

1 **Uncovering the prevalence and drivers of antibiotic resistance genes in soils**
2 **across different land-use types**

3 Jie Wu^a, Shumin Guo^a, Haiyan Lin^a, Kejie Li^a, Zhutao Li^a, Jinyang Wang^{a,b*}, William H. Gaze^c, Jianwen
4 Zou^{a,b}

5 ^a *Key Laboratory of Green and Low-carbon Agriculture in Southeastern China, Ministry of Agriculture*
6 *and Rural Affairs, College of Resources and Environmental Sciences, Nanjing Agricultural University,*
7 *Nanjing 210095, China*

8 ^b *Jiangsu Key Laboratory of Low Carbon Agriculture and GHGs Mitigation, Jiangsu Collaborative*
9 *Innovation Center for Solid Organic Waste Resource Utilization, Nanjing 210095, China*

10 ^c *European Centre for Environment and Human Health, University of Exeter Medical School,*
11 *Environment & Sustainability Institute, Penryn Campus, TR10 9FE, United Kingdom*

12 ***Corresponding author:**

13 Nanjing Agricultural University, Nanjing 210095, Jiangsu, China

14 tel.: +86 25 8439 6286; fax: +86 25 8439 5210; e-mail: jywang@njau.edu.cn (J. Wang)

15

16

17

18

19

20

21

22

23

24 **Abstract**

25 The emergence and spread of antibiotic resistance genes (ARGs) in soil due to animal excreta and
26 organic waste is a major threat to human health and ecosystems, and global efforts are required to tackle
27 the issue. However, there is limited knowledge of the variation in ARG prevalence and diversity
28 resulting from different land-use patterns and underlying driving factors in soils. This study aimed to
29 comprehensively characterize the profile of ARGs and mobile genetic elements and their drivers in soil
30 samples collected from 11 provinces across China, representing three different land-use types, using
31 high-throughput quantitative polymerase chain reaction and 16S rRNA amplicon sequencing. Our
32 results showed that agricultural soil had the highest abundance and diversity of ARGs, followed by tea
33 plantation and forest land. A total of 124 unique ARGs were detected in all samples, with shared
34 subtypes among different land-use patterns indicating a common origin or high transmission frequency.
35 Moreover, significant differences in ARG distribution were observed among different geographical
36 regions, with the greatest enrichment of ARGs found in southern China. Biotic and abiotic factors,
37 including soil properties, climatic factors, and bacterial diversity, were identified as the primary drivers
38 associated with ARG abundance, explaining 71.8% of total ARG variation. The findings of our study
39 demonstrate that different land-use patterns are associated with variations in ARG abundance in soil,
40 with agricultural practices posing the greatest risk to human health and ecosystems regarding ARGs.
41 Our identification of biotic and abiotic drivers of ARG abundance provides valuable insights into
42 strategies for mitigating the spread of these genes. This study emphasizes the need for coordinated and
43 integrated approaches to address the global antimicrobial resistance crisis.

44 **Keywords:** antibiotic resistance genes; large-scale field survey; land-use patterns; shared ARGs;
45 environmental factors

46 **1 Introduction**

47 Worldwide, the prevalence of antibiotic resistance genes (ARGs) in microbial communities is
48 considered a public health challenge, which is associated with the increased use of antibiotics in humans
49 and animals and has generated extensive concern (Kuppusamy et al., 2018; Martinez, 2009; Sarmah et
50 al., 2006; Wang et al., 2022). Soil has been well-documented as a rich reservoir of ARGs, and China
51 has witnessed the most serious symptoms of this urgent challenge (Lyu et al., 2020; Qiao et al., 2018).
52 In recent years, a few studies focused on ARGs in cropland and pristine forest soils, revealing large-
53 scale spatial patterns in China (Du et al., 2020; Song et al., 2021). The prevalence of ARGs with
54 different fertilization and cropping regimes has also been previously reported (Wang et al., 2018).
55 Clearly, further research is warranted to better understand the ARG profiles for different land-use types
56 at a national scale.

57 The distribution profile of ARGs in soil ecosystems with different land-use types located in the
58 same geographical region remains poorly characterized. The enrichment of ARGs in soils under
59 different land-use patterns may be different owing to the degree of anthropogenic impact, which
60 underscores the imperative for complementary studies. Forest soils contain less ARG contamination
61 primarily due to minimal impact from anthropogenic activities (Song et al., 2021; Xiang et al., 2020).
62 Several reports indicate that high diversity of ARGs has been documented in forest soils, emphasizing
63 forest environments are also likely hidden reservoirs of ARGs (Hu et al., 2018; Radu et al., 2021). A
64 regional case in the Amazon rainforest confirmed that deforestation and subsequent shifts to other land-
65 use regimes enriched the diversity and abundance of the ARGs (Lemos et al., 2021). Further evidence
66 suggests that ARG abundance is significantly higher in cropland than in forest land (Zhu et al., 2022).
67 In stark contrast to forest environments that have relatively limited human impact, common agricultural

68 practices include the introduction of livestock manure, sewage sludge, and biosolids to tea plantations
69 and agricultural soil that are assumed to promote the enrichment of ARGs (Cerqueira et al., 2020; Heuer
70 et al., 2011; Urra et al., 2019). Furthermore, this phenomenon may have been exacerbated by
71 agricultural practices that replaced synthetic fertilizers with manure in recent years (Ji et al., 2020;
72 Udikovic-Kolic et al., 2014; Zhang et al., 2020). In particular, agricultural soils that are more exposed
73 to long-term application of manure have significantly altered patterns of antibiotic-resistant bacteria
74 with proliferation and increased persistence of ARGs in soils (Liu et al., 2022; Peng et al., 2017). Also,
75 as noted previously, ARG pollution can be transferred to the surrounding environment through runoff
76 and aerosols in a long-term and persistent manner, which enhances the dissemination of ARGs among
77 different land-use types (Bai et al., 2022; Cáliz et al., 2022; Zhang et al., 2021).

78 Several studies have confirmed that soil properties and bacterial community diversity are the
79 dominant drivers in shaping ARG diversity and abundance (Wei et al., 2022; Wu et al., 2023; Yang et
80 al., 2019). For example, soil physicochemical variables, such as soil organic carbon (SOC) and total
81 nitrogen (TN) content, are strongly associated with the distribution and prevalence of ARGs (Guo et al.,
82 2020; Wang et al., 2020). Clay affects the distribution of ARGs by shaping bacterial communities
83 (Awasthi et al., 2019; Wang et al., 2020). pH can alter the hydrophobicity of antibiotics in soils by
84 affecting electrostatic repulsion between the negatively charged soil surface and anionic forms of
85 antibiotics and further reduce the abundance of ARGs by removing antibiotics (Conde-Cid et al., 2020;
86 Li et al., 2020). Heavy metals as widespread and persistent contaminants in soil may trigger the co-
87 selection of ARGs and contribute to their maintenance and propagation (Baker-Austin et al., 2006;
88 Seiler and Berendonk, 2012). A nationwide survey of the geographic distribution of major ARGs in
89 agricultural soils revealed a strong correlation between sulfonamide resistance genes and mean annual

90 temperature (MAT) and mean annual precipitation (MAP) (Zhou et al., 2017). Warming had a positive
91 effect on the amount and proportion of natural ARGs in forest soils, and climatic seasonality is
92 associated with the dissemination of ARGs and was found to be a critical driver of their global
93 occurrence (Delgado-Baquerizo et al., 2022; Z. Li et al., 2022). Soil properties and the bacteria
94 associated with them act jointly to determine ARG profiles (Chen et al., 2016; Xie et al., 2018; Zhou et
95 al., 2020). The primary driver of ARG diversity and geographic distribution in natural wetlands is the
96 bacterial community structure (Yang et al., 2019). Furthermore, bacterial abundance and diversity are
97 the key drivers affecting the evolution of ARGs in aerobic composting (Chen et al., 2022). However,
98 previous large-scale investigations of the prevalence and drivers of ARGs have not considered
99 geographic differences in regional climate factors.

100 Given the above, the factors affecting ARG distribution are complex, and this study was aimed to
101 investigate the variations in prevalence of ARGs and potential driving mechanisms involved in the
102 propagation of ARGs in three land-use regimes (forest land, tea plantation, and agricultural land) at a
103 national scale. By combining high-throughput quantitative polymerase chain reaction (HT-qPCR)-
104 based SmartChip systems and Illumina sequencing of bacterial 16S rRNA genes, the specific objectives
105 of this study were as follows: (1) to compare the occurrence of the abundance and diversity of ARGs
106 among different land-use patterns; (2) to reveal the controlling biotic and abiotic factors driving ARG
107 profiles; and (3) to discuss practical measures to control ARG pollution under specific land-use patterns.
108 The outcomes will provide insights into feasible strategies regarding ARG reduction in soils.

109 **2 Materials and Methods**

110 **2.1 Sampling sites and sample collection**

111 The representative sites were selected from 11 provinces (AH, Anhui; FJ, Fujian; GZ, Guizhou; HEN,

112 Henan; HUN, Hunan; JS, Jiangsu; JX, Jiangxi; SC, Sichuan; SX, Shanxi; YN, Yunnan; ZJ, Zhejiang)
113 in China (range of latitude, 25.0°-33.5°N, and range of longitude, 101.6°-120.3°E), covering multiple
114 climate zones. Each sampling point selected had to cover three different land use patterns (forest land,
115 tea plantation, and agricultural land) and without livestock farms and farmers raising livestock in a
116 scattered manner nearby. At each sampling site, the upper 20 cm of soil was collected from three
117 neighboring ecosystems. Three individual samples were combined randomly selected within the target
118 areas into large composite samples to ensure the representativeness of samples (three samples per land-
119 use pattern for the same sampling site and 1.5 kg of soil per sample).

120 According to our survey, the addition of exogenous fertilizers with minimal human interference
121 did not affect forest land. Each tea plantation had been in operation for at least 10 years. The agricultural
122 land had been annually cropped with traditional crops (e.g., rice, wheat, and vegetable). Tea plantations
123 and agricultural land were managed according to conventional agricultural practices, which usually
124 received manure application (e.g., swine manure, cattle manure, and chicken manure). Collected soil
125 samples were transported to the laboratory within three days. A total of 33 soil samples were sieved
126 through a 2 mm mesh and thoroughly homogenized. A portion of each soil sample was air-dried for
127 chemical property measurements, and another portion was stored at -80°C until DNA extraction.
128 Detailed information on the sampling sites is listed in Supporting Information (SI) Table S1.

