Niche and local geography shape the pangenome of wastewater- and livestock-associated *Enterobacteriaceae*

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***Escherichia coli* and other *Enterobacteriaceae* are highly diverse species with 'open' pangenomes[1,2], where genes move intra- and inter-species via horizontal gene transfer[3]. These species can cause clinical infections[4,5] as well as persist environmentally[6,7]. Environmental populations have been suggested as important reservoirs of antimicrobial resistance (AMR) genes. However, as most analyses focus on clinical isolates[8,9], the pangenome dynamics of natural populations remain understudied, particularly the role of plasmids. Here, we reconstructed near-complete genomes for 828 *Enterobacteriaceae*, including 553 *Escherichia* spp. and 275 non-*Escherichia* species with 2,293 circularised plasmids in total, collected from nineteen locations (livestock farms and wastewater treatment works in the United Kingdom) within a 30km radius at three timepoints over the course of a year. We find different dynamics for the chromosomal and plasmid-borne components of the pangenome, showing that plasmids have a higher burden of both AMR genes and insertion sequences, and AMR plasmids show evidence of being under stronger selective pressure. Focusing on *E. coli*, we observe that plasmid dynamics are more strongly dominated by niche and local geography, rather than phylogeny or season. Our results highlight the diversity of the AMR reservoir in these species and niches, and the importance of local strategies for controlling the emergence and spread of AMR.**

Enterobacteriaceae can persist across diverse environmental niches[10] and also cause clinical infections, with AMR in *Enterobacteriaceae* emerging as a major problem in the last decade[11,12]. Dissemination of AMR genes often occurs via mobile genetic elements (MGEs) which can transfer intra- and inter-species, both locally[13] and globally[14]. Freshwater, wastewater and livestock-associated strains of *Enterobacteriaceae* have been proposed as reservoirs for AMR genes in clinical isolates[15–18], but the links between these remain

cryptic[19]. Current understanding of the ecology and evolution of pangenomes is incomplete[20], with ongoing debate about the roles of niche adaptation and selection[21–24]. Published *Enterobacteriaceae* genomes are biased towards clinical isolates, sampling frames reflecting truly interlinked communities are limited, and much remains unknown about the population genetics of *Enterobacteriaceae*[25] and the role of plasmids in non-clinical contexts[26].

Genomic studies of *Enterobacteriaceae* have predominantly used short-read whole genome sequencing (WGS). AMR genes and their flanking regions are frequently fragmented in short-read assemblies due to repetitive elements and structural rearrangements[13,27]. Combining short- and long-reads ('hybrid assembly') produces complete, high-quality genomes[28], allowing accurate structural resolution. Here, we used hybrid assembly of 828 sympatric *Enterobacteriaceae* (*Citrobacter*, *Enterobacter*, *Escherichia* spp., and *Klebsiella*) to characterise the pangenome of these genera considering both niche (cattle, pig, sheep, or wastewater treatment works (WwTW)-associated) and geography (sampling location).

A diverse collection of complete genomes from livestock and water-borne niches

We collected samples from nineteen locations ≥5 km apart (maximum distance: 60km) in South-central England (United Kingdom) in 2017, namely: fourteen livestock farms (four pig, five cattle, five sheep) and water sources around five WwTWs over three seasonal timepoints (Fig. 1a). We pooled samples within each location and timepoint, then cultured isolates from these pooled samples. A selected subset of 832/2098 cultured isolates underwent short- and long-read sequencing and hybrid genome assembly (Fig. 1b, see Methods), resulting in 828 high-quality genomes (Table S1: *n*=496 from livestock farms, *n*=332 from WwTWs), from four genera: *Citrobacter* (*n*=128), *Enterobacter* (*n*=71), *Escherichia* (*n*=553), and *Klebsiella* (*n*=76). Most farm isolates were *Escherichia* spp.

(451/496, 90.9%), with WwTW isolates having roughly even proportions of genera (Fig. S1). Isolates contained a median of 1 AMR gene (range: 0-23), with variation by genus: *Klebsiella* isolates carried a median of 4 (range: 1-18).

Isolates were highly diverse, containing novel diversity not present in published genomes (Fig. S2). *Escherichia* diversity included all main *E. coli* phylogroups (Fig. 1c), as well as 53 *E. fergusonii*, and 13 isolates from clades I, II, III, and V (Table S1). Phylogroup B2 was strongly associated with WwTWs compared to livestock (34.3% vs. 5.3% of *Escherichia* isolates in each category respectively; χ^2=70.4, p<0.001) particularly in influent and effluent samples (Fig. 1c). This observation is in line with Ludden et al.[29] who found that phylogroup B2 was rare in livestock-associated *E. coli* sampled in England (4/431 genomes). Pigs had a greater proportion of phylogroup A isolates (Fig. 1c). Of 187 identified *E. coli* multilocus sequence types (STs), 56.1% (105/187) were seen only once, similar to the 61% observed by Touchon et al. in a study of non-clinical *E. coli*[30]. Only 12 *Escherichia* STs were seen in both livestock and WwTW isolates, with phylogroup B1 the most represented (5/12 STs). ST10 was the most prevalent ST (n=45), seen in 10/14 farms and 3/5 WwTWs. This is in agreement with recent studies in England: Ludden et al.[29] also found ST10 to be the most prevalent *E. coli* ST in livestock-associated isolates, and AbuOun et al.[18] found the ST10 clonal complex was the most prevalent in isolates from pig farms. Our observations and good agreement with recent work in this setting suggest that our dataset, although sampled from a relatively small geographical region, is representative of broader non-clinical *E. coli* populations.

Considering only livestock *E. coli* isolates, over time, there was a persistent phylogroup signature of both livestock host and farm, with individual farm explaining slightly more variance than livestock type (R^2=28.1% vs. 25.5%, Fig. S3). In other words, between-farm

differences in *E. coli* phylogroup abundance were of the same order as differences between livestock species. However, livestock type explained less variance for STs than phylogroups (R^2=8.5%), with only 39/131 STs (29.8%) seen on more than one farm. This suggests that associations between livestock species and *E. coli* population structure are more important at the higher level of phylogroup, and clear associations between ST and livestock species are not the norm. There were only 26 instances where an *E. coli* ST was observed over time on the same farm (involving 16 STs) and the majority of these (22/26) were STs also seen across farms (Fig. S4); these could therefore represent more generalist STs.

