Avian-specific *Salmonella enterica* Serovar Gallinarum transition to endemicity is accompanied by localized resistome and mobilome interaction

Chenghao Jia[1], Chenghu Huang[1,2], Haiyang Zhou[1,2], Xiao Zhou[5], Zining Wang[1,2], Abubakar Siddique[1,2], Xiamei Kang[1], Qianzhe Cao[1], Yingying Huang[1,2], Fang He [1,6], Yan Li [1,2], Min Yue [1,2,3,4#]

[1] Department of Veterinary Medicine, Zhejiang University College of Animal Sciences, Hangzhou 310058, China

[2] Hainan Institute of Zhejiang University, Sanya 572000, China

[3] School of Life Science, Hangzhou Institute for Advanced Study, University of Chinese Academy of Sciences, Hangzhou 310024, China.

[4] State Key Laboratory for Diagnosis and Treatment of Infectious Diseases, National Clinical Research Center for Infectious Diseases, National Medical Center for Infectious Diseases, The First Affiliated Hospital, College of Medicine, Zhejiang University, Hangzhou 310003, China

[5] Ningbo Academy of Agricultural Sciences, Ningbo 315040, China

[6] ZJU-Xinchang Joint Innovation Centre (TianMu Laboratory), Gaochuang Hi-Tech Park, Zhejiang, China

Email address:

CJ: 22017118@zju.edu.cn

CH: 22217106@zju.edu.cn

ZH: zhouhaiyang@zju.edu.cn

XZ: xiaozhou_2023@163.com

ZW: znwang@zju.edu.cn

AS:abubakars@zju.edu.cn

XK: 12017032@zju.edu.cn

QC: 22317128@zju.edu.cn

YH: 22317127@zju.edu.cn

FH: hefangzj@zju.edu.cn

YL: yanli3@zju.edu.cn

Corresponding author:

Min Yue. Tel: +86 0571 88982832; E-mail: myue@zju.edu.cn

Abstract:

Bacterial regional demonstration after global dissemination is an essential pathway for selecting distinct finesses. However, the evolution of the resistome during the transition to endemicity remains unaddressed. Using the most comprehensive whole-genome sequencing dataset of *Salmonella enterica* serovar Gallinarum (*S*. Gallinarum) collected from 15 countries, including 45 newly recovered samples from two related local regions, we established the relationship among avian-specific pathogen genetic profiles and localization patterns. Initially, we revealed the international transmission and evolutionary history of *S*. Gallinarum to recent endemicity through phylogenetic analysis conducted using a spatiotemporal Bayesian framework. Our findings indicate that the independent acquisition of the resistome via the mobilome, primarily through plasmids, transposons, and prophages, shapes a unique antimicrobial resistance profile among different lineages. Notably, the mobilome-resistome combination among distinct lineages exhibits a geographical-specific manner, further supporting a localized endemic mobilome-driven process. Collectively, this study elucidates resistome adaptation in the endemic transition of an avian-specific pathogen, likely driven by the localized farming style, and provides valuable insights for targeted interventions.

Keywords:

Salmonella; Mobilome; Resistome; Whole genome sequencing; Horizontal gene transfer

Introduction

The acquisition of antimicrobial resistance (AMR) by pathogens is well-established as one of the most severe threats of the 21st century (1–3). It is estimated that over 10 million people could die per year due to AMR by 2050 (4). Antimicrobial resistance genes (ARGs), collectively known as the resistome, play a pivotal role in AMR development, progression, and amplification (5). Pathogens typically acquire resistome through mobilomes, like plasmids, via horizontal gene transfer (HGT) (6, 7). From an evolutionary perspective, the dynamic change in the resistome is crucial for enhancing pathogen fitness and enabling it to colonize new ecological niches (8). Key factors contributing to resistome alterations include disease burden, human activities, climate change, and geographical selection forces (9–13). Understanding the evolutionary trajectory of the AMR and related reservoirs is necessary, as disease management strategies will differ substantially (14). However, knowledge regarding the overall profile of pathogen regional adaptation related to the stepwise dynamics of the resistome remains limited.

Before recent advancements in whole-genome sequencing (WGS) technology, little was understood about resistome diversity that affects pathogen endemicity (15). Traditional typing methods generally exhibit low resolution, making monitoring and quantitatively comparing horizontal ARG transfer events challenging. The scarcity of appropriate models also poses a limitation. *Salmonella*, a pathogen with distinct geographical characteristics, has more than 90% of its serovars frequently categorized as geo-serotypes (16). Among the thousands of geo-serotypes, *Salmonella enterica* serovar Gallinarum (*S. Gallinarum*) is an avian-specific pathogen that causes severe mortality, with particularly detrimental effects on the poultry industry in low- and middle-income countries (17, 18). As once a globally prevalent pathogen in the

20th century, *S.* Gallinarum was listed by the World Organization for Animal

Health (WOAH) and gradually became an endemic pathogen with sporadic

outbreaks following the implementation of eradication programs in most

high-income countries, making it a perfect model for study (19, 20).