129 **2.2 Measurements of soil properties**

130 Soil physicochemical properties, including soil pH, SOC, TN, NH_4^+ , NO_3^- , and soil texture (clay, sand,
131 and silt) were measured using conventional methods as described previously (Cheng et al., 2016; Wang
132 et al., 2020; Zhu et al., 2022). Soil pH was measured by a pH probe (PHS-3C, Shanghai, China) at a
133 soil-to-water ratio of 1:2.5 (w/v). Soil total C and N were measured with a Multi N/C 3100 analyzer

134 (AnalytikJena, Jena, Germany). Skalar San Plus segmented flow analyzer (Skalar Analytical, Breda,
135 Netherlands) was used to measure the soil mineral N (NH_4^+ and NO_3^-). Heavy metal (Cu, Zn, Pb, Cd,
136 Ni, Cr, As, and Hg) concentrations were measured by inductively coupled plasma-mass spectrometry
137 (ICP-MS, Thermo Fisher Scientific, USA) after acid digestion (HClO_4 , HNO_3 , and HF) according to
138 EPA Method 3050B. The annual average temperature and precipitation were extracted from the
139 WorldClim database (www.worldclim.org/). Detailed information on the soil physicochemical
140 properties is presented in SI Table S1.

141 **2.3 DNA extraction and Illumina sequencing**

142 Soil DNA was extracted from 0.25 g of fresh soil by the MoBio PowerSoil™ DNA Isolation kit (Mo
143 Bio Laboratories, Carlsbad, CA, USA). The concentration and quality of DNA were measured with a
144 Nanodrop ND-2000 spectrophotometer (Thermo Scientific, USA).

145 To analyze the composition of the bacterial community, the V4 hypervariable region of the 16S
146 rRNA gene was amplified using universal primers 520F (5'-AYTGGGYDTAAAGNG-3') and 802R
147 (5'-TACNVGGGTATCTAATCC-3'). After the libraries passed quality control, the results of the Smear
148 analysis were examined using a Fragment Analyzer. The mixed libraries were purified by gel cutting
149 using the QIA quick gel recovery kit (Qiagen, Venio, The Netherlands). The sequencing was performed
150 using an Illumina MiSeq PE300 (Illumina Inc., San Diego, CA, USA). The data were analyzed by the
151 QIIME pipeline (version 1.9.1) (Caporaso et al., 2010; Kemp and Aller, 2004). Sequences were
152 clustered into operational taxonomic units (OTUs) by the UCLUST clustering method to classify at a
153 97% similarity level (Edgar, 2013), and used the SILVA (v132) SSU reference database to define the
154 taxonomy of OTUs. The observed species, Chao1 index, and Shannon index were used to evaluate alpha
155 diversity for each sample (Schloss et al., 2009). The Illumina raw sequence reads were submitted to the

156 National Center for Biotechnology Information Sequence Read Archive database (accession number:
157 PRJNA914204).

158 **2.4 High-throughput quantitative PCR of ARGs**

159 HT-qPCR was performed by the SmartChip Real-time PCR system (Wafergen Inc., CA, USA) to
160 determine the abundance and composition of ARGs and mobile genetic elements (MGEs) in soil
161 samples. The system is capable of processing 5184 nano-well reactions in parallel at one time. A total
162 of 144 primer sets were selected (Table S2), targeting 137 ARGs conferring resistance to major
163 antibiotics subtypes, 6 MGEs, and 16S rRNA gene (Su et al., 2015; Wolters et al., 2018; Zhu et al.,
164 2013). Amplification conditions were: 95 °C for 10 min, followed by 40 cycles of 30 s at 95 °C and 30
165 s at 60 °C. The melting curve was automatically generated by Wafergen software. Wells with multiple
166 melting peaks or amplification efficiencies above the acceptable range (90-110%) were dropped. To
167 ensure reproducibility, a threshold cycle (CT) of 31 was used as the detection limit in all three technical
168 replicates. Only samples with two or more detections in three replicates and meeting a deviation of <20%
169 were regarded as positive. The abundance of ARGs and MGEs was normalized using 16S rRNA genes
170 from the Wafergen platform to minimize the error caused by the different abundance of background
171 bacteria and variation in DNA extraction and analysis efficiency. The CT measured by the WaferGen
172 qPCR was used to calculate the copy number of ARGs via $\text{Copy Number} = 10^{(31-CT)/(10/3)}$ (Chen et al.,
173 2017; Zhu et al., 2018). The absolute abundance (copies/g) of ARGs and MGEs was calculated by
174 multiplying the relative abundance by the absolute number of gene copies of 16S rRNA in the same
175 sample (Fu et al., 2021).

176 **2.5 Statistical analysis**

177 Statistical analyses were performed in R version 4.0.3 (R Core Team, 2020). The absolute abundance

178 and number of ARGs in soil are presented as mean values. The differences in the abundance of ARGs
179 and MGEs and the relative abundance of the dominant bacteria in soil under different land-use patterns
180 were determined using a one-way analysis of variance (ANOVA) followed by Tukey's test. Non-metric
181 multidimensional scaling (NMDS) analyses based on the Bray-Curtis distance and the permutational
182 multivariate analysis of variance (PERMANOVA, Adonis test) with 999 permutations were employed
183 to determine the dissimilarity in soil microbial diversity based on the OTU in different samples
184 (Legendre and Gallagher, 2001). The "randomForest" package was used for random forest analysis.
185 Correlations between ARGs and bacterial communities were determined using the Mantel test and
186 Procrustes analysis (Delgado-Baquerizo et al., 2018; Peres-Neto and Jackson, 2001). Spearman's
187 correlation matrix between bacterial OTUs and ARGs was calculated using the Hmisc package in R,
188 and the correspondence between bacteria and ARGs was visualized by network analysis in the Gephi
189 platform (version 0.9.2) (Li et al., 2015). Redundancy analysis (RDA) and variance partitioning analysis
190 (VPA) was applied to investigate the relationship between ARGs and the components of the bacterial
191 community, soil physicochemical properties, and climatic factors. Heatmap diagrams were generated
192 to visualize the abundance of ARG classification in different samples with the "ComplexHeatmap"
193 package in R software.

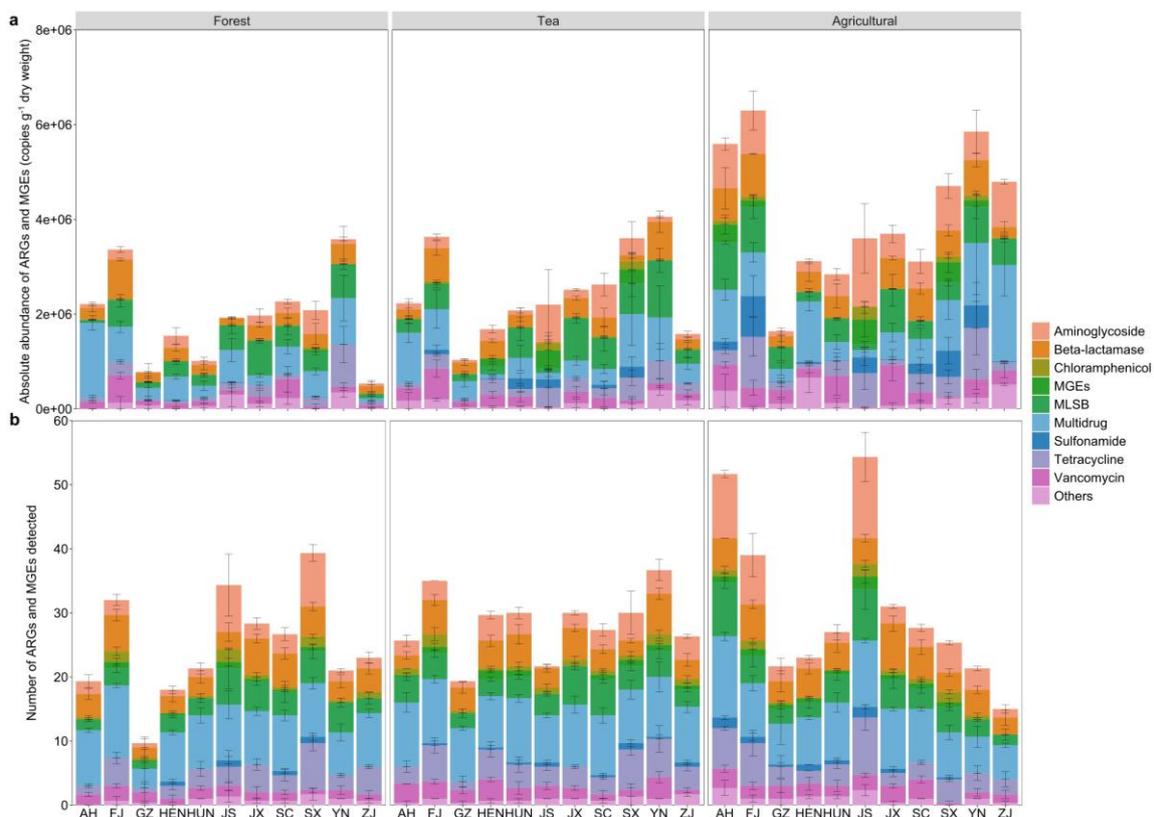
194 **3 Results**

195 **3.1 Characterization of the antibiotic resistome in soils**

196 Resistomes were classified into eight dominant ARGs categories (i.e., aminoglycoside, beta-lactamase,
197 chloramphenicol, macrolide-lincosamide-streptogramin B (MLSB), multidrug, sulfonamide,
198 tetracycline, vancomycin, and others) and MGEs, with the distribution of ARGs in each soil sample
199 shown in Fig. 1. A total of 124 unique ARGs and five MGEs were detected across all samples conferring

200 resistance to a wide range of antibiotics. Overall, the absolute abundance of ARGs for beta-lactamase
 201 (13.5%), MLSB (16.8%), and multidrug (23.6%) accounted for the majority of constituents in soils,
 202 followed by tetracycline (10.6%) and aminoglycoside (13.2%) resistance genes (Fig. S1 and Table S3).