However, STs can encompass large genomic variation. We therefore also considered *E. coli* strain clusters using a core genome distance of <100 single-nucleotide variants (SNVs) (maximal diversity observed across sampled *E. coli*: 211,251 SNVs; median pairwise distance 46,144 SNVs), there were 280 isolate pairs with <100 SNVs, of which 181 (64.6%) were isolates cultured from the same pooled sample (i.e. same farm, same timepoint) (Fig. S5a). Overall, 10.5% of all isolate pairs from the same pooled sample had <100 SNVs between them, compared to 1.4% (n=52) of isolate pairs from different timepoints on the same farm and 0.2% (n=44) between different farms of the same animal (Fig. S5b). Notably, of the latter, 41/44 were between cattle farms, and 36 involved a single cattle farm (RH06). There were only three isolate pairs with <100 SNVs between farms of different animals (Fig. S5a). Notably, all of these were between farms in close geographic proximity (two instances from pig farm RH03 and cattle farm RH10, one instance from cattle farm RH07 and sheep farm RH12; see Fig. 1a for distances), suggesting local strain movement. There were no isolate pairs with <100 SNVs between WwTW and livestock niches, and only three isolate pairs occurred across timepoints at WwTWs (all at a single WwTW).

Taken together, these results indicate that different livestock hosts have a stable balance of *E. coli* phylogroups and that each farm setting can harbour substantial strain-level diversity which, in our dataset, is unique to that farm and can persist over time. In contrast, isolates from locations proximal to WwTWs do not have a stable population of strains, reflecting the more transitory nature of this setting. We did not recover any transmission links at the strain level between WwTWs and livestock, although the vast diversity of natural *E. coli* populations means this should not be interpreted as evidence that such links do not exist.

Plasmid gene repertoires are linked to genus and niche

We recovered 2,293 circularised plasmids across all *Enterobacteriaceae*, ranging in size from 1,240-824 kbp (median: 43 kbp; Table S2). There were 298/2,293 (13.0%) with no identifiable plasmid replicon and the majority of these were from WwTW isolates (192/298, 64.4%). Multiple replicons were carried by 723/2,293 (31.5%) and these plasmids tended to be larger (median length: 106,811 bp vs. 6,275 bp for single replicon plasmids). Of *E. coli* isolates with complete genomes, over two thirds (70.4%, 245/348) carried a plasmid with an IncFII replicon. 43.0% of circularised plasmids (986/2,293) had at least one match with >99% identity to other publicly available plasmid sequences (Fig. S2b). However, 12.3% (282 of 2,293) had a top identity score of <95% to a previous known sequence (Fig. S2b), and 17 plasmids with no match were identified, suggesting novel plasmid diversity in our setting. We grouped circularised plasmids into 611 distinct plasmid clusters using alignment-free distances (see Methods), which closely matched their gene content (Fig. S6a). A recent analysis by Redondo-Salvo et al.[31], clustering 10,000 plasmids from prokaryotes using average nucleotide identity, found that plasmids within the same cluster contained a common genomic backbone. We also found that the synteny of shared genes was strongly conserved

within plasmid clusters (Fig. S6b), supporting this concept of common backbones that remain stable while allowing genes to be gained by insertion.

A median of 3.3% of genes were on plasmids (range: 0-16.5%), with substantial variation by genus and niche (Fig. S7a). This is a comparable value to the 3% finding of Touchon et al. in *E. coli* with variation over a similar range (Fig. 4a of ref. [30]). We also considered the effect of plasmid copy number (i.e. multiplying plasmid lengths by their inferred copy number) to to calculate the total DNA in both chromomes and plasmids within isolates (Fig. S7b). *E. coli* isolates had a median of 5.7% of DNA present on plasmids, which was substantially higher in pig farm isolates (median: 10.1%; Fig. S7b), linked to the presence of large plasmids. We analysed the variation of gene content with phylogeny and niche (Fig. 2). Chromosomal genes were highly genus-specific (R^2=55.0%); the plasmid-borne pangenome was far more variable, but still had a weak association with genus (R^2=6.5%) (Fig. 2). Within *E. coli*, plasmid gene content was linked to niche (R^2=5.6%) and phylogroup (R^2=5.2%), with a stronger interaction between niche and phylogroup (R^2=7.9%) (Fig. 2).

Plasmids were predicted to be conjugative, mobilizable, or non-mobilizable (see Methods), and we explored whether this affected their distribution. Non-mobilizable plasmid clusters were less commonly shared between different phylogroups within farms compared to mobilizable or conjugative plasmids (Fig. S8). Although AMR genes were predominantly found in conjugative or mobilizable plasmid clusters, consistent with their frequent acquisition and movement between strains, plasmid clusters with AMR genes were not more commonly distributed across multiple phylogroups (Chi-squared test χ^2=0.64, p=0.42; Fig. S8). However, on pig farms the majority of conjugative plasmid clusters seen across multiple phylogroups carried AMR genes, suggesting they play an important role within this niche.

Since many isolates had multiple plasmids in their genome, we also considered the possibility of plasmid-plasmid interactions. Positive epistasis between large (>10 kbp) and small plasmids has been suggested to promote plasmid stability in *Enterobacteriaceae*, based on analyses of genomes in public databases[32]. In *E. coli* isolates with complete genomes (n=348), we observed a significant association between small and large plasmid presence (Chi-squared test χ^2=4.44, p=0.035), with 45.7% carrying at least one large (>10 kbp) and one small plasmid and only 3.7% carrying a small plasmid without a large plasmid. Thus, previous results that support the existence of positive large-small plasmid epistasis are reproducible in this natural population, suggesting this is an important feature of *Enterobacteriaceae* plasmid distributions. We also found evidence for specific plasmid-plasmid associations. For example, cattle *E. coli* isolates showed co-occurrence of a ColRNA plasmid (cluster 37: median length 4.6 kbp) and an IncFII plasmid cluster (cluster 279: median length 106 kbp), with 14/16 isolates with the ColRNA plasmid also carrying the larger IncFII plasmid. Isolates were from three phylogroups (A: n=2, B1: n=5, E: n=9) and four farms, suggesting a robust association which reflects plasmid epistasis independent of chromosomal background.

Plasmids carry an over-representation of AMR genes and insertion sequences

Plasmids carried more diverse and less genus-restricted genes, as expected from their role as the more flexible component of the genome. Despite carrying just 3.3% of total gene content, plasmid-borne genes accounted for 11.5% of unique genes (8.9-17.0% considering each genus; Fig. S9) and 40.1% were seen in more than one genus (19.6-55.6% considering each genus; Table S3). Plasmids are important vehicles of AMR genes in clinical contexts; similarly, plasmids had a much greater burden of AMR genes in the niches studied here. Considering isolates with circularised chromosomes (see Methods), 901/1,876 AMR genes

(48.0%) were found on plasmids i.e. a 14.5x relative burden in plasmids. They also had a higher burden of insertion sequences (ISs), which are linked to the movement of genes by horizontal gene transfer. Of 26,565 ISs, 3,695 (21.7%) were found on plasmids (6.6x relative burden). There was a weak correlation between the number of plasmid- and chromosome-associated AMR genes within an isolate (Spearman's ρ=0.11, p=0.004) but a strong positive correlation for the number of ISs (Spearman's ρ=0.40, p<0.001) (Fig. S9a), seen across genera (Fig. S9b).

