Supporting information

Supplementary Tables

Table S1: Metagenome datasets used in this study for the populationlevel analyses of the gut antibiotic resistome. The table gives the number of adult stool metagenome samples from each country that were used in the analyses of country-level correlation to antibiotic consumption rate and interpersonal variation of gut resistome. $\frac{1}{1}$ Health status was determined from the description of the original study subject recruitment criteria, and on the sample's categorization into disease-control grouping whenever the original study was presented in a diseased-versus-healthy comparison scheme. The 'disease' label here encompasses several different unhealthy states including CDI, cholera, colorectal adenoma, CRC, IBD, liver cirrhosis, T2D, fatty liver, hypertension, rheumatoid arthritis, and STEC infection. See Figure S11 for full disease acronym definitions. ² We labelled the sample with 'No' for 'Antibiotic Exposure' 'if the original metadata description explicitly indicated that the given sample's donor was not taking any form of antibiotic at the time of sample collection, or if the original study's recruitment criteria included the exclusion of subjects that had been taking antibiotics within a defined period preceding the enrollment. T Populations with traditional lifestyle, N Noncontemporary samples.

Table S2: Clustering of ARG catalogue. All ARG ORFs collected from the metagenome assemblies of the human microbiome dataset and RefSeq genome assemblies were pooled and clustered by average nucleotide identity. Clustering was performed with the cascaded clustering workflow using a greedy set cover algorithm, according to the default settings of the cluster command in the program mmseqs2 at four different nucleotide identity and coverage cut-offs (see Methods).

Table S3: Country level comparison of the abundance and diversity of ARGs in healthy individuals. Stool metagenomes sampled from healthy adults who were not taking antibiotics were used to calculate the country-based population level statistics on the abundance and diversity of ARGs. Only countries with at least 10 samples that met the criteria were analyzed. We used readbased profiling to derive median total abundance of ARGs (copies per genome cpg) for each country. Diversity of ARGs was calculated by randomly picking subsamples from each country until the total raw reads in the picked samples amount to 100 ± 10 Gbp and counting the number of ARG cluster 95 discovered in the subsamples. The subsampling was repeated 99 times. Countries that had a total of less than 100 Gbp of raw reads from the eligible individuals according to the above criteria were excluded from this analysis. The resulting richness estimates were scaled to the number of clusters per 100 Gbp to adjust for slight variations in the actual size of subsampled datasets. The median, the first quartile $(Q1)$, and the third quartile $(Q3)$ values are presented for each parameter.