203 The resistomes varied significantly among different land-use regimes. The abundance of ARGs in
 204 agricultural soil was considerably greater than in either tea plantation or forest land ($P < 0.001$; Fig. 1a
 205 and Table S3). Specifically, the ARG abundance was significantly increased by 27.9% in soils for tea
 206 plantation compared to forest land and 66.1% in agricultural soil compared to the tea plantation, and
 207 variations were mainly reflected in aminoglycoside and sulfonamide resistance genes ($P < 0.01-0.001$;
 208 Table S3). For collection sites, the survey in 11 provinces of China for major categories of ARGs and
 209 MGEs showed that ARG pollution was most severe in southern China, such as Fujian and Yunnan,
 210 demonstrating the large discrepancy in the ARG abundance among geographical regions (Fig. 1a).

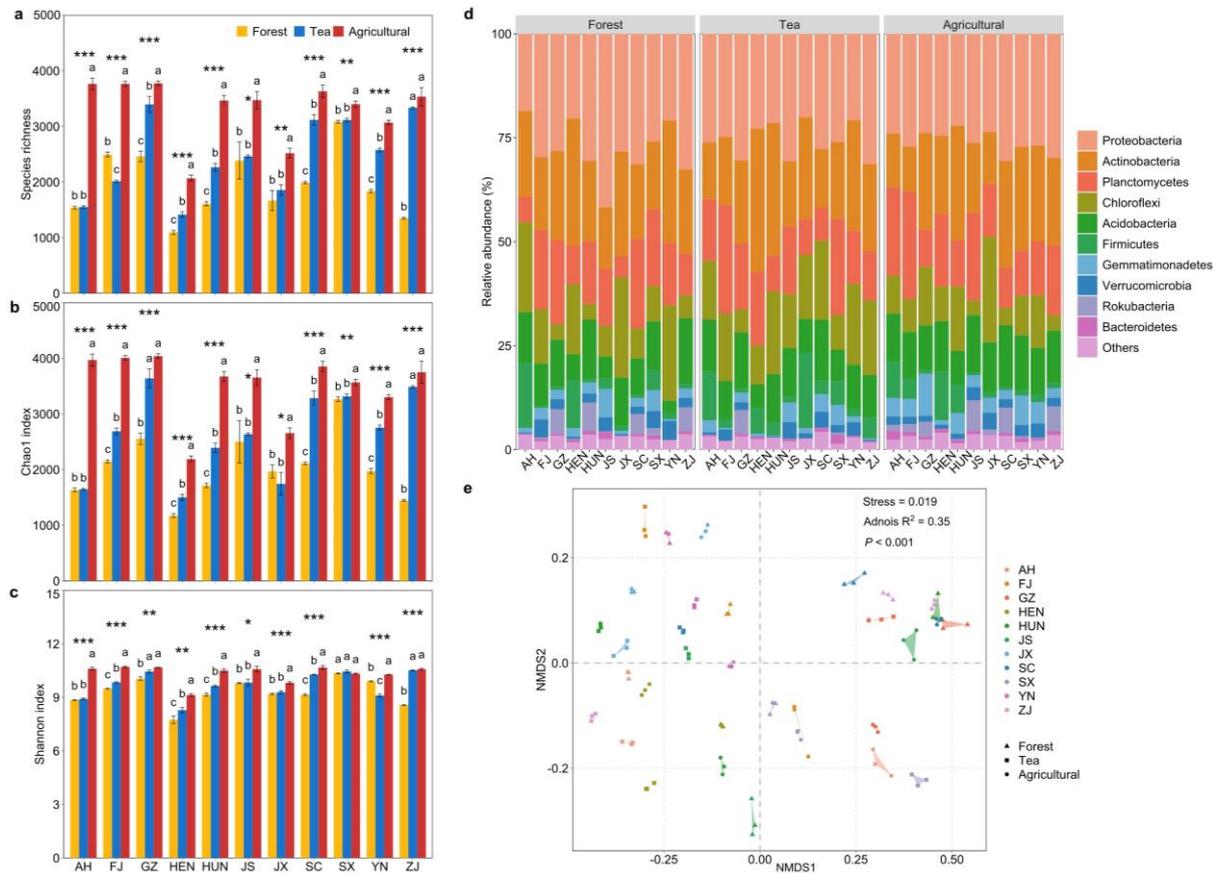


211

212 **Fig. 1 Absolute abundance (a) and number (b) of ARGs and MGEs detected in soil under different**

213 **land-use patterns.** AH, Anhui; FJ, Fujian; GZ, Guizhou; HEN, Henan; HUN, Hunan; JS, Jiangsu; JX,
 214 Jiangxi; SC, Sichuan; SX, Shanxi; YN, Yunnan; ZJ, Zhejiang.

215 As shown in Fig. 1b, the detected number of ARG subtypes and MGEs ranged from 10 to 54 in
 216 each sample. In general, the maximum ARG diversity for each region occurred mostly in agricultural
 217 soil, and the largest number of ARGs was detected in agricultural soil in Jiangsu. As expected, the
 218 diversity results of ARGs revealed by the Shannon index were generally consistent with a change in the
 219 detected number, and the Shannon index of Anhui, Henan, and Zhejiang was significantly different
 220 among the three land-use patterns ($P < 0.05-0.001$; Fig. S2). The behavior (including diversity and
 221 abundance) of MGEs was similar to that of ARGs among different land-use patterns. Furthermore, a
 222 total of 88 ARGs and MGEs were shared among soil samples in different land-use regimes, suggesting
 223 that ARG may have a high potential for transferring between soil environments (Fig. S3).



224

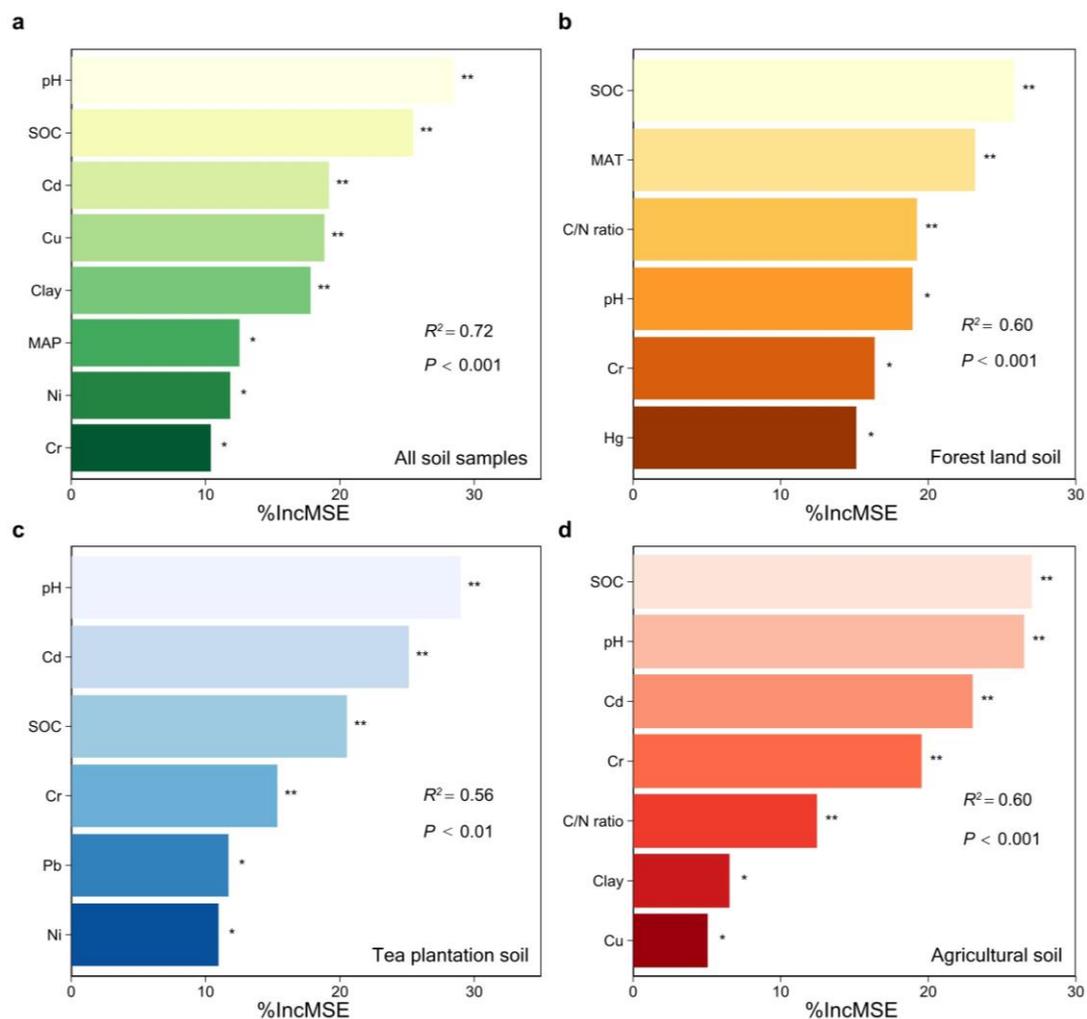
225 **Fig. 2 The alpha diversity (Species richness, Chao1 index, and Shannon index) (a, b, and c),**
226 **taxonomic profiles of bacterial communities at the phylum level (d), and non-metric**
227 **multidimensional scaling (NMDS) ordination plot of the Bray-Curtis dissimilarity matrices and**
228 **the Adonis PERMANOVA analyses of bacterial distribution (e) in soil under different land-use**
229 **patterns. AH, Anhui; FJ, Fujian; GZ, Guizhou; HEN, Henan; HUN, Hunan; JS, Jiangsu; JX, Jiangxi;**
230 **SC, Sichuan; SX, Shanxi; YN, Yunnan; ZJ, Zhejiang. Different letters refer to significant differences**
231 **among different land-use patterns at the $P < 0.05$. The alpha diversity of soil bacteria with significant**
232 **differences among different land-use patterns in the same region is indicated by asterisks. * $P < 0.05$,**
233 **** $P < 0.01$, and *** $P < 0.001$.**

234 **3.2 Characterization of the bacterial community in soils**

235 After assembly and quality filtering, a total of 10,317,764 high-quality sequences were obtained from
236 the soil by amplicon sequencing, ranging from 46,943 to 122,811 per sample, and these sequences were
237 classified into 87,169 OTUs at the 97% similarity cutoff. Rarefaction curves revealed that the depth of
238 bacterial sequencing had reached saturation and was adequate for subsequent community analysis (Fig.
239 S4). The alpha diversity of bacteria, as represented by Species richness, Chao1 index, and Shannon
240 index, showed consistent responses to land-use type, with samples from agricultural soil having
241 significantly higher bacterial diversity, followed by tea plantation and forest land ($P < 0.05-0.001$; Fig.
242 2a-c).