We observed different patterns of ISs across chromosomes and plasmids (Fig. S10). Some ISs were strongly associated with plasmids, the strongest association being for IS*26*. However, 27.5% of isolates carrying IS*26* on a plasmid also carried it on their chromosome, consistent with its characteristically active behaviour. It has been shown that IS26 can drive the reorganization of plasmids in clinical settings by replicative transposition[33] as well as amplify AMR genes[34] and create within-plasmid heterogeneity in a single isolate[35]. Its high prevalence in *Enterobacteriaceae* plasmids here suggests that it may play a similar role outside the context of clinical AMR. The most prevalent IS on both chromosomes and plasmids was IS*Kpn26*, with 50.2% of IS*Kpn26*-positive isolates having it both chromosomally and plasmid-borne.

Considering *Escherichia*, WwTW isolates showed a greater diversity of ISs, with 65% of ISs found in a higher proportion of WwTW isolates compared to those from farms (Fig. S11), including IS*30* which has been proposed as a marker for naturalized wastewater populations of *E. coli*[36]. Touchon et al.[30] suggest that water-borne *E. coli* strains are adapted to this niche rather than being fecal contaminants, and can therefore adapt by acquiring genetic material from not only gut bacteria (as in e.g. a livestock or human host) but also other diverse

environmental bacteria. The observation of greater IS diversity in WwTW isolates here would be consistent with this hypothesis.

We also investigated the overall patterns of co-occurrence of ISs to see if they were strongly linked on plasmids. Overall, ISs had random levels of co-occurrence on *Escherichia* plasmids (upper-tail p=0.85 from null model simulations of checkerboard score, see Methods; Fig. S12a), suggesting that ISs frequently move independently between plasmid backgrounds. In contrast, applying the same method to AMR genes, we found they significantly co-occurred (upper-tail p=0.02; Fig. S12b), suggesting co-selection and underlining the tendency of some AMR genes to co-occur in specific regions of plasmids.

Plasmids carrying AMR genes show features suggestive of selection

Plasmids fell into two broad classes across genera: small multicopy plasmids (<10 kbp, 10-100X copy number) and large low-copy plasmids (>10 kbp, <10X) (Fig. 3a). AMR plasmids were almost all large low-copy plasmids (173/184, 94.0%). While small multicopy plasmids are of interest in facilitating evolutionary innovation[37,38] this finding suggests that they do not play a major, direct role in AMR in livestock- and WwTW-associated *Enterobacteriaceae*.

Experimental evidence shows that selective advantages favour lower GC-content in intracellular elements[39], and this is what explains the tendency of plasmids to typically have lower GC-content than their host chromosome. Under this hypothesis, the more dependent on the bacterial host the plasmid is, the lower its relative GC-content should be. We investigated this in our dataset. Overall, plasmids had a lower relative GC-content than their host chromosomes (median difference -2.5%, Fig. 3b). Notably, there was an approximate gradient of relative GC-content with predicted plasmid mobility (Fig. 3b), with plasmids predicted to be mobile having a smaller relative difference. Furthermore, this difference was

less marked for AMR plasmids (median -0.3%) across mobility categories (Fig. 3b). Nearly half had a higher GC-content than their host chromosome (46.7% vs. 17.7% of non-AMR plasmids). Taken together, this suggests that AMR plasmids are being selected for in these environments, which counteracts the usual selective advantages for lower GC-content. Alternatively, this may be a signal of their relatively recent acquisition by their host.

Evidence for recent horizontal gene transfer across genera and within isolates

We identified 2,364 potential horizontal gene transfer (HGT) events involving transfers of sequence >5,000 bp between isolates of different genera (see Methods). These represent possible instances of the recent movement of genetic material across species boundaries. Isolates from the same farm were ~10x more likely to show evidence of cross-genera HGT than would be expected (Chi-squared test χ^2=1159, p<0.001; Fig. S13), and 12.3% of these cross-genera HGT events involved at least one AMR gene, with most of these AMR HGT events between pig isolates (37/48, 77.0%).

The movement of genes can also occur within individual genomes. We therefore also investigated occurrences where the same gene was present on both the chromosome and plasmid(s) within an *E. coli* genome. We observed distinct differences between niches, with increased amounts of chromosome-plasmid sharing in pig and WwTW isolates compared to cattle and sheep (Fig. S14). This may be a signature of increased selection for AMR in these niches, such that usually transitory gene movements and duplications are retained in genomes for long enough to be detected.

Quantifying the roles of phylogeny, niche and geography in the *E. coli* pangenome

To understand the strength of different factors shaping the pangenome, we analysed the pangenome of a single species, *E. coli*, in more detail. Isolates recovered from the same

location spanned total *E. coli* diversity (Fig. 4a). Inter-isolate core genome distances were strongly correlated with chromosomal gene repertoire relatedness (GRR) (Fig. 4a). Core genome distance explained the majority of variance in chromosomal GRR (Fig. 4b), but there was a consistent contribution from geography and time: isolates from the same pooled sample sharing more genes than would be expected (+1.2%), as did isolates from the same farm at different timepoints (+0.5%) (Fig. 4b). There was no such effect for isolates from different farms of the same livestock, suggesting this reflects local geography rather than adaptation to livestock host. Although the variance explained was much lower, local geography effects were also observed for plasmid GRR (Fig. 4c), but core genome distance was uncorrelated with plasmid GRR apart from for near-identical strains (Fig. 4d). Isolates from different STs from different farms of the same livstock could still have high plasmid GRR (Fig. 4e), suggesting that host-specific plasmids may facilitate niche adaptation.

Conclusions

We have investigated the pangenome of major genera of sympatric *Enterobacteriaceae* from locations within a 30km radius, using a diverse set of non-clinical isolates cultured from the same samples, and focusing in detail on *E. coli*. Despite high overall diversity, with the majority of strains only observed once in the dataset, we observed the persistence of strains and plasmids on farms over the course of a year. Our results highlight the combination of persistence and dynamism that characterises *Enterobacteriaceae* genomes at multiple scales, with relevance both for understanding the population structure of species within *Enterobacteriaceae* and for managing AMR. The existence of farm-level differences in *E. coli* populations which persist over time, with a small number of possible inter-farm transfers, suggests that livestock farms function as distinct but linked niches. It could be that "everything is everywhere" (frequent movement of strains and genes between farms) but "the

environment selects" (different farms have different selective pressures). However, the observation of persistent strains over the course of a year on farms, despite presumably varying selective conditions, and the overrepresentation of putative cross-genera HGT events in isolates at the same location suggests that geographical effects or intrinsic properties of certain bacterial/MGE lineages could affect the evolution of AMR on such timescales. Future modelling work and investigation will be required to distinguish these hypotheses. Overall, our findings underline the importance of local control strategies for the emergence and spread of AMR beyond clinical settings.

