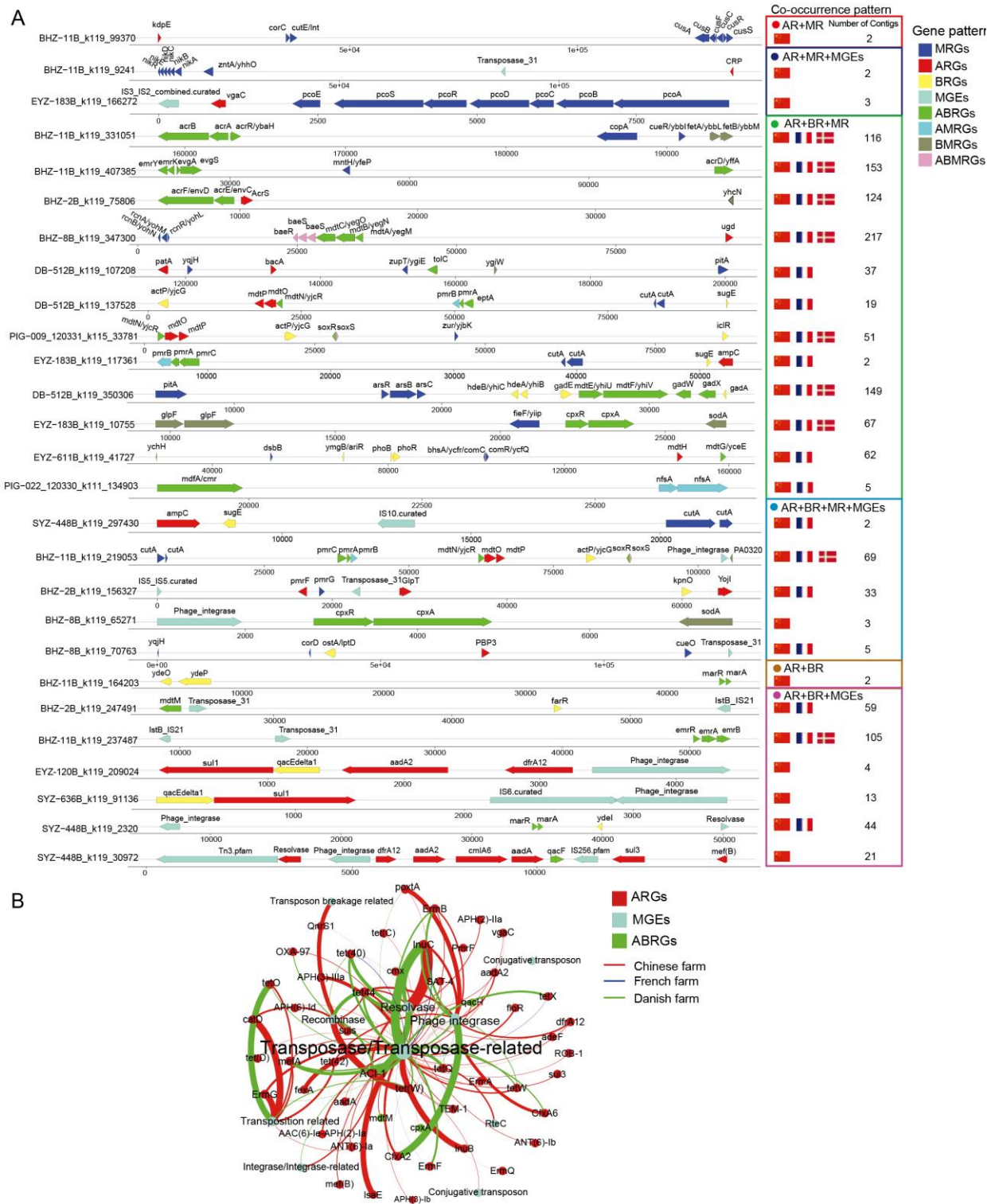


Fig. 3. The patterns of Co-occurrence in assembled contigs. A) The arrangement of resistance genes within the co-occurrences-carrying representative contigs. Only those co-occurrences present in at least