243 The composition of the bacterial community exhibited distinct changes at the phylum level across
244 land-use patterns. Proteobacteria, Actinobacteria, Planctomycetes, Chloroflexi, Acidobacteria, and
245 Firmicutes were the most dominant phyla in the three land-use regimes, accounting for more than 90%
246 of the whole bacterial community structure (Fig. 2d), and Proteobacteria was the most abundant

247 bacterial phyla in all samples. Differences in the taxonomic composition of bacteria among land-use
 248 patterns occurred at the level of Planctomycetes, Acidobacteria, and Gemmatimonadetes, the proportion
 249 of which was significantly increased in the agricultural soil ($P < 0.05$; Table S4). NMDS analysis of
 250 bacterial profile based on the Bray-Curtis distance revealed a clear separation among different land-use
 251 regimes, which was also confirmed by the Adonis test. Further analysis revealed that the bacterial
 252 communities also strongly clustered according to geographical location for 11 provinces ($P < 0.001$;
 253 Fig. 2e).

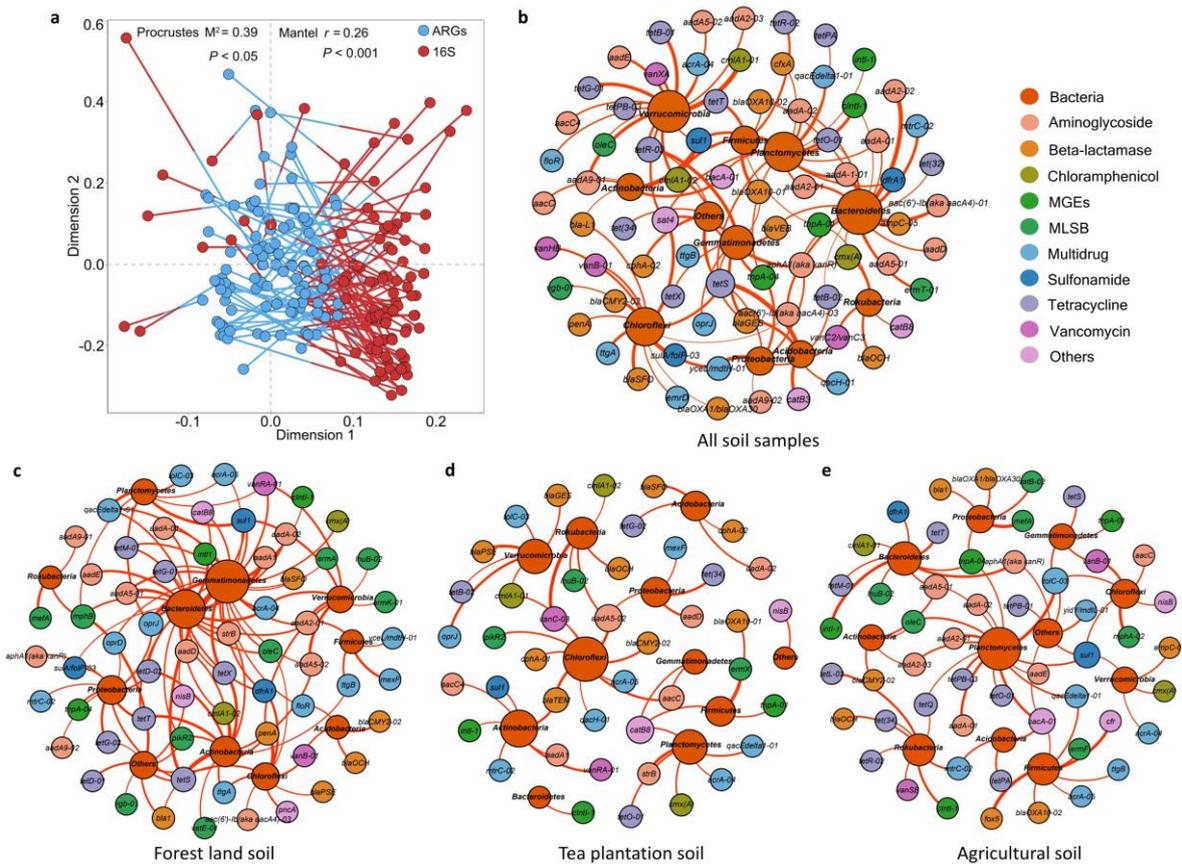


254
 255 **Fig. 3 Main factors affecting the distribution of ARGs characterized by random forest modeling**
 256 **analysis in all soil samples (a), forest land (b), tea plantation (c), and agricultural soils (d),**
 257 **respectively.** The figure shows the random forest analysis results with primary predictor importance

258 (i.e., the percent increase in mean square error [%IncMSE]) of soil properties and climatic factors on
259 ARG profiles from the soil under different land-use patterns. * $P < 0.05$, ** $P < 0.01$, and *** $P < 0.001$.

260 **3.3 Driving forces for the distribution of ARGs in soils**

261 The random forest model analysis used to identify the key soil properties and climatic factors that
262 influenced ARG abundances in soils (Fig. 3). Overall, in the model of ARG profiles ($R^2 = 0.72$, $P <$
263 0.001), the significant importance of drivers was in the order of soil pH > SOC > Cd > Cu > clay >
264 MAP > Ni > Cr in all samples (Fig. 3a). Specifically, in the forest land ($R^2 = 0.60$, $P < 0.001$), ARG
265 distribution was greatly affected by SOC, MAT, and C/N ratio (Fig. 3b). Similarly, the relatively vital
266 soil properties in the tea plantation ($R^2 = 0.56$, $P < 0.01$) were soil pH, Cd, and SOC. Comparatively,
267 the effect of soil properties on the profile of ARGs was most significant in the agricultural system (R^2
268 $= 0.60$, $P < 0.001$), and the relative importance was as follows: SOC > pH > Cd > Cr > C/N ratio >
269 clay > Cu (Fig. 3b-d).



270

271 **Fig. 4 Association of ARG distribution with the bacterial community (a) and network**

272 **visualization of ARGs and bacterial community in all soil samples (b), forest land (c), tea**

273 **plantation (d), and agricultural soils (e), respectively. Procrustes analysis and the Mantel test are**

274 **used to reveal significant associations between ARG abundance and the bacterial community at the**

275 **phylum level. Procrustes test between ARG profiles (Bray-Curtis) and bacterial communities using**

276 **taxonomic dissimilarity metrics (Bray-Curtis), where M^2 represents the sum of the squared deviations**

277 **(vector residuals) over the first two dimensions (999 permutations). The connection between ARGs and**

278 **bacterial taxa represents a strong (Spearman's correlation coefficient $r > 0.7$) and significant (P -value $<$**

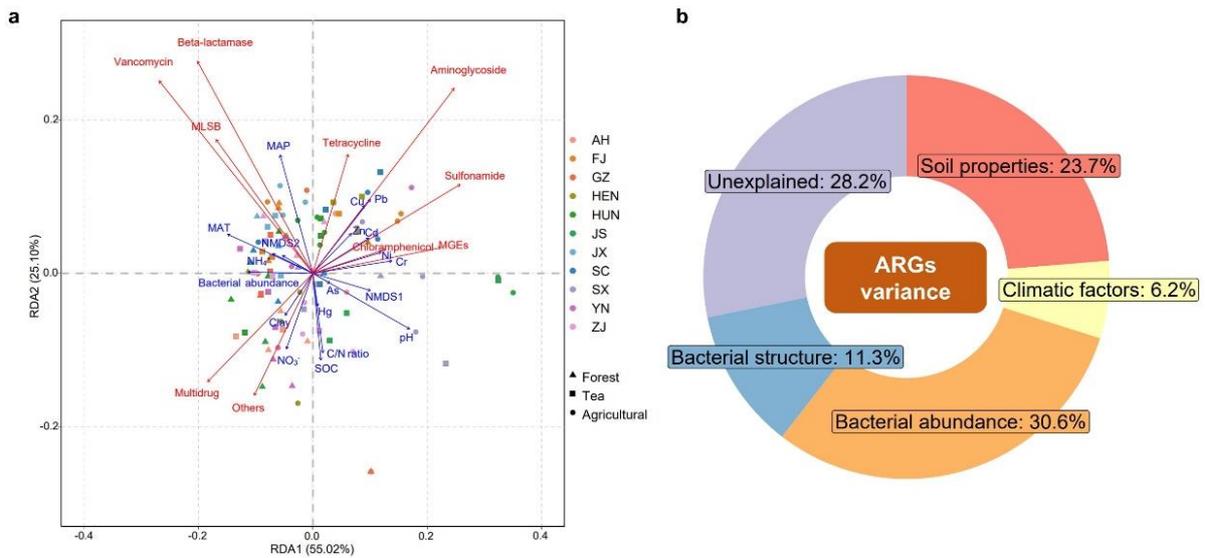
279 **0.01) correlation. The nodes are colored according to bacterial and ARG types. The node size is**

280 **proportional to the number of connections, and the edge thickness is proportional to the correlation**

281 **coefficient.**

282 Using the Procrustes and Mantel tests to explore whether the ARG profiles correlated with the
283 bacterial OTU composition using (Fig. 4a). The results of the Mantel test indicated that ARG profiles
284 were significantly correlated with bacterial composition based on Bray-Curtis distances ($r = 0.26$, $P <$
285 0.001 , permutations = 999). Moreover, the Procrustes analysis showed that most of the ARG data and
286 bacterial 16S rRNA gene OTU data of soil samples exhibited goodness-of-fit based on the Bray-Curtis
287 dissimilarity metrics ($M^2 = 0.39$, $P < 0.005$, permutations = 999).

288 The co-occurrence patterns among potential host bacteria at the phylum level and individual ARG
289 subtypes were further explored using network analysis covering all main types of ARGs and MGEs,
290 and only strong ($r > 0.7$) and significant ($P < 0.01$) correlations were displayed (Fig. 4b-e). Overall,
291 there were 85 nodes and 115 edges in the network for all samples, and the average degree and modularity
292 index were 2.706 and 0.567, respectively. ARGs and MGEs showed high connectivity with
293 Planctomycetes, Acidobacteria, Gemmatimonadetes, Verrucomicrobi, and Bacteroidetes in soils (Fig.
294 4b). Comparing specific land-use types, several bacterial phyla, i.e., Gemmatimonadetes and
295 Bacteroidetes, were most significantly associated with multiple ARGs in forest land (Fig. 4c). The
296 ARGs were weakest associated with bacteria in tea plantation soils (Fig. 4d). In contrast, agricultural
297 soil significantly increased the nodes (63) and edges (72) between ARGs and bacteria compared with
298 the tea plantation soils (Fig. 4e). The most abundant bacteria, i.e., Planctomycetes, Firmicutes, and
299 Bacteroidetes, were tightly associated with ARGs, and may be the major carriers of ARGs in agricultural
300 soil. In addition, one ARG may be carried by different host bacteria.