Here, we have focused on genome dynamics across different niches, but have not attempted to integrate our findings with detailed contextual information from the participating sites. This will be crucial to give further insight into how differences in antimicrobial usage on farms and other management practices can affect AMR in multi-species pangenomes, such as the decline in *mcr-1* prevalence in *Enterobacteriaceae* on pig farms after a ban on colistin in feed, observed in both China[40] and the UK[41]. Other factors beyond antimicrobial usage could include herd size, herd management practices, and cleaning and disinfection practices. Investigating these is part of our ongoing work. Similarly, differences in management between WwTWs may affect the prevalence of AMR genes in effluent, and this also the subject of ongoing work.

Resource limitations meant that we were unable to sequence and genetically evaluate all isolates that were cultured, and despite detailed sampling we will not have captured all the persistence, HGT and strain sharing events across niches: the instances of sharing we did identify remain fairly anecdotal. This is a problem faced by even intensive sampling efforts, which can only capture a tiny fraction of the population diversity and so are unlikely to retrieve close links in transmission chains[42] (of strains or MGEs). Future analyses could still

investigate the links between such sharing and possible transmission routes between both farms and WwTWs, including the land application of manure[43], the land application of sewage sludge[44], and groundwater flow[45]. Older studies have established that specific AMR genes can be more prevalent in groundwater closer to manure storage[46], but working out how to do equivalent genomic analyses for strains and MGEs across larger geographic distances with such diversity is difficult. Although this study is unprecented in evaluating four genera in such detail, AMR gene dissemination and important structural associations of AMR genes and MGEs may also be occurring within other genera not studied here. Furthermore, we did not investigate the relationship between isolates in this study and clinical human compartments in the same study area. Ludden et al. previously reported limited direct overlap for *E. coli* or *K. pneumoniae* strains from livestock and humans in a different region of England[29,47], but highlighted that 5% of human *E. coli* isolates potentially shared closely-related AMR-associated MGEs with those found in livestock. We intend to see if this finding holds in our ongoing work.

In conclusion, our study highlights the plastic and dynamic nature of AMR gene dissemination within the pangenome of major Enterobacteriaceae in several important non-clinical niches. It also demonstrates how robustly evaluating the flow of AMR genes and MGEs across highly diverse and dynamic niches is challenging even with extensive sampling. The implications of this for adequately understanding dissemination and selection of AMR genes in a 'One Health' context should not be under-estimated.

Figures

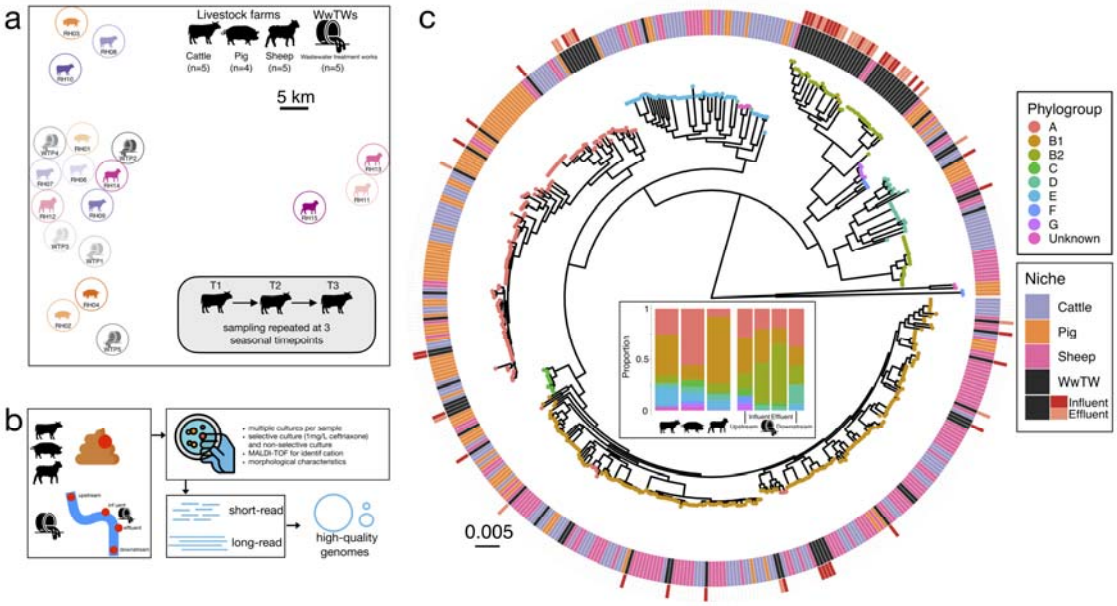

Figure 1. Overview of the diverse *Escherichia coli* isolates in this study. **(a)** Relative sampling locations of the farms (cattle, pig, sheep) and wastewater treamtent plants (WwTWs) in this study, sampled at three different timepoints. **(b)** Schematic illustration of the sampling, culture and sequencing workflow, resulting in high-quality genome assemblies with a median of 1 circularised chromosome and 2 circularised plasmids per assembly. **(c)** Mid-point rooted core genome phylogeny of *E. coli* isolates (*n*=488), with tips coloured by phylogroup and ring colours showing sampling niche. Inset panel at centre of phylogeny shows phylogroup abundances (as proportion of isolates) from different sampling niches.