Nowadays, antimicrobial therapy remains a priority choice against *S.*

Gallinarum infections (21). The overuse or misuse of antimicrobials has led to

increased epidemiological escalation of *S.* Gallinarum, with a regionally higher

risk of AMR, especially for sulfonamides, penicillin, and tetracyclines (22, 23).

To fill the gaps in understanding the evolution of *S.* Gallinarum under

regional-associated AMR pressures and its adaptation to endemicity, we

collected the most comprehensive set *S.* Gallinarum isolates, consisting of 580

genomes, spanning the period from 1920 to 2023. Using such unique WGS

dataset, we investigated: 1) the population structure and potential geographical

transmission history of *S.* Gallinarum at a single base level; 2) the dynamic

resistome and mobilome changes that are associated with *S.* Gallinarum

transition to an endemic variant; and 3) horizontal resistome transfer frequency

and pattern between distinct regions.

Results

Global distribution of *S.* Gallinarum links with distinct sublineages

To understand the global geographic distribution and genetic relationships of *S.*

Gallinarum, we assembled the most comprehensive *S.* Gallinarum WGS

dataset (n=580), comprising 535 publicly available genomes and 45 newly

sequenced genomes (24)*.* The core-genome single nucleotide polymorphism

sites (cgSNPs) were obtained by using the fully sequenced genome of *S.*

Gallinarum R51 as a reference. Through hierarchical Bayesian analysis of

cgSNPs, it was confirmed that *S.* Gallinarum divides into three biovars: *S.*

Gallinarum biovar Pullorum (bvSP) (n=528/580, 91.03%), *S*. Gallinarum biovar

Gallinarum (bvSG) (n=50/580, 8.6%), and *S*. Gallinarum biovar Duisburg

(bvSD) (n=2/580, 0.34%). An association was identified between the biovar

type of *S*. Gallinarum and its global geographic distribution. Most bvSP isolates

were from Asia (436/528), with fewer occurrences in Europe (54/528), South

America (12/528), and North America (5/528). In bvSG, South America (18/50,

36%) is the primary source of isolates, followed by North America (9/50, 18%),

Europe (9/50, 18%), Africa (7/50, 14%), and Asia (3/50, 6%) (S1 Fig).

From a lineage perspective, bvSP can be further classified into five lineages: L1

(10/528), L2a (45/528), L2b (163/528), L3b (169/528), and L3c (141/528).

Importantly, L3a, previously considered distinct, has been revealed as a

sub-lineage of L3c. The predominant lineage types vary across different

continents. Regarding predominant bvSP sources, Asia stands out, with L3c,

L3b, and L2b being the main lineage types. On the other hand, L3b and L2a are

more widespread in Europe and America, respectively (Fig 1a). For the bvSP

strains from Asia included in our dataset, we found that all originated from

China. To further investigate the distribution of bvSP across different regions in

China, we categorized them into three distinct regions: southern, eastern, and

northern (Supplementary Table 3). Our findings indicate significant variations in

bvSP across various regions from geo-temporal aspects. Prevalence was

highest in the eastern region (233/436), with lower incidences in the northern

(137/436) and southern regions (65/436). Interestingly, the predominant bvSP

lineage type also varied. In eastern and southern China, L2b and L3c exhibited

as the predominant lineage types, while in northern China, L3b and L3c held

nearly equivalent positions (Fig 1c).

An analysis of the temporal prevalence of bvSP in China revealed a gradual

replacement pattern of the lineages. Before the 2000s, the predominant lineage

type was L3b. Then, from the 2000s to the 2010s, L3b and L1 were continually

replaced by L3c, with L2b becoming the predominant lineage after the 2010s

(Fig 1b). However, the replacement pattern varied among different regions.

Specifically, the same pattern was observed in eastern China, but for northern

and southern China, only replacements from L3b to L2b and L1 to L3c were

observed (S2 Fig). These findings might indicate accelerated localized

adaptation of bvSP.

Genomics portrait bvSP historic transmission

Considering the historical pandemic status of bvSP, we then investigated its

global geographic transmission routes to understand its evolution into an

endemic pathogen. L2b and L3b were identified as the dominant global

lineages due to their antimicrobial resistance risks, and potential

intercontinental transmission events. Other lineages were also analyzed using

a Bayesian model with a relaxed molecular clock to infer their historical

evolution (S3-S5 Figs). The temporal structure was verified (S6 Fig). Our

findings show that the origin of L3b in China can be traced back to as early as

1683 (95% CI: 1608 to 1839). In contrast, the earliest possible origin of L2b in

China dates back to 1880 (95% CI: 1838 to 1902) (Fig 2). Then, we specifically

estimated the time points for the first intercontinental transmission events for

the two major lineages. Our results indicate that L2b, likely underwent two

major intercontinental transmission events. The first occurred around 1893 (95%

CI: 1870 to 1918), with transmission from China to South America. The second

major transmission event occurred in 1923 (95% CI: 1907 to 1940), involving

the spread from South America to Europe. In contrast, the transmission pattern

of L3b, an *S.* Gallinarum lineage originating in China, appears relatively more

straightforward. It underwent only one intercontinental transmission event, from

China to Europe, likely around 1790 (95% CI: 1661 to 1890) (S7 Fig).