301

302

303

304

305

306

307

308

309

310

311

312

313

314

315

316

Fig. 5 Redundancy analysis (RDA) (a) and variation partitioning analysis (VPA) (b) depict the contribution of biotic and abiotic factors to the variation of ARGs in soils. AH, Anhui; FJ, Fujian; GZ, Guizhou; HEN, Henan; HUN, Hunan; JS, Jiangsu; JX, Jiangxi; SC, Sichuan; SX, Shanxi; YN, Yunnan; ZJ, Zhejiang. Using soil physicochemical characteristics, climatic factors, bacterial abundance, and bacterial community structure (two axes of NMDS) as explanatory variables. The blue and red arrows represent the explanatory and response variables, respectively, where greater lengths indicate stronger correlations between the pivotal factor and ARG distribution. The angles between arrows reveal the correlations between respective environmental parameters and individual ARGs.

Redundancy analysis (RDA) was used to analyze the effect of soil properties, climatic factors, bacterial abundance, and bacterial structure on the ARG profiles. The results revealed that 80.12% of the total variation could be explained by the first two axes of the RDA (Fig. 5a). Variation partitioning analysis (VPA) indicated that our selected biotic and abiotic factors explained 71.8% of total ARG variation (Fig. 5b). The soil properties contributed 23.7% of total ARG variation and climatic factors (i.e., MAT and MAP) influenced 6.2% of the variance in ARGs. For the biotic factors, bacterial abundance (30.6%) has a larger impact on the distribution of ARGs than bacterial community structure

317 (11.3%).

318 **4 Discussion**

319 **4.1 Occurrence of ARGs in different land-use patterns**

320 Variations in fertilizer application regimes result in different artificial pressure for the proliferation of
321 ARGs among the three land-use patterns. Compared to forest and tea plantation soils, the anthropogenic
322 inputs of manure, biosolids, and sewage sludge to agricultural soils are considered to significantly
323 increase the levels of ARGs as they are reservoirs of ARGs and resistant bacteria (Heuer et al., 2011;
324 Kuppusamy et al., 2018; Udikovic-Kolic et al., 2014). Several studies indicate that agricultural practices
325 are the main source of antibiotics entering the soil, and the concentration of veterinary antibiotics found
326 in cultivated land was generally higher than in forest land, associated with increased ARGs in soils
327 (Zhao et al., 2020b, 2020a). In particular, agricultural areas with continual manure application enhanced
328 the occurrence and number of ARGs in the soil compared to chemical fertilizer (Liu et al., 2022; Peng
329 et al., 2017). Likewise, our results were consistent with the above-mentioned findings that the
330 abundance and diversity of ARGs were sequentially higher in the forest land, tea plantation, and
331 agricultural soils. Nevertheless, for the specific ARG types, the dissimilarity was only in
332 aminoglycoside and sulfonamide, which may be directly related to the type of antibiotics used in the
333 local livestock farming management (Fig. 1 and Table S3). In addition, from the geographical location,
334 the southern region, like Fujian province, has more serious pollution of ARGs, and the trend of higher
335 ARG contamination was also presented in the less economically developed areas like Yunnan province,
336 accounted for by the abuse of manure containing antibiotics (Fig. 1a). Research on large scale patterns
337 of antibiotic resistance in Chinese farmland soils also revealed that ARGs presented distinct geographic
338 regional differences and highlighted the role of anthropogenic activities in shaping ARGs in soils (Du

339 et al., 2020).

340 Undoubtedly, we cannot ignore the fact that forest soil is a huge potential reservoir of ARGs (Table
341 S3). A nationwide survey of soil resistome from pristine forests across China still detected ARGs
342 conferring to major classes of antibiotics, and a high abundance of ARGs can be found in forest soils
343 (Radu et al., 2021; Song et al., 2021). Geographical distance and environmental variables considerably
344 influence the distribution and abundance of ARGs in forest soils with little human disturbance (Song et
345 al., 2021). More importantly, 68.2% of ARG subtypes co-existed in soil samples with different land-use
346 patterns, indicating a common origin or high transmission frequency (Fig. S3). Spread of ARGs through
347 runoff and airborne between land-use types within the same region is supported by several evidences.
348 Distribution characteristics of ARGs in a representative rivers-reservoir system suggested rainfall-
349 induced seasonal runoff may promote the abundance of ARGs and MGEs (Chen et al., 2019). Long-
350 term atmospheric deposition monitoring found sulfonamide resistance genes and *int11* to be long-range
351 and continuously dispersed in free tropospheric aerosols and ARGs from animal feces and feed yards
352 origin accumulated in bioaerosols and were dispersed over long distances (Bai et al., 2022; Cáliz et al.,
353 2022; McEachran et al., 2015; Zhou et al., 2023).

354 **4.2 Biotic and abiotic factors influencing the distribution of ARGs**

355 As we expected, the soil properties and climatic factors that influence the persistence and spread of
356 ARGs and antibiotic residues vary in relative importance under different land-use practices (Cycoń et
357 al., 2019; Wu et al., 2023). Overall, ARG distribution was strongly influenced by pH, SOC, and multiple
358 heavy metals (Cu, Zn, Pb, Cd, Ni, Cr, As, and Hg) (Fig. 3a). A study targeting the large-scale profile of
359 antibiotic resistome in forest soils found significant relationships between temperature and spatial
360 variation in ARGs (Song et al., 2021). Our study confirmed that forest soil ARGs were mainly

361 associated with SOC, MAT, and C/N ratios, and further supported this explanation (Fig. 3b). This may
362 promote the degradation of antibiotics due to the high organic carbon and nitrogen content induced by
363 plant litter as well as the fact that forest vegetation regulates temperature differences (Zhang et al., 2023).
364 Shared bacteria and resistome in the soil-plant systems suggested the possibility of transfer for ARGs
365 through the internal plant tissue (Zhang et al., 2019). Resistomes were altered by plant identity (Zheng
366 et al., 2022), and forest vegetation may have a better capacity to absorb ARGs compared to tea and
367 agricultural crops. pH is the primary abiotic factor affecting the variation of ARG abundance in tea
368 plantation soils (Fig. 3c). Increasing the pH was associated with a decrease in target ARGs from manure
369 slurries (Li et al., 2020). SOC and pH were the top two soil attributes affecting ARG abundance in
370 agricultural soils (Fig. 3d), which showed a significant negative correlation with ARGs (Wang et al.,
371 2020; Wu et al., 2023). Heavy metals, such as Cd and Cu (Figs. 3 and 5b), introduced into the soil with
372 long-term fertilization driving the risk of antibiotic resistance co-selection is a concern (Baker-Austin
373 et al., 2006). The use of zinc as an agricultural feed additive has driven the development of antibiotic
374 resistance in metal-exposed bacteria due to the selective effect of metals on genetic elements that carry
375 both metal and antibiotic resistance genes (Dickinson et al., 2019; Poole, 2017). Co-occurrence of ARGs
376 and heavy metals are characterized much more frequently in human pathogens than in bacteria that are
377 less frequently associated with humans (Li et al., 2017). Furthermore, it was found that warming and
378 increased precipitation have positive impacts on specific ARGs (e.g., beta-lactamase, vancomycin, and
379 MLSB) but have negative impacts on other categories (Fig. 5a), and climate seasonality was also
380 determined as a key driver of their global profile (Delgado-Baquerizo et al., 2022; Z. Li et al., 2022).
381 On the whole, it can be observed that pH and SOC have a negative effect on the majority of ARG types,
382 while several heavy metals and MAP have a positive effect in the present study.

383 Bacterial community composition (i.e., abundance and structure) is also a primary driver shaping
384 ARG diversity and evolution (Chen et al., 2022; Yang et al., 2019), with a greater influence than abiotic
385 factors (Fig. 5b). By network analysis, Planctomycetes, Acidobacteria, and Gemmatimonadetes were
386 considered as the major potential hosts of ARGs, and the variation in ARG abundance may be related
387 to their carrying aminoglycoside and sulfonamide in soils (Fig. 4b and Tables S3-4). The high
388 connectivity of ARGs with several bacterial phyla (i.e., Gemmatimonadetes and Bacteroidetes) in forest
389 systems reflected the capacity of forest soils as potential reservoirs of ARGs (Fig. 4c). In contrast, the
390 weakest correlation between ARG and host bacteria was in tea plantation soils, and ARGs were mainly
391 associated with Chloroflexi (Fig. 4d). Previous studies have also found enrichment of soil resistome
392 was strongly influenced by bacterial taxa in manure-amended soils, including Gemmatimonadetes and
393 Chloroflexi (Zhang et al., 2019). Although Proteobacteria, as the most enriched bacterial phylum, was
394 considered the major potential host of ARGs in agricultural soils (T. Li et al., 2022), Planctomycetes,
395 Firmicutes, and Bacteroidetes may be the main carriers of ARGs in this study (Fig. 4e). In addition, the
396 positive correlation of MGEs with sulfonamide and aminoglycoside further indicates their role in
397 carrying ARGs (Gaze et al., 2011; Peng et al., 2017).