AMU		Resistome parameter	Countries tested Shapiro-Wilks			Kendall			Pearson		
data source	Value type	Scope of ARGs and/or gene diversity unit	Number Outlier		\mathbf{p}	tau	p	FDR	r	p	FDR
CDDEP	Abundance from reads	Total	12	CHN	6.7E-2	0.63	5.7E-3 *	$1.9E-2$ *	0.89	$2.3E-4$ *	$8.3E - 3$ *
	Abundance based on assembled catalog	Total	12	CHN	$6.8E-1$	0.49	4.1 $E-2$ *	$7.4E - 2$	0.80	$2.9E-3$ *	$1.7E-2$ *
		Plasmid ORFs			2.5E-1	0.64	5.7E-3 *	$1.9E-2$ *	0.84	$1.1E-3$ *	$1.3E-2*$
		Non-plasmid ORFs			$1.0E + 0$	0.53	$2.6E - 2$ *	$5.2E - 2$	0.77	$5.9E-3$ *	$1.9E-2$ *
		Multi-species clusters			$6.0E-1$	0.60	$9.9E - 3*$	$2.5E-2$ *	0.83		$1.6E-3$ * $1.4E-2$ *
		Single-species clusters			$3.5E - 2$ *	0.26	$2.9E-1$	$3.IE-1$	0.45	$1.7E-1$	$2.0E-1$
		LCA-unassigned clusters			$2.9E-4*$	0.30	$2.2E-1$	$2.5E-1$	0.09	8.0E-1	8.0E-1
WHO	Abundance from reads	Total	10	None	$7.0E - 2$	0.60	$1.7E-3$ *	$9.3E-3*$	0.65	$4.0E-2$ *	$7.6E - 2$
	Abundance based on assembled catalog	Total	10	None	$4.8E-1$	0.73	$2.2E-3$ *	9.9E-3 *	0.82	$3.9E-3$ *	$1.8E-2$ *
		Plasmid ORFs			$1.8E-1$	0.69	4.7E-3 *	$1.9E-2$ *	0.84	$2.3E-3$ *	$1.7E-2$ *
		Non-plasmid ORFs			$4.6E-1$	0.64	$9.1E-3*$	$2.5E-2$ *	0.71	$2.2E-2$ *	$5.0E-2$ *
		Multi-species clusters			$2.1E-1$	0.64	$9.1E-3*$	$2.5E-2$ *	0.75		$1.2E-2$ * 3.1E-2 *
		Single-species clusters			8.2E-2	0.45	7.8E-2	$1.3E-1$	0.40	2.5E-1	$2.6E-1$
		LCA-unassigned clusters			$1.5E-4*$	0.31	$2.3E-1$	2.5E-1	0.10	$7.8E-1$	$8.0E-1$
	CDDEP Richness	ARG cluster90, total	10	CHN	$6.5E-1$	0.42	$1.2E-1$	$1.7E-1$	0.57	$1.1E-1$	$1.4E-1$
		ARG cluster95, total			$6.6E-1$	0.39	$1.8E-1$	$2.1E-1$	0.57	$1.1E-1$	$1.4E-1$
		ARG cluster99, total			$4.8E-1$	0.39	$1.8E-1$	$2.IE-1$	0.63	$6.7E - 2$	$1.0E-1$
		ARG cluster100, total			$4.3E-1$	0.39	$1.8E-1$	$2.1E-1$	0.62	$7.7E - 2$	$1.1E-1$
WHO	Richness	ARG cluster90, total	8	None	$7.4E-1$	0.86	$1.7E-3*$	9.3E-3 *	0.85	$7.3E - 3*$	$2.0E-2$ *
		ARG cluster95, total			$8.4E-1$	0.86	$1.7E-3$ *	$9.3E-3*$	0.86	$6.3E-3*$	$1.9E-2$ *
		ARG cluster99, total			$3.7E-1$	0.86	$1.7E-3$ *	9.3E-3 *	0.87	5.5E-3 *	$1.9E-2$ *
		ARG cluster100, total			$6.7E-1$	0.71	$1.4E - 2$ *	$3.2E-2$ *	0.80	$1.8E-2*$	$4.3E - 2$ *
	CDDEP Richness	ARG cluster99, multi-species	10	CHN	$4.0E-1$	0.44	$1.2E-1$	$1.7E-1$	0.63	$6.8E - 2$	$1.0E-1$
		ARG cluster99, single-species			$4.0E-1$	0.39	$1.8E-1$	$2.1E-1$	0.59	9.1E-2	$1.3E-1$
		ARG cluster99, LCA-unassigned			5.5E-1	0.20	$4.6E-1$	$4.6E-1$	0.47	$2.0E-1$	$2.3E-1$
WHO	Richness	ARG cluster99, multi-species	$\overline{\mathbf{8}}$	None	$6.6E-1$	0.71	$1.4E - 2$ *	$3.2E - 2$ *	0.69	$6.1E-2$	$1.0E-1$
		ARG_cluster99, single-species			8.0E-1	0.86	$1.7E-3*$	9.3E-3*	0.93	$9.5E-4*$	$1.3E - 2$ *
		ARG cluster99, LCA-unassigned			$6.2E-1$	0.47	$1.1E-1$	$1.7E-1$	0.46	$2.5E-1$	$2.6E-1$
	CDDEP Richness	ARG cluster95, from plasmid	10	CHN	$4.5E - 2$ *	0.25	$3.5E-1$	$3.6E-1$	0.70	$3.7E - 2$ *	$7.4E - 2$
		ARG cluster99, from plasmid			$6.7E - 2$ *	0,40	$1.4E-1$	$1.9E-1$	0.65	5.6E-2	$1.0E-1$
		ARG cluster95, from non-plasmid			$4.3E-1$	0.42	$1.2E-1$	$1.7E-1$	0.52	$1.5E-1$	$1.8E-1$
		ARG cluster99, from non-plasmid			5.4E-1	0.44	$1.2E-1$	$1.7E-1$	0.63	6.8E-2	$1.0E-1$
WHO	Richness	ARG cluster95, from plasmid	8	None	1.5E-1	0.69	$1.8E - 2$ *	$3.8E-2$ *	0.74	$3.5E-2$ *	$7.4E - 2$
		ARG cluster99, from plasmid			$4.6E-1$	0.62	$3.4E - 2$ *	6.4E-2	0.61	$1.1E-1$	$1.4E-1$
		ARG cluster95, from non-plasmid			2.9E-1	0.91	$1.8E - 3$ *	9.3E-3*	0.87	5.3E-3*	$1.9E-2*$
		ARG cluster99, from non-plasmid			$6.7E-1$	0.93	$4.0E - 4*$	$9.3E-3*$	0.88		$3.7E-3*1.8E-2*$