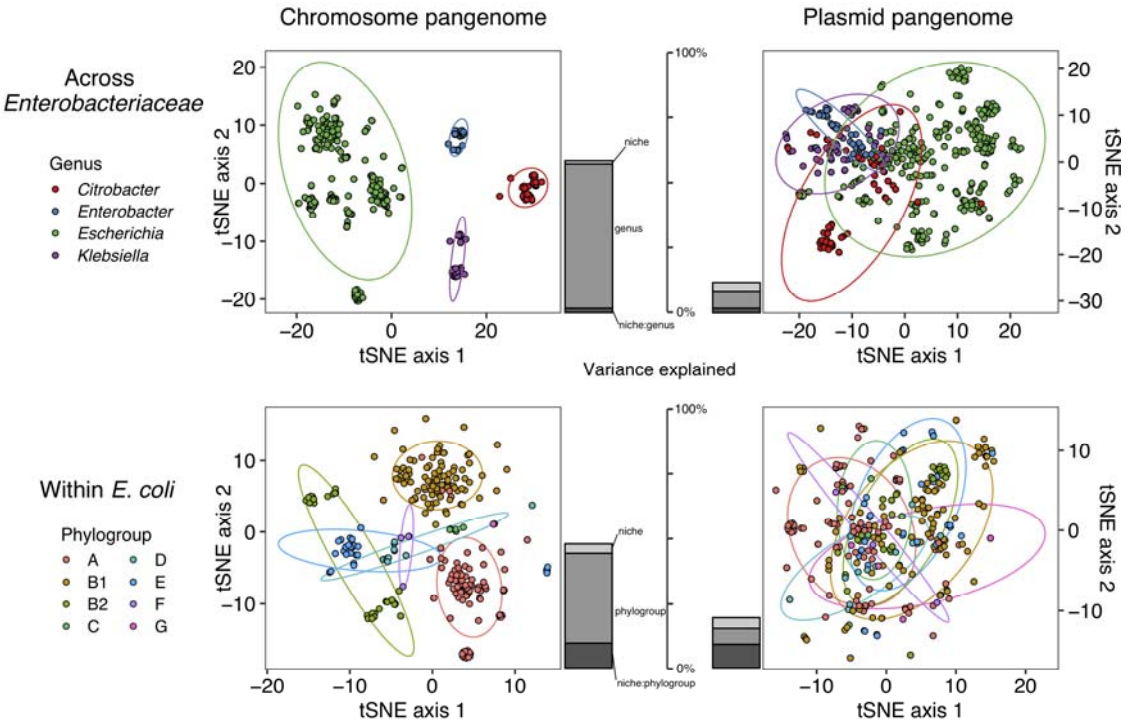

Figure 2. The plasmid-borne component of the pangenome is structured by niche and phylogeny, with greater variation than in the chromosomal component. Plots are shown for all isolates in four genera across *Enterobacteriaceae* (top row) and for *E. coli* (bottom row), for both the chromosomal component of the pangenome (left column) and the plasmid-borne component (right column). Stacked bar charts show the variance in gene content explained by niche, phylogeny (genus or phylogroup) and their interaction. The plasmid-borne component has greater residual variance than the chromosomal component, with a comparatively stronger niche-phylogeny interaction (darkest shaded bar).

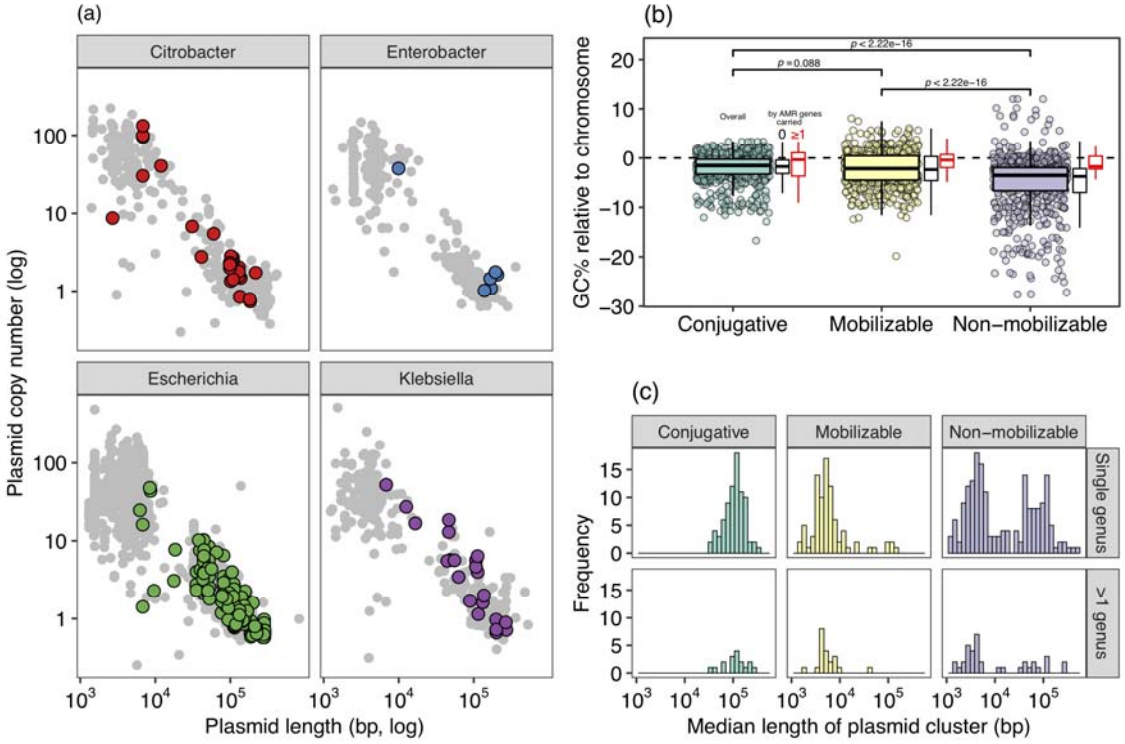

Figure 3. Distinct plasmid lifestyles between AMR and non-AMR plasmids. (a) Plasmid length (x-axis) and inferred copy number (y-axis) of all circularised plasmids (*n*=2,293), faceted by genus. Plasmids with ≥1 AMR gene (coloured points) tended to be larger and present in lower copy numbers. **(b)** Relative GC-content of all plasmids to their host chromosome for all circularised plasmids present in an assembly with a circularised chromosome (*n*=1,753 plasmids across 616 isolates), split by predicted plasmid mobility. Boxplots are additionally shown classifying plasmids within predicted mobility types by the number of AMR genes carried: those ≥1 AMR gene (red) or no AMR genes (black). Comparisons with *p*-values are shown for all plasmids within a predicted mobility class. **(c)** Length distributions of plasmid clusters (see Methods).