398 **4.3 Mitigation options for ARG contamination from different land-use patterns**

399 To date, ARGs are an increasing global crisis in human health (Zhang et al., 2022), and it is crucial to
400 propose strategies to mitigate ARGs for specific soil environments. Antibiotic-resistant bacteria and
401 ARGs can persist and transmit to vegetables with the application of manure (Pu et al., 2019). Hence,
402 developing cost-effective antibiotic abatement methods are fundamental approaches to mitigate ARG
403 contamination from the source. Currently, aerobic composting and aerobic and anaerobic digestion of
404 manure and sewage sludge are effective methods to reduce ARGs in soils (Wang et al., 2022). For forest

405 systems, the absorption and accumulation of ARGs by plants should be fully utilized. It has been
406 reported that plants with an enrichment effect on organic pollutants in combination with biochar showed
407 stronger attenuation of typical antibiotics and ARGs in soils (Liang et al., 2017). Increasing vegetation
408 coverage near the tea plantation and agricultural land can be considered a sustainable ecological
409 approach. Agricultural field and tea plantation soils are more likely to be contaminated by
410 anthropogenic sources such as cultivation activities compared to forest systems (Xiang et al., 2020). pH,
411 as the most influential factor affecting ARGs in tea plantation soils (Fig. 3c), implied that reasonable
412 fertilization measures to mitigate the process of soil acidification in tea plantations might contribute to
413 the alleviation of ARG contamination. A systematic global analysis of ARGs found that the abundance
414 of ARGs was negatively correlated with SOC and TN in agricultural soil (Wu et al., 2023), which was
415 consistent with the observations in this study (Fig. 5a). This implied that maintaining high soil nutrients
416 may outcompete introduced ARG bearing bacteria. Meanwhile, soil quality should be selected carefully
417 when growing crops. For example, higher clay and heavy metals may significantly enhance the
418 abundance of antibiotic-resistant bacteria (Awasthi et al., 2019; Seiler and Berendonk, 2012). Moreover,
419 as ARGs are normally carried by host bacteria, it could be promising to explore efficient biological
420 control agents to maintain soil nutrients for crop growth while reducing the reproduction of bacteria
421 carrying ARGs. We also acknowledge that there are some limitations in this study, such as soil samples
422 were collected from three land-use patterns in several provinces in this study, subsequent studies could
423 enhance the representativeness of the samples by collecting multiple samples from a particular location.
424 In addition, the environmental risk of ARGs closely associated with human pathogens in each land-use
425 pattern was not clarified and the dynamic pattern of ARGs over time was not consistently tracked.

426 **5 Conclusion**

427 Through the large-scale field investigation of the presence of ARGs in different land-use practices
428 across 11 provinces across China, this study has provided compelling field evidence that the abundance
429 of ARGs increased by 66.1% in agricultural soils compared to tea plantations and by 27.9% in tea
430 plantation compared to forest land soils. Notably, the shared subtypes suggest that ARGs may have a
431 common origin or high transmission frequency between different soil environments. Although there
432 was no further evidence to determine the transfer pathway of ARGs in our study, the findings still can
433 extend our sights for regulating ARGs. The forest soil also showed a high abundance and detected
434 number of ARGs despite the forest environments being rarely affected by human activities. Intrinsically,
435 variance in ARGs induced by changes in soil properties (i.e., pH, SOC, Cd, Cu, and clay), climatic
436 factors (i.e., MAT and MAP), and bacterial community (i.e., abundance and structure) are the main
437 drivers shaping ARGs profiles in different land-use patterns. Finally, these findings are essential for
438 evaluating the potential public risks posed by ARGs in soil systems, and we highlight the necessity to
439 target ARG reduction by adopting efficient practices under different land-use regimes, especially in soil
440 types and properties of importance.

441

442 **Acknowledgment**

443 This work was supported by the National Natural Science Foundation of China (42177285), Jiangsu
444 Agriculture Science and Technology Innovation Fund (JASTIF-CX(21)3007), Jiangsu Provincial
445 Special Project for Carbon Peak Carbon Neutrality Science and Technology Innovation (BE2022308,
446 BE2022423), the Startup Foundation for Introducing Talent of Nanjing Agricultural University
447 (030/804028) and the China Scholarship Council (202106850038).

448 **References**

- 449 Awasthi, M.K., Chen, H., Liu, T., Awasthi, S.K., Wang, Q., Ren, X., Duan, Y., Zhang, Z., 2019.
450 Respond of clay amendment in chicken manure composts to understand the antibiotic resistant
451 bacterial diversity and its correlation with physicochemical parameters. *J. Clean. Prod.* 236,
452 117715. <https://doi.org/10.1016/j.jclepro.2019.117715>
- 453 Bai, H., He, L.Y., Wu, D.L., Gao, F.Z., Zhang, M., Zou, H.Y., Yao, M.S., Ying, G.G., 2022. Spread
454 of airborne antibiotic resistance from animal farms to the environment: Dispersal pattern and
455 exposure risk. *Environ. Int.* 158, 106927. <https://doi.org/10.1016/j.envint.2021.106927>
- 456 Baker-Austin, C., Wright, M.S., Stepanauskas, R., McArthur, J. V., 2006. Co-selection of antibiotic
457 and metal resistance. *Trends Microbiol.* 14, 176–182. <https://doi.org/10.1016/j.tim.2006.02.006>
- 458 Cáliz, J., Subirats, J., Triadó-Margarit, X., Borrego, C.M., Casamayor, E.O., 2022. Global dispersal
459 and potential sources of antibiotic resistance genes in atmospheric remote depositions. *Environ.*
460 *Int.* 160. <https://doi.org/10.1016/j.envint.2022.107077>
- 461 Caporaso, J.G., Kuczynski, J., Stombaugh, J., Bittinger, K., Bushman, F.D., Costello, E.K., Fierer, N.,
462 Peña, A.G., Goodrich, J.K., Gordon, J.I., Huttley, G.A., Kelley, S.T., Knights, D., Koenig, J.E.,
463 Ley, R.E., Lozupone, C.A., Mcdonald, D., Muegge, B.D., Pirrung, M., Reeder, J., Sevinsky,
464 J.R., Turnbaugh, P.J., Walters, W.A., Widmann, J., Yatsunenko, T., Zaneveld, J., Knight, R.,
465 2010. correspondence QIIME allows analysis of high- throughput community sequencing data
466 Intensity normalization improves color calling in SOLiD sequencing. *Nat. Publ. Gr.* 7, 335–336.
467 <https://doi.org/10.1038/nmeth0510-335>
- 468 Cerqueira, F., Christou, A., Fatta-Kassinos, D., Vila-Costa, M., Bayona, J.M., Piña, B., 2020. Effects

469 of prescription antibiotics on soil- and root-associated microbiomes and resistomes in an
470 agricultural context. *J. Hazard. Mater.* 400, 123208.
471 <https://doi.org/10.1016/j.jhazmat.2020.123208>

472 Chen, Q., An, X., Li, H., Su, J., Ma, Y., Zhu, Y.G., 2016. Long-term field application of sewage
473 sludge increases the abundance of antibiotic resistance genes in soil. *Environ. Int.* 92–93, 1–10.
474 <https://doi.org/10.1016/j.envint.2016.03.026>

475 Chen, Q.L., An, X.L., Zhu, Y.G., Su, J.Q., Gillings, M.R., Ye, Z.L., Cui, L., 2017. Application of
476 Struvite Alters the Antibiotic Resistome in Soil, Rhizosphere, and Phyllosphere. *Environ. Sci.*
477 *Technol.* 51, 8149–8157. <https://doi.org/10.1021/acs.est.7b01420>

478 Chen, Y., Su, J.Q., Zhang, J., Li, P., Chen, H., Zhang, B., Gin, K.Y.H., He, Y., 2019. High-throughput
479 profiling of antibiotic resistance gene dynamic in a drinking water river-reservoir system. *Water*
480 *Res.* 149, 179–189. <https://doi.org/10.1016/j.watres.2018.11.007>

481 Chen, Z., Fu, Q., Wen, Q., Wu, Y., Bao, H., Guo, J., 2022. Microbial community competition rather
482 than high-temperature predominates ARGs elimination in swine manure composting. *J. Hazard.*
483 *Mater.* 423, 127149. <https://doi.org/10.1016/j.jhazmat.2021.127149>

484 Cheng, J., Ding, C., Li, X., Zhang, T., Wang, X., 2016. Soil quality evaluation for navel orange
485 production systems in central subtropical China. *Soil Tillage Res.* 155, 225–232.
486 <https://doi.org/10.1016/j.still.2015.08.015>

487 Conde-Cid, M., Ferreira-Coelho, G., Fernández-Calviño, D., Núñez-Delgado, A., Fernández-Sanjurjo,
488 M.J., Arias-Estévez, M., Álvarez-Rodríguez, E., 2020. Single and simultaneous adsorption of
489 three sulfonamides in agricultural soils: Effects of pH and organic matter content. *Sci. Total*

490 Environ. 744, 140872. <https://doi.org/10.1016/j.scitotenv.2020.140872>

491 Cycoń, M., Mroziak, A., Piotrowska-Seget, Z., 2019. Antibiotics in the soil environment—degradation
492 and their impact on microbial activity and diversity. *Front. Microbiol.* 10.
493 <https://doi.org/10.3389/fmicb.2019.00338>

494 Delgado-Baquerizo, M., Fry, E.L., Eldridge, D.J., de Vries, F.T., Manning, P., Hamonts, K., Kattge,
495 J., Boenisch, G., Singh, B.K., Bardgett, R.D., 2018. Plant attributes explain the distribution of
496 soil microbial communities in two contrasting regions of the globe. *New Phytol.* 219, 574–587.
497 <https://doi.org/10.1111/nph.15161>

498 Delgado-Baquerizo, M., Hu, H.-W., Maestre, F.T., Guerra, C.A., Eisenhauer, N., Eldridge, D.J., Zhu,
499 Y.-G., Chen, Q.-L., Trivedi, P., Du, S., Makhalanyane, T.P., Verma, J.P., Gozalo, B., Ochoa, V.,
500 Asensio, S., Wang, L., Zaady, E., Illán, J.G., Siebe, C., Grebenc, T., Zhou, X., Liu, Y.-R.,
501 Bamigboye, A.R., Blanco-Pastor, J.L., Duran, J., Rodríguez, A., Mamet, S., Alfaro, F., Abades,
502 S., Teixido, A.L., Peñaloza-Bojacá, G.F., Molina-Montenegro, M., Torres-Díaz, C., Perez, C.,
503 Gallardo, A., García-Velázquez, L., Hayes, P.E., Neuhauser, S., He, J.-Z., 2022. The global
504 distribution and environmental drivers of the soil antibiotic resistome. *bioRxiv*
505 2022.07.11.499543. <https://doi.org/10.1186/s40168-022-01405-w>

506 Dickinson, A.W., Power, A., Hansen, M.G., Brandt, K.K., Piliposian, G., Appleby, P., O'Neill, P.A.,
507 Jones, R.T., Sierocinski, P., Koskella, B., Vos, M., 2019. Heavy metal pollution and co-selection
508 for antibiotic resistance: A microbial palaeontology approach. *Environ. Int.* 132, 105117.
509 <https://doi.org/10.1016/j.envint.2019.105117>