two different contigs are listed. ABRG stands for the gene conveying resistance to antibiotic and biocide; AMRG stands for the gene conveying resistance to antibiotic and metal; BMRG stands for the gene conveying resistance to biocide and metal; ABMRG stands for the gene conveying resistance to antibiotic, biocide and metal. The location and size of genes are proportional to the actual condition; the country flag stands for the location of the farm raising the pigs that have the corresponding co-occurrences and its row order stands for the decreasing co-occurrence frequency. **B)** Network of co-occurring ARGs and MGEs on the same contigs. For view purposes, only the co-occurrences between ARGs and MGEs which are present in at least 5 different contigs are shown in this network. The size of node, node label and the weight of edge are proportional to the number of co-occurred contigs.

Mobilizable contigs carrying ARG, BRG/MRG and MGEs are concentrated in *Enterobacteriaceae*

To investigate which clades of bacteria tended to maintain and spread ARGs in the form of co-selection and also confirm the accuracy of contig assembly, we blasted the contigs carrying ARGs, MGEs, BRGs/MRGs against the NCBI RefSeq database consisting of 138, 683 genomes. All the contigs could be found in reference genomes. The tree in Fig. 4 demonstrated that those mapped bacteria and the size of nodes was proportional to the mapped occurrence of the bacterial taxa. Proteobacteria was the most often hit phylum, under which family *Enterobacteriaceae* had the biggest class branch. Notably, the opportunistic pathogen *E. coli* and the pathogen *Shigella flexneri* were the most mapped species (Fig. 4).

Fig. 4.

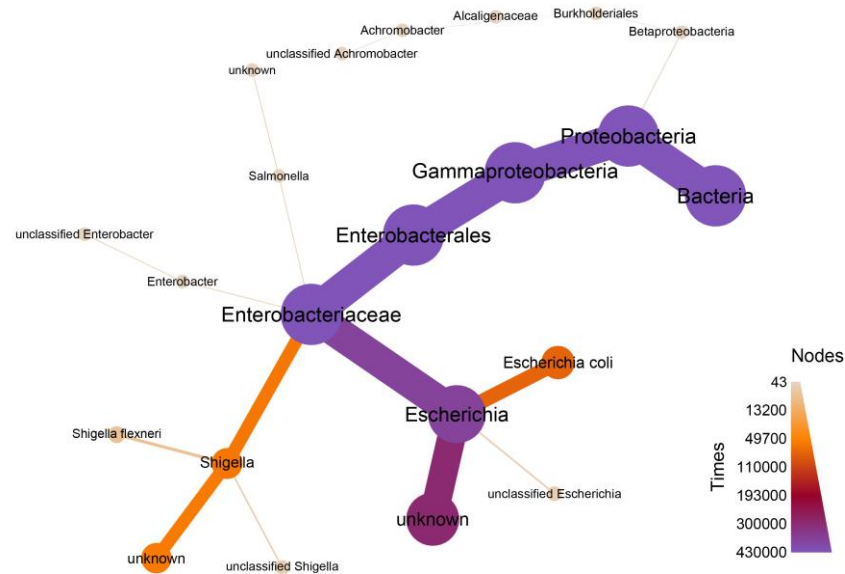


Fig. 4. Tree showing bacteria harboring contigs with ARGs, MGEs, MRGs/BRGs. The node size, node label size, and edge weight were proportional to mapped frequency. All the bacteria taxa that had a successful match have been shown in Table S7.