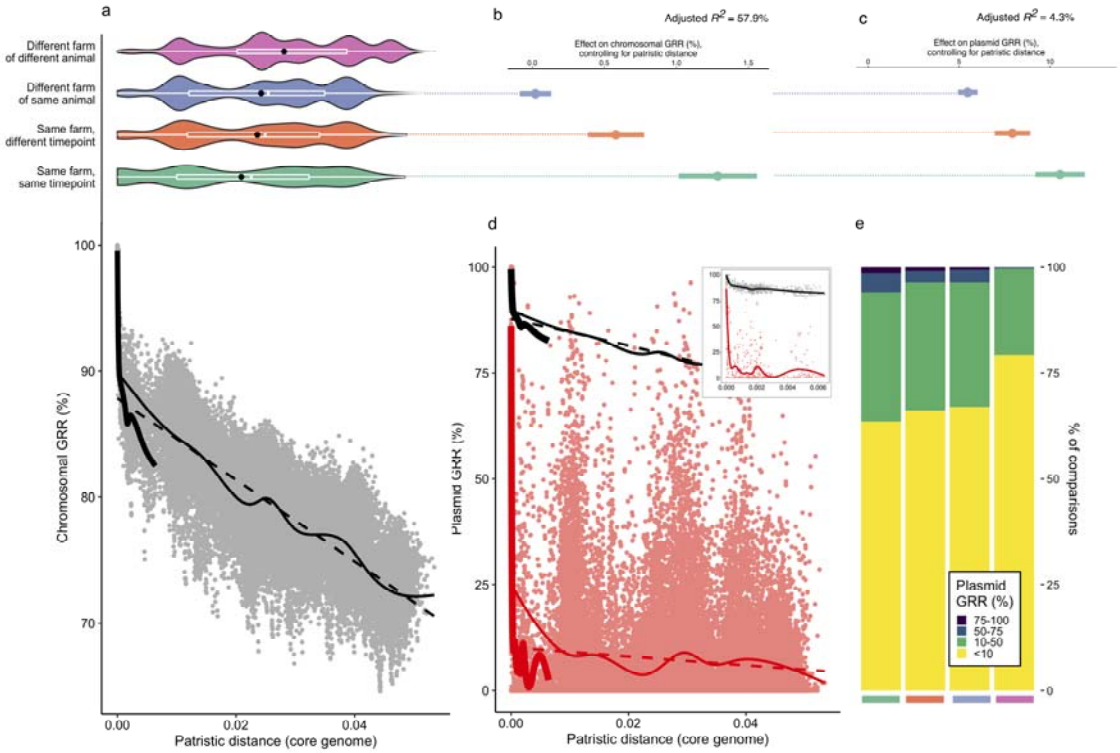

Figure 4. The interplay of phylogeny and niche in the *E. coli* pangenome. **(a)** Pairwise comparisons of gene repertoire relatedness (GRR) for chromosomal genes show that chromosomal GRR falls off rapidly at small patristic distances, followed by an approximately linear decrease. Fits show intra-ST comparisons (thick black line), all comparisons (thin black line), and a linear model (dashed black line). Violin plots above show the distribution of patristic distances depending on the relative sample source of the two isolates in the pairwise comparison (white boxplot: median and IQR; black point: mean), showing that even isolates cultured from the same sample (same farm, same timepoint) span equivalent diversity to isolates cultured from different locations. **(b)** Coefficients from a linear model for chromosomal GRR with an interaction term with patristic distance (excluding intra-ST comparisons). **(c)** Variance explained by phylogeny and geography for chromosomal and plasmid GRR. **(d)** GRR for plasmid-borne genes with patristic distance. Fits show intra-ST comparisons (thick red line), all comparisons (thin red line), and a linear model (dashed red line). Inset panel shows left-hand region of the plot with only intra-ST comparisons, with chromosomal GRR relationship also shown (grey points, black line). **(e)** Plasmid GRR comparisons shown by isolate sources, excluding intra-ST comparisons. Colours on x-axis are the same as in (a). Plots include all *E. coli* isolates with a circularised chromosome (*n*=363).

Methods

Isolates were sequenced from samples collected as part of the REHAB project in 2017, which aimed to characterise non-clinical *Enterobacteriaceae* populations in four different niches within a defined study area of South-central England: cattle farms, pig farms, sheep farms, and water environments linked to wastewater treatment works (WwTWs). Sampling occurred at each location at three separate timepoints (TPs): January-April 2017 (TP1), June-July 2017 (TP2) , October-November 2017 (TP3).

Farms. Five cattle farms, five sheep farms and four pig farms were recruited from the study area following a defined recruitment process (described in more detail in MA et al.; paper in preparation). Briefly, we aimed to recruit the five largest farms for each livestock type within the area using local APHA databases, progressively inviting the next largest farm if a farm declined. All participating farmers provided written consent for farm sampling for research purposes and farm samples were taken between January and November 2017 on three separate visits ('timepoints') for each farm. Each farm was divided in five or fewer 'epidemiological groups', defined as a group of animals expected to share similar characteristics and managed in the same way. Ten pooled samples were collected from each of these groups, with each sample composed of small pinches of fresh faeces from the floor combined into a small composite sample around 5cm in diameter. Each group's ten samples were pooled, diluted up to 10^{-5} in phosphate buffer solution (PBS) (pH 7.2) and plated on to CHROMagar™ ECC (CHROMagar Microbiology, Paris, France) and CHROMagar™ ECC plates containing 1mg/L cefotaxime as a marker for multi-drug resistance. Up to ten colonies were collected from 1mg/L cefotaxime-supplemented plates and fourteen colonies from CHROMagar™ ECC plates; where ten colonies were not recovered, additional colonies were taken from the CHROMagar™ ECC plates, resulting in 24 isolates per farm. Pure isolate sub-cultures were subsequently stored at -80˚C in MicroBank beads (Pro-Lab Diagnostics, Neston, Cheshire, UK), and the bacterial species confirmed using MALDI-TOF (Bruker, Coventry, UK) or 16S rRNA sequencing[48]. The median number of sequenced isolates for a farm-timepoint combination was twelve (range: 9-14), with 496 farm isolates in total: cattle (n=178), pig (n=144), sheep (n=174).

Wastewater treatment works (WwTWs). Five WwTWs were selected based on a defined recruitment process (described in more detail in DSR et al.; paper in preparation) including; geographic location within the study area, wastewater treatment configuration, wastewater population equivalent served, consented flow, and the accessibility of the effluent receiving river for sampling both upstream and downstream. Sampling took place in 2017 over three sampling rounds: February–March (TP1), June–July (TP2), and October–November (TP3). Sewage influent samples were collected after WwTW coarse screens and effluent samples were collected at the last sampling point before entering the river. For each sampling round, ~6 repeated 200 ml samples of influent and effluent were collected between 9 am and 12 pm, using a sampling pole and sterile Whirl-Pak collection bags. Repeat samples in each round were pooled prior to processing, to reduce the impact of temporal variability in wastewater flows and composition. Sediment samples were collected from 100 m upstream and 250 m downstream of the effluent entry point into the river. Sediment samples were collected using a custom sampling pole that held a removable 50 ml plastic centrifuge tube (Sigma, UK). Using a fresh sterile 50 ml tube each time, sediment from the riverbed was collected from the surface layer at three points at each sampling site; near bank, the centre of the river, and the far bank. These samples were pooled prior to analysis to account for spatial variability in sediment composition. Influent, effluent and sediment samples were stored in an insulated box at ~4 °C until