510 Du, S., Shen, J.P., Hu, H.W., Wang, J.T., Han, L.L., Sheng, R., Wei, W.X., Fang, Y.T., Zhu, Y.G.,

511 Zhang, L.M., He, J.Z., 2020. Large-scale patterns of soil antibiotic resistome in Chinese
512 croplands. *Sci. Total Environ.* 712, 136418. <https://doi.org/10.1016/j.scitotenv.2019.136418>

513 Edgar, R.C., 2013. UPARSE: Highly accurate OTU sequences from microbial amplicon reads. *Nat.*
514 *Methods* 10, 996–998. <https://doi.org/10.1038/nmeth.2604>

515 Fu, Y., Jia, M., Wang, F., Wang, Z., Mei, Z., Bian, Y., Jiang, X., Virta, M., Tiedje, J.M., 2021.
516 Strategy for Mitigating Antibiotic Resistance by Biochar and Hyperaccumulators in Cadmium
517 and Oxytetracycline Co-contaminated Soil. *Environ. Sci. Technol.* 55, 16369–16378.
518 <https://doi.org/10.1021/acs.est.1c03434>

519 Gaze, W.H., Zhang, L., Abdousslam, N.A., Hawkey, P.M., Calvo-Bado, L., Royle, J., Brown, H.,
520 Davis, S., Kay, P., Boxall, A.B.A., Wellington, E.M.H., 2011. Impacts of anthropogenic activity
521 on the ecology of class 1 integrons and integron-associated genes in the environment. *ISME J.* 5,
522 1253–1261. <https://doi.org/10.1038/ismej.2011.15>

523 Guo, X. pan, Zhao, S., Chen, Y. ru, Yang, J., Hou, L. jun, Liu, M., Yang, Y., 2020. Antibiotic
524 resistance genes in sediments of the Yangtze Estuary: From 2007 to 2019. *Sci. Total Environ.*
525 744, 140713. <https://doi.org/10.1016/j.scitotenv.2020.140713>

526 Heuer, H., Schmitt, H., Smalla, K., 2011. Antibiotic resistance gene spread due to manure application
527 on agricultural fields. *Curr. Opin. Microbiol.* 14, 236–243.
528 <https://doi.org/10.1016/j.mib.2011.04.009>

529 Hu, H.W., Wang, J.T., Singh, B.K., Liu, Y.R., Chen, Y.L., Zhang, Y.J., He, J.Z., 2018. Diversity of
530 herbaceous plants and bacterial communities regulates soil resistome across forest biomes.
531 *Environ. Microbiol.* 20, 3186–3200. <https://doi.org/10.1111/1462-2920.14248>

- 532 Ji, L., Ni, K., Wu, Z., Zhang, J., Yi, X., Yang, X., Ling, N., You, Z., Guo, S., Ruan, J., 2020. Effect of
533 organic substitution rates on soil quality and fungal community composition in a tea plantation
534 with long-term fertilization. *Biol. Fertil. Soils* 56, 633–646. [https://doi.org/10.1007/s00374-020-](https://doi.org/10.1007/s00374-020-01439-y)
535 01439-y
- 536 Kemp, P.F., Aller, J.Y., 2004. Bacterial diversity in aquatic and other environments: What 16S rDNA
537 libraries can tell us. *FEMS Microbiol. Ecol.* 47, 161–177. [https://doi.org/10.1016/S0168-](https://doi.org/10.1016/S0168-6496(03)00257-5)
538 6496(03)00257-5
- 539 Kuppusamy, S., Kakarla, D., Venkateswarlu, K., Megharaj, M., Yoon, Y.E., Lee, Y.B., 2018.
540 Veterinary antibiotics (VAs) contamination as a global agro-ecological issue: A critical view.
541 *Agric. Ecosyst. Environ.* 257, 47–59. <https://doi.org/10.1016/j.agee.2018.01.026>
- 542 Legendre, P., Gallagher, E.D., 2001. Ecologically meaningful transformations for ordination of
543 species data. *Oecologia* 129, 271–280. <https://doi.org/10.1007/s004420100716>
- 544 Lemos, L.N., Pedrinho, A., Vasconcelos, A.T.R. de, Tsai, S.M., Mendes, L.W., 2021. Amazon
545 deforestation enriches antibiotic resistance genes. *Soil Biol. Biochem.* 153, 108110.
546 <https://doi.org/10.1016/j.soilbio.2020.108110>
- 547 Li, B., Yang, Y., Ma, L., Ju, F., Guo, F., Tiedje, J.M., Zhang, T., 2015. Metagenomic and network
548 analysis reveal wide distribution and co-occurrence of environmental antibiotic resistance genes.
549 *ISME J.* 9, 2490–2502. <https://doi.org/10.1038/ismej.2015.59>
- 550 Li, L.G., Xia, Y., Zhang, T., 2017. Co-occurrence of antibiotic and metal resistance genes revealed in
551 complete genome collection. *ISME J.* 11, 651–662. <https://doi.org/10.1038/ismej.2016.155>
- 552 Li, M.M., Ray, P., Knowlton, K.F., Pruden, A., Xia, K., Teets, C., Du, P., 2020. Fate of pirlimycin

553 and antibiotic resistance genes in dairy manure slurries in response to temperature and pH
554 adjustment. *Sci. Total Environ.* 710, 136310. <https://doi.org/10.1016/j.scitotenv.2019.136310>

555 Li, T., Li, Ruochen, Cao, Y., Tao, C., Deng, X., Ou, Y., Liu, H., Shen, Z., Li, Rong, Shen, Q., 2022.
556 Soil antibiotic abatement associates with the manipulation of soil microbiome via long-term
557 fertilizer application. *J. Hazard. Mater.* 439, 129704.
558 <https://doi.org/10.1016/j.jhazmat.2022.129704>

559 Li, Z., Sun, A., Liu, X., Chen, Q.L., Bi, L., Ren, P.X., Shen, J.P., Jin, S., He, J.Z., Hu, H.W., Yang,
560 Y., 2022. Climate warming increases the proportions of specific antibiotic resistance genes in
561 natural soil ecosystems. *J. Hazard. Mater.* 430. <https://doi.org/10.1016/j.jhazmat.2022.128442>

562 Liang, Y., Pei, M., Wang, D., Cao, S., Xiao, X., Sun, B., 2017. Improvement of Soil Ecosystem
563 Multifunctionality by Dissipating Manure-Induced Antibiotics and Resistance Genes. *Environ.*
564 *Sci. Technol.* 51, 4988–4998. <https://doi.org/10.1021/acs.est.7b00693>

565 Liu, W., Cheng, Y., Guo, J., Duan, Y., Wang, S., Xu, Q., Liu, M., Xue, C., Guo, S., Shen, Q., Ling,
566 N., 2022. Long-term manure inputs induce a deep selection on agroecosystem soil antibiotic
567 resistome. *J. Hazard. Mater.* 436, 129163. <https://doi.org/10.1016/j.jhazmat.2022.129163>

568 Lyu, J., Yang, L., Zhang, L., Ye, B., Wang, L., 2020. Antibiotics in soil and water in China—a
569 systematic review and source analysis. *Environ. Pollut.*
570 <https://doi.org/10.1016/j.envpol.2020.115147>

571 Martinez, J.L., 2009. Environmental pollution by antibiotics and by antibiotic resistance determinants.
572 *Environ. Pollut.* 157, 2893–2902. <https://doi.org/10.1016/j.envpol.2009.05.051>

573 McEachran, A.D., Blackwell, B.R., Hanson, J.D., Wooten, K.J., Mayer, G.D., Cox, S.B., Smith, P.N.,

574 2015. Antibiotics, bacteria, and antibiotic resistance genes: Aerial transport from cattle feed
575 yards via particulate matter. *Environ. Health Perspect.* 123, 337–343.
576 <https://doi.org/10.1289/ehp.1408555>

577 Peng, S., Feng, Y., Wang, Y., Guo, X., Chu, H., Lin, X., 2017. Prevalence of antibiotic resistance
578 genes in soils after continually applied with different manure for 30 years. *J. Hazard. Mater.* 340,
579 16–25. <https://doi.org/10.1016/j.jhazmat.2017.06.059>

580 Peres-Neto, P.R., Jackson, D.A., 2001. How well do multivariate data sets match? The advantages of
581 a procrustean superimposition approach over the Mantel test. *Oecologia* 129, 169–178.
582 <https://doi.org/10.1007/s004420100720>

583 Poole, K., 2017. At the Nexus of Antibiotics and Metals: The Impact of Cu and Zn on Antibiotic
584 Activity and Resistance. *Trends Microbiol.* 25, 820–832.
585 <https://doi.org/10.1016/j.tim.2017.04.010>

586 Pu, C., Yu, Y., Diao, J., Gong, X., Li, J., Sun, Y., 2019. Exploring the persistence and spreading of
587 antibiotic resistance from manure to biocompost, soils and vegetables. *Sci. Total Environ.* 688,
588 262–269. <https://doi.org/10.1016/j.scitotenv.2019.06.081>

589 Qiao, M., Ying, G.G., Singer, A.C., Zhu, Y.G., 2018. Review of antibiotic resistance in China and its
590 environment. *Environ. Int.* 110, 160–172. <https://doi.org/10.1016/j.envint.2017.10.016>

591 Radu, E., Woegerbauer, M., Rab, G., Oismüller, M., Strauss, P., Hufnagl, P., Gottsberger, R.A.,
592 Krampe, J., Weyermaier, K., Kreuzinger, N., 2021. Resilience of agricultural soils to antibiotic
593 resistance genes introduced by agricultural management practices. *Sci. Total Environ.* 756,
594 143699. <https://doi.org/10.1016/j.scitotenv.2020.143699>

595 Sarmah, A.K., Meyer, M.T., Boxall, A.B.A., 2006. A global perspective on the use, sales, exposure
596 pathways, occurrence, fate and effects of veterinary antibiotics (VAs) in the environment.
597 *Chemosphere* 65, 725–759. <https://doi.org/10.1016/j.chemosphere.2006.03.026>

598 Schloss, P.D., Westcott, S.L., Ryabin, T., Hall, J.R., Hartmann, M., Hollister, E.B., Lesniewski, R.A.,
599 Oakley, B.B., Parks, D.H., Robinson, C.J., Sahl, J.W., Stres, B., Thallinger, G.G., Van Horn,
600 D.J., Weber, C.F., 2009. Introducing mothur: Open-source, platform-independent, community-
601 supported software for describing and comparing microbial communities. *Appl. Environ.*
602 *Microbiol.* 75, 7537–7541. <https://doi.org/10.1128/AEM.01541-09>