Table S4: Country level correlation between total antibiotic consumption rate and gut resistome prevalence and richness. Country median gut resistome profiles were derived from the subjects that were identifiable as healthy adults that were currently unexposed to antibiotics at the time of sampling. Median relative abundance of ARGs was calculated only for the countries that had at least 10 samples that met the criteria. ARG richness was calculated only for the countries where at least 100 Gbp total sequencing reads were collected from the subjects that met the criteria. The resulting richness estimates were scaled to the number of clusters per 100 Gbp to adjust for slight variations in the actual size of subsampled datasets. Total antibiotic consumption rates in the units of - defined daily does - DDD per 1,000 capita per year was used. P-values and Benjamini-Hochberg-adjusted FDR values below 0.05 are highlighted (*).

Table S5: Country level correlation between antibiotic consumption rate in each antibiotic class versus median abundance and rarefied richness of the corresponding ARGs in the gut metagenomes. Six broad antibiotic classes that were covered by both WHO and CDDEP statistics were used in the analyses. Abundance of ARG families in each sample were summarized at the level of antibiotic classes based on the annotation given in the CARD database. Only stool samples from subjects that were identifiable as healthy adults that were currently unexposed to antibiotics at the time of sampling were included in the analysis. Median abundance of ARGs in the countries that have at least 10 samples were tested for correlation with the defined daily dose (DDD per 1,000 per year) statistics from WHO or CDDEP. Richness of ARG cluster99s were calculated for each country for the ARG cluster99s affiliated to each target antibiotic class. From each country, random selections of samples was performed to reach 100 ± 5 Gbp of total sequence reads with 100 iterations, and the number of ARG cluster99s in the resulting subsamples were scaled to the number of clusters per 100 Gbp to adjust for slight variations in the actual size of subsampled datasets. Median number of clusters per 100 Gbp derived from 100 iterations represented the richness at country level. P-values or Benjamini-Hochberg-adjusted FDR values below 0.05 were highlighted (∗).

Table S6: Proportion of multi-species ARG cluster99s charted by the antibiotic class the genes confer resistance to or the resistance mechanism. Only the ARG cluster99s that contain one or more ORFs from adult gut metagenomes $(n = 6,104)$ were counted in this table. Classification of the ARG clusters into the categories (antibiotic class and resistance mechanism) was made based on the annotation of the representative ORF of the cluster. The annotations of the ARG ORFs were derived from the CARD database (see Methods).

Table S7: Resistance classes and species enriched in the resistotypes (A) Normalized abundance of ARGs (cpg) was summed by the antibiotic class and compared between the two resistotypes. Stool metagenomes assigned to one of the two resistotypes $(n = 5,372)$ were used. A pseudo-abundance value of 2.5e-5 cpg, the smallest non-zero value in the matrix of ARG relative abundances, was added before log transformation. Resistance classes with Benjamini-Hochberg-adjusted FDR value < 0.05 following two-sided Mann-Whitney tests were included in the table. (B) Compositional abundance of species-level genome bins (SGBs) was compared between the two resistotypes. A pseudo-abundance value of 1.02E-9 was added before the log transformation. Stool metagenomes assigned to resistotypes were used except for the samples that were not profiled at species level \langle < 1,000 ORFs recovered for single-copy core COGs). SGBs were labelled as pathogen or not based on the 463 pathogen names identified from the reference (see Methods). Detection frequency of each SGB among the healthy adults who were not taking antibiotics $(n = 3,096)$ was given as 'preval. in the healthy'.