getting back to the laboratory (<6 h). Influent, effluent, 100 m upstream and 250 m downstream environmental samples collected from each sewage treatment works were transferred to the laboratory on ice and processed within 24 hours of collection. Each sample was vortexed briefly, serial diluted to 10^{-3} in nutrient broth containing 10% glycerol (Oxoid, Basingstoke, UK) and plated on to CHROMagar™ Orientation agar (Chromagar, Paris, France) and CHROMagar™ Orientation agar supplemented with 1 µg/ml cefotaxime (Cambridge Biosciences, Cambridge, UK). Colonies with putative morphology for species of interest were subcultured from dilution plates with suitably isolated growth. A total of up to 20 colonies was picked per sample: up to ten colonies were picked from the 1mg/L cefotaxime-supplemented plates and the remainder picked from the non-supplemented plates. Pure isolates subcultured on Columbia blood agar (CBA) (Oxoid, Basingstoke, UK) were subsequently stored at -80˚C in nutrient broth containing 10% glycerol, and bacterial species confirmed using MALDI-TOF (Bruker, Coventry, UK).

DNA sequencing. A subset of isolates were selected for sequencing to represent diversity within the four major genera within each niche, including the use of third-generation cephalosporin resistance as a selective marker to identify a sub-group of multi-drug resistant isolates within each genus. 832 isolates were each sequenced with both a short-read (Illumina HiSeq 4000) and a long-read sequencing approach (four isolates selected for sequencing failed subsequent hybrid assembly and were not included in further analyses). For the first timepoint, the latter involved sequencing using either PacBio SMRT (*n*=192) or Oxford Nanopore Technologies (ONT) methodologies (*n*=127). The results of a pilot study comparing sequencing and assembly approaches using a subset of REHAB isolates[28] were used to inform the choice of ONT as the long-read sequencing approach for all isolates from the second (*n*=255) and third (*n*=254) timepoints.

Isolate stocks from -80˚C storage were cultured on to CBA and supplemented with cefpodoxime (Fisher Scientific, USA) 10 µg discs for isolates not sensitive to cefotaxime during original sample isolation. DNA was extracted using the Qiagen Genomic tip/100G (Qiagen, Venlo, Netherlands) according to the manufacturer's instructions. DNA concentration was quantified by Qubit® 2.0 fluorimeter (Invitrogen, UK), and quality and fragment size distribution assessed by TapeStation 2200 (Agilent, Santa Clara, USA). ONT sequencing libraries were prepared by multiplexing 6-8 DNA extracts per flow cell using kits SQK-RBK004, SQK-LSK108 and EXP-NBD103 according to the manufacturer's protocol. Libraries were loaded onto flow cell versions FLO-MIN106 R9.4(.1) SpotON and sequenced for 48 h on a GridION (ONT, Oxford, UK).

Genome assembly. We used the hybrid assembly and sequencing methods described in our pilot study[28] to produce high-quality *Enterobacteriaceae* genomes from short and long reads. In brief, we used Unicycler (v0.4.7)[49] with 'normal' mode, --min_component_size 500, --min_dead_end_size 500, and otherwise default parameters. Our pilot study[28] explored the accuracy of assemblies from this method (and others) in detail using a multiple metrics, including: the mapping of Illumina short reads back to assemblies, the mapping of long reads back to assemblies, the comparison of assemblies from the same isolate, using known marker genes to assess overall quality, the assembly's overall 'circularity', and the presence of indel errors. Notably, we found excellent concordance in structural agreement of circular structures from different assemblies (see e.g. Fig. S4 of ref.[28]), giving us confidence that circular structures from hybrid assemblies (chromosomes and plasmids) are accurate. One possible concern could be that chimeric reads in long-read datasets could lead to incorrect structures in the final hybrid assembly. However, an independent investigation[50,51] using data from our pilot study alongside simulated data showed that long-read-only assemblies with Unicycler are only

affected by the presence of chimeric reads at rates of ~15% of reads, and are thus not a concern for real datasets which typically have rates of <2%.

Final assemblies from all isolates had a median of four contigs (IQR: 3-8, range: 1-391), with a median of two circularised plasmids (IQR: 1-4, range: 0-14). The majority (616/828, 74.4%) of assemblies had a circularised chromosome, and 558/828 (67.3%) were complete i.e. chromosome and all plasmids circularised (Table S1).

Genome assignment and typing. We assigned species and sequence type (ST) from assembled genomes using mlst v2.16.4[52]. We also validated species assignments by downloading all NCBI Refseq complete genomes for the four genera under study as of June 4 2020 and using fastANI (v1.3)[53] to compute average nucleotide identity scores against reference genomes for each assembled genome. We took the species assignment of the top hit for each assembled genome. Furthermore, we manually checked genus assignments using a tSNE plot of isolate genomes against a collection of reference genomes (not shown) and made corrections to the assignment if necessary. We used ClermonTyping (v1.4.1)[54] to assign phylogroup to *n*=553 *Escherichia* isolates. Considering the genus *Escherichia*, there were 553 isolates, 410 with circularised chromosomes, and of these 379 were complete genomes containing 961 complete plasmids in total. Considering only *E. coli*, there were 502 *E. coli* isolates, 372 with circularised chromosomes, and of these 348 were complete genomes containing 878 complete plasmids in total. A minority of genomes were *E. fergusonii* (n=51), from clades I-V (*n*=14), or could not be typed (*n*=7), with *n*=481 genomes from within the principal *E. coli* phylogroups (A: *n*=131, B1: *n*=193, B2: *n*=59, C: *n*=11, D: *n*=25, E: *n*=50, F: *n*=6, G: *n*=6).

Sequenced isolates from three other *Enterobacteriaceae* genera included: *Citrobacter* (*n*=128: 82 *C. freundii* and 46 unassigned *Citrobacter sp.*), *Enterobacter* (*n*=71: 59 *E. cloacae* and 12 unassigned *Enterobacter sp.*)*;* and *Klebsiella* (*n*=76: 40 *K. pneumoniae*, 30 *K. oxytoca*, 2 *K. aerogenes*, and 4 unassigned *Klebsiella sp*.). The majority of farm-associated isolates were *E. coli*, whereas WwTW-associated isolates had roughly equal numbers of genera (Fig. S1). This reflects both the diversity present in each niche and the selection strategy to sequence equal numbers across genera where feasible.