Discussion

The current study provides a comprehensive analysis based on the genetic linkage of bacterial co-occurrence profiles between ARGs and other genetic elements including MRGs, BRGs and MGEs in the gut microbiome of 278 pigs from Danish, French and Chinese farms. The antibiotic feeding mode of pigs in the Chinese, Danish and French farms can stand for an unmonitored mode, European standard mode and organic mode, respectively. Compared to antibiotics, the usage of metals as a growth promotor in pig farming has received little restriction around the world. But even worse, unmonitored biocide usage as a disinfectant in the pigsty and its residual in vegetable feed did so far not cause much attention. In this study, we found a positive association between the extent of antibiotic use and the load of mobile ARGs. The gut of pigs in the Chinese farm contained the most abundant mobile ARGs, followed by pigs from Danish farms and lastly the pigs from French farms. The abundance and type of MRGs and BRGs did not vary a lot among the different farms. Pigs in all farms had abundant MRGs and BRGs, especially

genes encoding resistances for Zn, Cu, Ag, Ni, As, Cd, QAC. In general, Cu and Zn are the most common growth promoters in animal feeding and residuals can be always found in manure (Mazhar et al., 2021). QAC resistance BRG was frequently found in the pigs from all farms. Over the past decade, the use of QACs as detergent, disinfectant and preservative has dramatically increased in industry, hospitals, and cosmetics (Buffet-Bataillon et al., 2012), accompanied by an elevated occurrence of QAC resistance genes in many bacteria (Bischoff et al., 2011; Heir et al., 1999; Langsrud et al., 2003; Seier-Petersen et al., 2015; Zmantar et al., 2011). Unfortunately, we cannot provide an effects size estimate using our study, since no quantitative or qualitative data on the use of these compounds was collected. In addition to the widespread use of QACs, the spread of these resistance genes via HGT among prokaryotes may be another reason for their widespread presence since a significant correlation between QAC resistance genes and MGEs was detected in this study. The spread of resistance determinants through HGT is a favored way for microbes to adapt to complex environmental pressures (Ye et al., 2017). In this study, around 10.8% (282/1261) ARGs, which co-occurred with MGEs especially genes encoding transposase, had the potential to be transferred via HGT. Some previous work had shown that the levels of transposases were highly correlated to the abundance of ARGs from other environments (Aziz et al., 2010; Zhu et al., 2013).

In this study, co-selection phenomena were investigated via co-occurrence patterns as a proxy. We found that regardless of antibiotics use or little/no antibiotic use, AMR seems to be maintained by co-occurrence with MRGs/BRGs. Despite little/no antibiotic use in Danish and French farms, cross-resistance, for example, *mdtA/B*, *mdtE/F*, *mdtN/O*, *cpxA/R*, *acrA/B/R*, *acrE/F*, *emrA/B/R*, *emrK/Y*, *gadA/X/W*, *baeR/S*, *soxS/R*, which have multidrug resistance towards antibiotics, biocides and/or metals, can still help to maintain AMR under exposure stress from biocides and/or metals. Of these, some multidrug resistance genes can encode multiple-component regulatory systems that mediate co-regulation as well. For example, the BaeRS two-component regulatory system can activate transcription of many

resistance genes (Nishino et al., 2005) such as *mdtA/B/C/D* transporter gene clusters and *acrD* (Baranova and Nikaido, 2002; Nagakubo et al., 2002). The CpxAR two-component regulatory system can activate transcription of the multiple antibiotic resistance regulatory operon *marA/B/R* (Weatherspoon-Griffin et al., 2014). *soxS* can enhance the expression of the Zn uptake system ZnuACB (Warner and Levy, 2012) and the AcrAB-TolC multidrug efflux protein complex (Pérez et al., 2012). In addition to cross-resistance, co-resistance phenomena detected in the pigs especially from French farms further highlighted the role of metals/biocides on the maintenance and spread of ARGs. These co-resistance patterns were either functionally connected such as in *acrA/B/R* or probably reflected the historical and current chemical environments (Ye et al., 2017). Zhu *et al* found that the co-selection of Cu, Zn, tetracycline resistance determinants and MGEs would be favored in exposed microbial communities due to the use of Cu, Zn, and antibiotics such as tetracycline in pig farming (Zhu et al., 2013). Accordingly, we speculate that microbes in the gut of pigs faced the selection pressures from metals, antibiotics and biocides including Cu, Ni, Zn, fluoroquinolone, penam, tetracycline, cephalosporin, Acid, QACs, Acridine, *etc.* In summary, antibiotics are not the sole factor for the spread and preservation of ARGs in their ecological environment.