Enterotype	SGB	Fold	FDR	Species	Genus	Family	Phylum
	4953	5.74	1.6E-195	Roseburia sp. CAG 182	Roseburia	Lachnospiraceae	Firmicutes
	4289	5.69	1.5E-110	Eubacterium sp. OM08 24	Eubacterium	Eubacteriaceae	Firmicutes
	4670	5.68	3.3E-219	Coprococcus catus	Coprococcus	Lachnospiraceae	Firmicutes
	4328	5.68	3.4E-66	Firmicutes bacterium CAG 341	Firmicutes unc.	Firmicutes unc.	Firmicutes
1	714	5.65	1.9E-112	Methanobrevibacter smithii	Methanobrevibacter	Methanobacteriaceae Euryarchaeota	
"Lachnospiraceae"	4809	5.65	2.4E-197	Blautia SGB4809	Blautia	Lachnospiraceae	Firmicutes
	14625	5.56	1.1E-148	Collinsella SGB14625	Collinsella	Coriobacteriaceae	Actinobacteria
	6778	5.56	1.7E-115	Catenibacterium mitsuokai	Catenibacterium	Erysipelotrichaceae	Firmicutes
	14624	5.52	4.2E-140	Collinsella SGB14624	Collinsella	Coriobacteriaceae	Actinobacteria
	14574	5.52	4.9E-125	Collinsella sp. AF28 5AC	Collinsella	Coriobacteriaceae	Actinobacteria
	1613	7.52	$0.0E + 00$	Prevotella sp. CAG 386	Prevotella	Prevotellaceae	Bacteroidetes
	1624	7.41	$0.0E + 00$	Prevotella SGB1624	Prevotella	Prevotellaceae	Bacteroidetes
	1657	7.35	$0.0E + 00$	Prevotella sp. CAG 1092	Prevotella	Prevotellaceae	Bacteroidetes
	1644	7.33	$0.0E + 00$	Prevotella sp. TF12 30	Prevotella	Prevotellaceae	Bacteroidetes
$\overline{2}$	1701	6.90	$0.0E + 00$	GGB1267 SGB1701	GGB1267	Prevotellaceae	Bacteroidetes
"Prevotella"	1680	6.77	$0.0E + 00$	Prevotella sp. CAG 520	Prevotella	Prevotellaceae	Bacteroidetes
	1699	6.68	1.1E-199	Prevotella Prevotella sp.		Prevotellaceae	Bacteroidetes
	1684	6.64	4.5E-299	Prevotella sp. CAG 924	Prevotella	Prevotellaceae	Bacteroidetes
	1560	6.46	2.1E-232	Prevotella intermedia	Prevotella	Prevotellaceae	Bacteroidetes
	1677	6.29	3.3E-253	Prevotella sp. 885	Prevotella	Prevotellaceae	Bacteroidetes
	1890	6.16	5.1E-105	GGB1385 SGB1890	GGB1385	Bacteroidaceae	Bacteroidetes
	5792	6.08	2.3E-87	Phascolarctobacterium faecium		Phascolarctobacterium Acidaminococcaceae	Firmicutes
	15132	5.95	4.8E-225	Flavonifractor plautii	Flavonifractor	Clostridiales unc.	Firmicutes
	1930	5.72	2.3E-79	Parabacteroides sp. CAG 409	Parabacteroides	Tannerellaceae	Bacteroidetes
3	1831	5.71	2.0E-66	Bacteroides sp. AM10 21B	Bacteroides	Bacteroidaceae	Bacteroidetes
"Bacteroides"	7050	5.69	5.0E-47	Lactobacillus amylovorus	Lactobacillus	Lactobacillaceae	Firmicutes
	1870	5.57	1.1E-105	Bacteroides sp. D2	Bacteroides	Bacteroidaceae	Bacteroidetes
	15143	5.38	1.1E-209	Flavonifractor sp.	Flavonifractor	Clostridiales unc.	Firmicutes
	1785	5.32	5.2E-131	Odoribacter sp. AF21 41	Odoribacter	Odoribacteraceae	Bacteroidetes
	1940	5.28	9.6E-109	Parabacteroides sp. AF18 52	Parabacteroides	Tannerellaceae	Bacteroidetes

Table S8: Species enriched in the enterotypes. For each of three enterotypes, we show the top ten species that are strongly enriched in the given enterotype against the other two enterotypes. Species composition matrix based on single-copy core gene ORFs was used as input data. Wilcoxon rank sum test was performed on each species for each given enterotype. Cases with Benjamini-Hochberg-adjusted p-value < 0.05 were selected and ordered by the fold difference between the mean compositional abundance in the target and nontarget enterotypes, to give the top 10 species.