Pangenome analysis. All genomes were annotated with Prokka (v1.14.0)[55]. We performed a multi-species pangenome analysis by clustering genes into gene groups using Roary (v3.12.0)[56] across all isolates at various sequence identity thresholds with the maximum number of clusters set to 300,000 (-g 300000) and without splitting paralogs (-s). At a 95% identity for blastp, there were 139,788 gene groups across all genera. Further to this analysis, genes were also clustered at a higher sequence identity (>99% identity threshold) in order to identify recent HGT events, which gave 214,743 gene groups across all genera. For *n*=616 isolates with circularised chromosomes, we split the genome into chromosomal and plasmid-borne components (i.e. all other contigs) to analyse the genomic location of genes. We excluded isolates without circularised chromosomes from this analysis. For within-species pangenome analyses, the more recently developed Panaroo gives lower annotation error rates and a more accurate core genome than Roary or other methods by using gene adjacency (i.e. synteny) information[57], although it is not suitable for cross-species analyses where no core genome is expected. Therefore, for a higher-resolution within-species analysis of *n*=488 E. *coli* isolates (excluding *E. fergusonii* and clades I-V), we used Panaroo (v0.1.0)[57] to extract a core genome alignment based on 2,915 concatenated core genes (Fig. 1c). The phylogeny was produced using iqtree (v1.6.11)[58], with branch lengths not corrected for recombination, and plotted with ggtree (v2.0.1)[59].

Plasmid annotation and clustering. We searched all plasmids against PLSDB (version: 2020-03-04)[60] which contains 20,668 complete published plasmids, using 'screen' in mash (v2.0)[61] and keeping the top hit. All plasmids had a match apart from 17 small plasmids predicted to be non-mobilizable (median length 4.8 kbp, range 2.9-20.7 kbp), from *Escherichia* (n=11), *Enterobacter* (n=2) and *Citrobacter* (n=4). We clustered plasmids using mob cluster and assigned replicon types with mob typer, both part of the MOB suite[62]. Mob cluster uses single linkage clustering with a cutoff of a mash distance of 0.05 (corresponding to 95% ANI), resulting in 611 clusters (Table S2). In total, there were 134 different combinations of replicons observed on plasmids ('replicon haplotypes'). The most abundant replicon was IncFIB (n=460) which was seen across all niches (pig [n=81], cattle [n=113], sheep [n=78], and WwTWs [n=188]). Only nine small multicopy plasmids (~6 kbp) carried AMR genes, all of which had a ColRNAI replicon; such ColRNAI plasmids have been proposed to be sources of evolutionary innovation[37,38]

We considered the relationship between such 'distance-free' clustering and plasmid gene content. Based on gene clustering with Roary (see above), we compared the structure of circularised plasmids using all connecting edges between two genes. We defined the resemblance for both gene content (gene presence/absence) and gene structure. The gene content resemblance between two plasmids with n_1 and n_2 genes respectively, with N genes in common, was defined as $r_{content}=2N/(n_1+n_2)$. The edge structure resemblance between two plasmids with g gene-gene edges in common, was defined as $r_{edge}=2g/(n_1+n_2)$. Typically $r_{edge}<r_{content}$ but this definition does allow for the case where repeated genetic elements produce $r_{edge}>r_{content}$ (e.g. Fig. S6b).

Comparison of plasmid-borne and chromosomal pangenome components. To visualize cross-genera pangenomes (e.g. Fig. 2), we used t-distributed Stochastic Neighbor Embedding (t-SNE). We used the Rtsne function with a perplexity of 30 on gene presence/absence matrices in the Rtsne R package (v0.15)[63]. To conduct permutational analyses of variance, we used the adonis function from the vegan R package (v2.5-6)[64] on the matrix of pairwise Jaccard distances, which was calculated using the vegdist function. For between-genera analyses, we used the formula *dist~niche*genus*. For within-*Escherichia* analyses, we used the formula *dist~niche*phylogroup*.

Detection of antimicrobial resistance genes and insertion sequences. We searched assemblies using ABRicate (v0.9.8)[65] for acquired resistance genes (i.e. excluding mutational resistance) in the the NCBI AMRFinder Plus database (PRJNA313047). We used a minimum identity threshold of 90% and a minimum coverage threshold of 90% (Table S4). Isolates cultured selectively from cefotaxime-supplemented plates carried more AMR genes than non-selectively cultured isolates (median of 7.5 vs. 1.0), as expected. We also searched for insertion sequences (ISs) using the ISFinder database[66] as a database in ABRicate with the same identity and coverage thresholds (Table S5).

Detection of recent horizontal gene transfer events. We performed an all-against-all comparison of assemblies with mummer (v3.23-2)[67] using the -maxmatch option to identify shared sequences of length >5,000 bp between genomes of different genera (these could include both transfer of whole plasmids or partial sequences). For comparing the observed distribution of cross-genera HGT events to the expected, we assumed a random distribution drawn from all possible cross-genera comparisons from livestock isolates.

Distribution of insertion sequences. We constructed the bipartite presence/absence network of ISs and replicon haplotypes for the 34 replicon haplotypes which were observed on 10 or more plasmids. We simulated null models of co-occurrence patterns using the cooc_null_model with null model sim9,

which fixes the row and column sums of the presence/absence matrix, in the R package EcoSimR (v0.1.0)[68]. Simulations used n=10,000 iterations with a burn-in of 500 iterations.

Modelling of gene repertoire relatedness (GRR). We selected a subset of *E. coli* genomes with a circularised chromosome (*n*=363) and used the core genome tree constructed with iqtree (Fig. 1, dropping other *E. coli* isolates) to calculate patristic distances between isolates. We calculated chromosomal and plasmid GRR for all pairwise comparisons using output from roary (95% identity threshold, as above) and fit linear models for GRR (Fig. 4).

Data availability. Sequencing data and assemblies for all isolates are available on NCBI under BioProject accession PRJNA605147. Biosample accessions for all isolates are provided in Table S1.

Image credits. The following images are used in Figure 1: Petri dish icon made by monkik; pig, cow, sheep and faeces icons made by Freepik; WwTW symbol made by Smashicons (all sourced from flaticon.com).

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Author contributions

Using the CRediT system, author contributions were as follows: conceptualization (LPS, MJB, DWC, DSR, MFA, ASW, NS), methodology (LPS, ASW, NS), software (LPS, JS), validation (LPS, KKC, JK, MA, ES, LB, GR, ATMH, HP, RS, NS), formal analysis (LPS), investigation (KKC, JK, MA, ES, HSG, LB, GR, MJBo, ATMH, HP, JS, DG, RPS, RS, NS), resources (all authors), data curation (LPS, KKC, JK, MA, ES, LB, GR, ATMH, HP, JS, DG, RPS, RS, NS), writing - original draft (LPS, ASW, NS), writing - review & editing (all authors), visualization LPS), supervision (TEAP, MJBa, DWC, DSR, MFA, ASW, NS), project administration (MA, RPS, SJH, DSR, MFA, ASW, NS), funding acquisition (MJB, DWC, DSR, MFA, NS).

Competing interest declarations

The authors declare no competing interests.

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