Notably, we found some polymyxin resistance genes and their co-occurrences with other resistance genes. As we know, polymyxins have reemerged as a final line of defense against Gram-negative 'superbugs' (Sun et al., 2018). The co-occurrences of polymyxin resistance genes and other resistance genes would aggravate their spread and maintenance. In this study, although we found integrons carrying ARG cassette in pigs only from the Chinese farm, the plasmids carrying these integrons have been found around the world (Fig. S2-b/d) and mostly in pathogens. The co-occurrence between QAC resistance genes *qacF/qacEΔ1* and ARG cassettes in the integrons could help AMR maintenance in pathogens. Similarly, Gaze *et al* found that the incidence of Class I integrons was significantly higher for bacteria exposure to QACs (Gaze et al., 2005).

415

416 Enterobacteriaceae tend to have more ARGs/MRGs/MGEs-carrying DNA fragments, especially in *E.*
 417 *coli* and *Shigella flexneri*. *E. coli* is known to carry a high degree of ARGs (Li et al., 2021) and *Shigella flexneri*
 418 can cause a variety of communicable bacterial dysenteries in their hosts (Jennison and Verma, 2004). The
 419 era of plentiful antibiotics is forcing more bacteria to develop ARGs to survive and grow in the newly
 420 established toxic environment, while non-degradable metals and easily accessible biocides probably
 421 further promote the maintenance and spread of AMR in the form of co-occurrence investigated in this
 422 study, which represents an increased public health risk.

423 While this study can distinguish patterns of co-selection that align with countries and farms that are
 424 known to have large differences in antibiotics usage, it is a limitation that there are no observations on
 425 many other aspects of farming practices. Cleaning agents and frequency would be particularly pertinent
 426 given the observed co-selection of biocide resistance genes, but other differences in cultural and
 427 regulatory practices could impact both the microbiome and the resistome this can harbor. Ideally,
 428 observations of fully factorial designs could be used to separate the variance of these, but ultimately
 429 intervention designs are needed to establish practices that can minimize the horizontal spread of ARGs.
 430

431 **Conclusion**

432 In this study, we have used publicly available data to map the mobilizable resistance genes in pig gut
 433 microbiomes and exploited the patterns of co-selection by non-antibiotic factors. The genetic evidence
 434 presented here clearly suggests these factors contribute to the maintenance of AMR in pig farming. We
 435 demonstrate the commonness of this co-selection with genetic evidence and augment this with an
 436 overview of mobilizing elements. We clarify that this maintenance is not a random selection from a
 437 mobilized pool but pertains to specific phylogenetic clades. We hope this work will give further insights
 438 into the genetics of co-selection and the implications of non-pharmaceutical antibiotic agent usage in the

propagation of AMR. Specifically, this work illustrates the need for a comprehensive survey of co-selection potential for effective agronomic practices policymaking aiming for a reduction of the global AMR burden in a One Health approach.

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CRedit authorship contribution statement

Xuanji Li: Data collection, Project administration, Analysis, Writing - original draft. **Christopher Rensing:** Project administration, Writing - review & editing. **Gisle Vestergaard:** Methodology, Writing - review & editing. **Joseph Nesme:** Project administration, Writing - review & editing. **Shashank Gupta:** Project administration, Writing - review & editing. **Manimozhiyan Arumugam:** Methodology, Writing - review & editing. **Asker Daniel Brejnrod:** Data collection, Supervision, Analysis, Writing - review & editing. **Søren Johannes Sørensen:** Supervision, Writing - review & editing, Funding acquisition.

Conflict of Interest

The authors declare no competing interests.

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