Table S9: Number of species-level genome bins counted according to the categorizations based on pathogenicity, gut residency, and resistotype association. We defined the species as human pathogen if it was described in the Manual of Clinical Microbiology (American Society of Microbiology, Eleventh edition) as a confirmed agent in human infections of any type. We defined the species as gut resident if the frequency of its detection among the stool metagenomes from healthy adults not taking antibiotics $(n = 3,096)$ was 10% or greater. We defined the species as associated with one of the two resistotypes when the means of the compositional abundance of the SGB was different by five-fold or greater and the Benjamini-Hochberg-adjusted p value was below 0.05 based on Mann-Whitney test.

Supplementary Figures

Figure S1: Proportion of clusters non-overlapping with the isolate genome database for each ARG family categorised by antibiotic class. Clusters were generated with a threshold of 99 percent nucleotide identity (ARG cluster99) from the ARG sequences pooled from human microbiome metagenomic ORFs (sample $n = 8.972$, regardless of the body site; ORF n $= 216,849$) and NCBI RefSeq prokaryotic genomic ORFs (genome $n = 152,407$; ORF $n = 2,349,728$. Each cluster was assessed for inclusion of microbiome and RefSeq ORFs. Clusters without RefSeq ORFs were considered unique to the metagenome dataset. Each ARG family was then assessed for the proportion of its member ARG cluster99s that are unique to metagenomes. ARG families were mapped to the antibiotic classes to which they confer resistance according to the CARD database. Semi-transparent bar plots in the background display the overall proportion of metagenome-unique clusters for each antibiotic class. Each of the bubble-shaped points visualizes an individual ARG family. We altered the colour scheme between antibiotic classes to aid visual distinction between classes.

Figure S2: The number of ARG clusters discovered with varying depth of subsampling compared across body sites. (a) ARG cluster99: subsample sizes expressed as the total amount of read bases in the subsamples. (b) ARG cluster99: subsample sizes expressed as the number of subsamples. (c) Number of clusters as a functional of sequencing depth at different cut-offs for the gut only. Smoothed lines were generated by the gam function of mgcv as implemented in the R library ggplot2. In (a) and (b) the x-axis was trimmed at the maximum sample size of oral cavity, as displaying the full sample size of stool would dwarf the other body sites.

Figure S3: Comparison between assembly-based and read-based profiling of ARGs. (a-b) Two dimensional heatmaps estimating sample distributions. Data points were generated from 5,341 adult stool samples analyzed by both methods. (a) Sample distribution of the number of ARG families detected per sample with read-based (x-axis) and assembly-based (y-axis) methods. (b) Sample distribution of the total ARG abundance estimated per sample with read-based (x-axis) and assembly-based (y-axis) methods. (c) Individual cpg estimations for an ARG family in a sample using read-based (x-axis) and assembly-based (y-axis) methods were compared. Data points originated from 228,635 occurrences of any ARG family throughout the adult stool metagenomes within either read- or assembly-based method. (d) Density plots comparing read-based cpg value distributions of the ARG family occurrences according to whether the occurrence was detected or not detected using the assembly-based method.

Figure S4: Country level antibiotic consumption rate data from CD-DEP and WHO. (a) The x-axis lists the 20 countries for which we obtained gut resistome profiles from at least 10 samples. Countries are ordered based on continent. If antibiotic consumption rate data was available from either source for a country then it is shown as a bar on the y-axis as defined daily dose (DDD) per 1,000 (capita) per day summed across all antibiotics. (b) Scatter plot of the 12 countries that were covered by both CDDEP and WHO data. Pearson's correlation $r = 0.77$, $p = 0.0036$.