603 Seiler, C., Berendonk, T.U., 2012. Heavy metal driven co-selection of antibiotic resistance in soil and
604 water bodies impacted by agriculture and aquaculture. *Front. Microbiol.* 3, 1–10.
605 <https://doi.org/10.3389/fmicb.2012.00399>

606 Song, M., Song, D., Jiang, L., Zhang, D., Sun, Y., Chen, G., Xu, H., Mei, W., Li, Y., Luo, C., Zhang,
607 G., 2021. Large-scale biogeographical patterns of antibiotic resistome in the forest soils across
608 China. *J. Hazard. Mater.* 403, 123990. <https://doi.org/10.1016/j.jhazmat.2020.123990>

609 Su, J.Q., Wei, B., Ou-Yang, W.Y., Huang, F.Y., Zhao, Y., Xu, H.J., Zhu, Y.G., 2015. Antibiotic
610 Resistome and Its Association with Bacterial Communities during Sewage Sludge Composting.
611 *Environ. Sci. Technol.* 49, 7356–7363. <https://doi.org/10.1021/acs.est.5b01012>

612 Udikovic-Kolic, N., Wichmann, F., Broderick, N.A., Handelsman, J., 2014. Bloom of resident
613 antibiotic-resistant bacteria in soil following manure fertilization. *Proc. Natl. Acad. Sci. U. S. A.*
614 111, 15202–15207. <https://doi.org/10.1073/pnas.1409836111>

615 Urrea, J., Alkorta, I., Mijangos, I., Epelde, L., Garbisu, C., 2019. Application of sewage sludge to

616 agricultural soil increases the abundance of antibiotic resistance genes without altering the
617 composition of prokaryotic communities. *Sci. Total Environ.* 647, 1410–1420.
618 <https://doi.org/10.1016/j.scitotenv.2018.08.092>

619 Wang, F., Xu, M., Stedtfeld, R.D., Sheng, H., Fan, J., Liu, M., Chai, B., Soares De Carvalho, T., Li,
620 H., Li, Z., Hashsham, S.A., Tiedje, J.M., 2018. Long-Term Effect of Different Fertilization and
621 Cropping Systems on the Soil Antibiotic Resistome. *Environ. Sci. Technol.* 52, 13037–13046.
622 <https://doi.org/10.1021/acs.est.8b04330>

623 Wang, Jinhua, Wang, L., Zhu, L., Wang, Jun, Xing, B., 2022. Antibiotic resistance in agricultural
624 soils: Source, fate, mechanism and attenuation strategy. *Crit. Rev. Environ. Sci. Technol.* 52,
625 847–889. <https://doi.org/10.1080/10643389.2020.1835438>

626 Wang, L., Wang, Jinhua, Wang, Jun, Zhu, L., Conkle, J.L., Yang, R., 2020. Soil types influence the
627 characteristic of antibiotic resistance genes in greenhouse soil with long-term manure
628 application. *J. Hazard. Mater.* 392, 122334. <https://doi.org/10.1016/j.jhazmat.2020.122334>

629 Wei, Z., Shen, W., Feng, K., Feng, Y., He, Z., Li, Y., Jiang, C., Liu, S., Zhu, Y.-G., Deng, Y., 2022.
630 Organic fertilizer potentiates the transfer of typical antibiotic resistance gene among special
631 bacterial species. *J. Hazard. Mater.* 435, 128985. <https://doi.org/10.1016/j.jhazmat.2022.128985>

632 Wolters, B., Fornfeldt, E., Jechalke, S., Su, J.Q., Zhu, Y.G., Sørensen, S.J., Smalla, K., Jacquioid, S.,
633 2018. Soil amendment with sewage sludge affects soil prokaryotic community composition,
634 mobilome and resistome. *FEMS Microbiol. Ecol.* 95, 1–14.
635 <https://doi.org/10.1093/femsec/fiy193>

636 Wu, J., Wang, J., Li, Z., Guo, S., Li, K., Xu, P., Ok, Y.S., Jones, D.L., Zou, J., 2023. Antibiotics and

637 antibiotic resistance genes in agricultural soils: A systematic analysis. *Crit. Rev. Environ. Sci.*
638 *Technol.* 53, 847–864. <https://doi.org/10.1080/10643389.2022.2094693>

639 Xiang, Q., Zhu, D., Giles, M., Neilson, R., Yang, X.R., Qiao, M., Chen, Q.L., 2020. Agricultural
640 activities affect the pattern of the resistome within the phyllosphere microbiome in peri-urban
641 environments. *J. Hazard. Mater.* 382, 121068. <https://doi.org/10.1016/j.jhazmat.2019.121068>

642 Xie, W.Y., Yuan, S.T., Xu, M.G., Yang, X.P., Shen, Q.R., Zhang, W.W., Su, J.Q., Zhao, F.J., 2018.
643 Long-term effects of manure and chemical fertilizers on soil antibiotic resistome. *Soil Biol.*
644 *Biochem.* <https://doi.org/10.1016/j.soilbio.2018.04.009>

645 Yang, Y., Liu, G., Ye, C., Liu, W., 2019. Bacterial community and climate change implication
646 affected the diversity and abundance of antibiotic resistance genes in wetlands on the Qinghai-
647 Tibetan Plateau. *J. Hazard. Mater.* 361, 283–293. <https://doi.org/10.1016/j.jhazmat.2018.09.002>

648 Zhang, M.S., Li, W., Zhang, W.G., Li, Y.T., Li, J.Y., Gao, Y., 2021. Agricultural land-use change
649 exacerbates the dissemination of antibiotic resistance genes via surface runoffs in Lake Tai
650 Basin, China. *Ecotoxicol. Environ. Saf.* 220, 112328.
651 <https://doi.org/10.1016/j.ecoenv.2021.112328>

652 Zhang, W.P., Fornara, D., Yang, H., Yu, R.P., Callaway, R.M., Li, L., 2023. Plant litter strengthens
653 positive biodiversity–ecosystem functioning relationships over time. *Trends Ecol. Evol.* 1–12.
654 <https://doi.org/10.1016/j.tree.2022.12.008>

655 Zhang, X., Fang, Q., Zhang, T., Ma, W., Velthof, G.L., Hou, Y., Oenema, O., Zhang, F., 2020.
656 Benefits and trade-offs of replacing synthetic fertilizers by animal manures in crop production in
657 China: A meta-analysis. *Glob. Chang. Biol.* 26, 888–900. <https://doi.org/10.1111/gcb.14826>

658 Zhang, Y.J., Hu, H.W., Chen, Q.L., Singh, B.K., Yan, H., Chen, D., He, J.Z., 2019. Transfer of
659 antibiotic resistance from manure-amended soils to vegetable microbiomes. *Environ. Int.* 130,
660 104912. <https://doi.org/10.1016/j.envint.2019.104912>

661 Zhang, Z., Zhang, Q., Wang, T., Xu, N., Lu, T., Hong, W., Penuelas, J., Gillings, M., Wang, M., Gao,
662 W., Qian, H., 2022. Assessment of global health risk of antibiotic resistance genes. *Nat.*
663 *Commun.* 13. <https://doi.org/10.1038/s41467-022-29283-8>

664 Zhao, F., Chen, L., Yang, L., Sun, L., Li, S., Li, M., Feng, Q., 2020a. Effects of land use and rainfall
665 on sequestration of veterinary antibiotics in soils at the hillslope scale. *Environ. Pollut.* 260,
666 114112. <https://doi.org/10.1016/j.envpol.2020.114112>

667 Zhao, F., Chen, L., Yen, H., Li, G., Sun, L., Yang, L., 2020b. An innovative modeling approach of
668 linking land use patterns with soil antibiotic contamination in peri-urban areas. *Environ. Int.*
669 134, 105327. <https://doi.org/10.1016/j.envint.2019.105327>

670 Zheng, F., Zhou, G.W., Zhu, D., Neilson, R., Zhu, Y.G., Chen, B., Yang, X.R., 2022. Does Plant
671 Identity Affect the Dispersal of Resistomes Above and Below Ground? *Environ. Sci. Technol.*
672 56, 14904–14912. <https://doi.org/10.1021/acs.est.1c08733>

673 Zhou, S. yi dan, Zhu, D., Giles, M., Daniell, T., Neilson, R., Yang, X. ru, 2020. Does reduced usage
674 of antibiotics in livestock production mitigate the spread of antibiotic resistance in soil,
675 earthworm guts, and the phyllosphere? *Environ. Int.* 136, 105359.
676 <https://doi.org/10.1016/j.envint.2019.105359>

677 Zhou, X.Y., Li, H., Zhou, S.Y.D., Zhang, Y. Sen, Su, J.Q., 2023. City-scale distribution of airborne
678 antibiotic resistance genes. *Sci. Total Environ.* 856, 1–8.

679 <https://doi.org/10.1016/j.scitotenv.2022.159176>

680 Zhou, Y., Niu, L., Zhu, S., Lu, H., Liu, W., 2017. Occurrence, abundance, and distribution of
681 sulfonamide and tetracycline resistance genes in agricultural soils across China. *Sci. Total*
682 *Environ.* 599–600, 1977–1983. <https://doi.org/10.1016/j.scitotenv.2017.05.152>

683 Zhu, D., Chen, Q.L., Li, H., Yang, X.R., Christie, P., Ke, X., Zhu, Y.G., 2018. Land Use Influences
684 Antibiotic Resistance in the Microbiome of Soil Collembolans *Orchesellides sinensis*. *Environ.*
685 *Sci. Technol.* 52, 14088–14098. <https://doi.org/10.1021/acs.est.8b05116>

686 Zhu, D., Ding, J., Wang, Y.F., Zhu, Y.G., 2022. Effects of Trophic Level and Land Use on the
687 Variation of Animal Antibiotic Resistome in the Soil Food Web. *Environ. Sci. Technol.* 56,
688 14937–14947. <https://doi.org/10.1021/acs.est.2c00710>

689 Zhu, Y.G., Johnson, T.A., Su, J.Q., Qiao, M., Guo, G.X., Stedtfeld, R.D., Hashsham, S.A., Tiedje,
690 J.M., 2013. Diverse and abundant antibiotic resistance genes in Chinese swine farms. *Proc. Natl.*
691 *Acad. Sci. U. S. A.* 110, 3435–3440. <https://doi.org/10.1073/pnas.1222743110>

692