Figure S5: Comparison of plasmid-borne and multi-species cluster distributions and assignments. (a) Scatter-plot of ARG families giving the proportion of ORFs in multi-species clusters (x-axis) and the proportion of ORFs on plasmid contigs (y-axis). (b) Proportions of metagenomic ORFs detected on plasmid contigs plotted by the antibiotic class the genes confer resistance to. (c) Proportions of metagenomic ORFs attributable to multi-species clusters in each ARG family plotted by the antibiotic class the genes confer resistance to. Colors alternated simply to aid visual recognition. Vertical bars on the background show the antibiotic class-wide proportion. (d) Scatter plot of ARG families giving the proportion of ORFs on plasmid contigs estimated for the RefSeq genomic ORFs (x-axis) and the gut metagenomic ORFs (y-axis). (e) Scatter plot of ARG families giving the proportion of ORFs in multi-species clusters estimated for the RefSeq genomic ORFs (x-axis) and the gut metagenomic ORFs $(y-axis)$.

Figure S6: Proportion of multi-species cluster ORFs according to their proximity to the marker genes of mobile genetic elements. For each of three types of mobile genetic elements (MGEs), we grouped the ARG ORFs based on the distance to the nearest MGE marker genes found on the same contig. The number and the proportion of ORFs that belong to multispecies clusters (99%-identity) were plotted for each interval. 'Unlinked' is when there is not a target MGE marker gene on the same contig. (a) Distance to the conjugative system marker genes as defined in ConjDB. (b) Distance to the insertion sequence element transposases as defined in the ISFinder. (c) Distance to the integron integrases using the target references (AAQ16665.1, AAT72891.1, AAO32355.1, and 99031763) and outgroup references (P0A8P6.1 and P0A8P8.1). For conjugative systems and IS transposases (a, b) we used distance bins ranging up to 100 Kbp, since the sizes of known conjugative mobile genetic elements typically span 20 - 100 Kbp [1] and composite transposons up to 80 Kbp [2]. For integron integrases (c) we used shorter intervals for the distance bins, up to 20 Kbp, as the majority of integrons in bacterial genomes have a length from 5'-CS to 3'-CS lower than 10 Kbp [3].

Figure S7: Country level correlation between antibiotic consumption rates versus median abundance and richness of ARGs across gene mobility categories. In (a,b) we display correlations between the median abundance (cpg) of ARGs in each gene mobility category and the antibiotic consumption rate (DDD per 1,000 per year) across the countries. (a) Using gene mobility categorization based on LCA assignments on ARG cluster99. (b) Using gene mobility categorization based on PlasmidNet. In (c,d) we display correlations between the number of ARG cluster99s recovered from rarefied subsamples $(100 \pm 10 \text{ Gbp per country})$ and the antibiotic consumption rates (DDD) per 1000 per year) across the countries. (c) Using gene mobility categorization based on LCA assignments on ARG cluster99. (d) Using gene mobility categorization based on PlasmidNet. Each panel is split with separate plots for CDDEP and WHO consumption rate data. Vertical lines indicate the range from the first quartile to the third quartile in per- sample ARG abundance (cpg) or the number of ARG cluster99. Pairwise correlation tests were performed, excluding China from the CCDEP correlations (see Table S4 for the correlation test statistics). The number of metagenome samples used to derive the median and the range bars shown for each country in A and B: AUT ($n =$ 16), CAN (n = 36), CHN (n = 340), DEU (n = 103), DNK (n = 401), ESP (n $=$ 139), FRA (n = 62), ITA (n = 33), KAZ (n = 168), NLD (n = 470), SWE (n $= 109$), USA (n $= 147$). The number of rarefactions performed to derive each box plot and range bar shown in C and D: $n = 99$.

Figure S8: Clustering of adult gut resistome profiles and species compositions. The log-transformed ARG family profiles (cpg) of 5,372 non-outlier adult gut metagenomes in which three or more ARG families were detected, used as input for Figure 3A, was also used here throughout (a) – (e). (a) NMDS projection of resistome profiles based on three different dissimilarity measures, with a sample color scheme corresponding to the partitioning-around-medoid (PAM) clustering at $k = 2$. (b) Average silhouette width assessed after running PAM clustering on three different multivariate dissimilarity measures, i.e., Bray-Curtis, Euclidean, and Manhattan, with various pre-defined number of clusters, k. (c) Elbow plot, showing the total within-cluster sum of squares resulting from k-means clustering using various k from 1 to 20. (d) UMAP projection of resistome profiles with a sample color scheme corresponding to the PAM clusterings at $k = 2$ shown in the panel (a). (e) NMDS projection of species compositional profiles based on Bray-Curtis dissimilarity metric. Samples were colored according to PAM chastering with k value of 3. (f) Optimal k value for PAM clustering screened with an average silhouette width plot and an elbow plot. (g) Cross counts between the two resistotypes and the three enterotypes. Enterotype numbering corresponds to what are shown in the panel (f): 1, Blautia sp.; 2, Bacteroides sp.; 3, Prevotella copri.

Figure S9: ARG diversity and species diversity of the two resistotypes. (a-b) Rarefaction curves of ARG clusters in the background and FAMP resistotype population. (a) Sample size on the x-axis expresses the total amount of bases in the raw sequence reads. (b) Sample size on the x-axis expresses the number of samples analyzed. All axes are in log scale. Fitting lines were drawn using a linear model using ggplot2 R package. (c-d) Species alpha-diversity of the stool metagenomes. (c) The number of observed species and Shannon's index (H') in the adult stool metagenomes that were subjected to resistotype analysis. The estimates are based on the species composition profiling based on the assembled ORFs that represent universal single-copy core COGs (see Methods). (d) Comparison of the distribution of the number of observed species in the background and the FAMP resistotypes. We repeated comparison by restricting the species counts to each of the four dominant phyla in the human gut microbiome. Comparisons were restricted to the samples from healthy subjects who were not taking antibiotics. The numbers displayed inside the plot area are the median value of each resistotype. P values displayed over the plots were determined by two-sided Wilcoxon rank sum tests.

Figure S10: Stratification of the differences between background and FAMP resistotypes based on the species association detected for ARG clusters. (a) Species-level genome bins (SGBs) in the stool metagenomes are categorized into pathogens and non-pathogens and into the residents and infrequent colonizers. (b) Enrichment in a resistotype was defined as a median abundance at least five-fold greater in one resistotype than the other and a Benjamini-Hochberg adjusted $p < 0.05$ using two-sided Mann-Whitney test. In (a) and (b) the population level frequency of each SGB was calculated as the proportion of positive samples among the stool metagenomes from healthy adult subjects who were not taking antibiotics ($n = 3,096$). (c) ARG cluster99s were categorized by the combination two factors: whether the cluster's species range contains multi-species or single-species; and whether the ORFs from pathogens and/or non-pathogenic residents were contained. ARG abundance (cpg) values in each sample were then summed by the resulting categories and shown as boxplots. (d) Additionally, the summed cpg values were calculated for each of the four major bacterial phyla found in the stool metagenomes. Affiliation of an ARG cluster99 into a phylum was determined by the least-common ancestor taxon of the SGBs that contributed ORFs to the cluster. The number of metagenome samples used in c and d: $n = 2823$ for background resistotype, $n =$ 2272 for FAMP resistotype. In the box plots, the box spans from 25th to 75th percentiles, line gives the median, and the whisker spans from the minimum to the maximum values.

Figure S11: Frequency of the background and FAMP resistotypes in the subgroups of subjects defined by disease status and antibiotic exposure status. (a) Resistotype frequencies in the subjects compared across disease status. We used stool metagenome samples from adults that were assigned to resistotype and had disease status metadata $(n = 5,260)$. Disease status of a subject was determined from the original publication that the sample was described in. (b) Resistotype frequencies in the healthy subjects compared across antibiotic exposure status. Starting with the samples in the 'healthy' category described in the panel (a) $(n = 3.522)$, we further restricted to the samples that we could identify either past antibiotic usage within the three months prior to the sampling or confirmable 'antibiotic-free' period of any length prior to the sampling $(n = 2,719)$. Categories for 'not exposed' periods were defined in a redundant way rather than in a non-exclusive way, thus an individual sample belonging to 'not exposed in 3 month' apparently also belong to 'not exposed currently' and 'not exposed in 1 month' categories. Abbreviations used to indicate health and disease states in (A): IBD, inflammatory bowel disease; CDI, Clostridium difficile infection; CDI-risk, the patients diagnosed to be at risk of developing CDI, which is known to be induced by antibiotics; RA, rheumatoid arthritis; Metabolic dis(eases), encompassing type 2 diabetes, hypertension, and fatty liver disease; LC, liver cirrhosis; STEC, shiga toxin-producing Escherichia coli infection.

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