Endemic isolates further support localized dissemination by bvSP

To investigate the dissemination pattern of bvSP, we obtained forty-five newly

isolated bvSP from 734 samples (6.1% overall isolation rate) collected from

diseased chickens at two farms in Yueqing and Taishun, Zhejiang Province.

Those bvSP can be classified into two dominant sequence types (STs):

ST3717 and ST92. Isolates from Yueqing were all identified as ST3717, while

most Taishun isolates were characterized as ST92. Only the E404 isolate was

identified as ST2151 (Fig 3a). Of note, a lineage-preferential association was

also observed, with bvSP isolated from Taishun belonging to L3b and those

from Yueqing belonging to L2b. To further understand the genetic relationship

among isolates in the two regions, we calculated cgSNPs distances between

isolates. Isolates with cgSNP distances less than five were determined as the

threshold to be involved in the same transmission event. Interestingly, genetic

clustering revealed that isolates from the same farm exhibited a significantly

close genetic correlation (Fig 3b). No historical transmission events of bvSP

were found between Taishun and Yueqing.

Other genetic characteristics also exhibited a similar pattern. The ARG and

plasmid profile disclosed that *IncFII(S)* and *ColpVC* plasmids exhibited the

highest carriage percentages (100%) among the 45 bvSP isolates. The *IncX1*

plasmid was exclusively identified in the E404 isolate. Furthermore, E404

harboured two distinctive resistance genes, bla_{TEM-1B} and *sul2*, in contrast to the

absence of ARGs in the remaining bvSP isolates. We also estimated the

invasiveness index to assess the invasiveness of isolated bvSP and showed

that bvSP from the same farm had comparable invasive abilities. However,

isolates from Taishun demonstrated a higher invasive ability than those from

Yueqing (Fig 3c, Supplementary Table 6).

Furthermore, we simulated potential transmission events between the bvSP

strains isolated from Zhejiang Province (n=95) and those from China with

available provincial information (n=435) using five SNPs as the threshold. As a

result, the transmission events showed a strong geographic-preferential

distribution. We identified a total of 91 potential transmission events, all of

which occurred exclusively within Zhejiang Province. No inter-provincial

transmission events were detected, supporting the statement that bvSP in

China is a highly localized pathogen (S8 Fig, Supplementary Table 8).

Mobilome and resistome distinct lineage-preferential association in a

regional manner

The mobilome drives partitions of the resistome and plays a pivotal role in

shaping the ecological niches and local adaptation of bacteria. Therefore, a

quantitative evaluation of the resistome and mobilome is necessary to enhance

the understanding of the localized distribution of bvSP. For *S.* Gallinarum, a

total of 13 classes of ARGs, which can be further classified into six categories,

were identified. The results revealed that bvSP exhibited a significantly greater

resistome than bvSG and bvSD (Fig 4a). Among them, *sul2* (196/528, 37.1%)

has the highest prevalence, followed by $bla_{\text{TEM-1B}}$ (183/528, 34.7%) and *tet(A)*

(104/528, 19.7%). Moreover, the resistome also demonstrated a

lineage-preferential association. We observed that L3b is more inclined to carry

$bla_{\text{TEM-1B}}$ and sul2 than other lineages, whereas tet(A) exclusively exists within

L3 (Supplementary Table 4). The diversity of resistome carried by L3 may

elevate the risk of multi-drug resistance (MDR). Notably, AMR risks vary among

different regions of China, with the highest risk observed in southern China,

followed by the northern and eastern regions (Fig 4c). Interestingly, *tet(A)*,

*bla*TEM-1B, and *sul2* were predominant resistome types prevalent across all

regions of China. Meanwhile, *aadA5* was typically more commonly found in

southern China (Fig 4b).

Four categories of the mobilome—prophage, plasmid, transposon, and

integron—were characterized, revealing strong lineage-specific patterns (S9a

Fig, Supplementary Table 5). In silico analyses revealed prophages emerge as

the most diverse mobilome type, with *Entero_mEp237, Escher_500465_1, and*

Klebsi_4LV2017 being more likely associated with L1, L2a, and L2b,

respectively, while the carrier percentage of *Escher_pro483* is significantly

higher among L3c. Interestingly, *Salmon_SJ46, Gifsy_2, Escher_500465_2,*

and *Shigel_SfIV* were consistently observed across all bvSP lineages and were

found to be duplicated on the chromosome. For plasmids, *IncFII(S)* was

predominantly carried across all bvSP lineages (520/528, 98.5%) (S9b Fig).

Notably, *ColpVC* had the lowest carriage percentage among L1 (2/10, 20%),

whereas *IncX1* was explicitly associated with L3c (140/141, 99.3%).

The diversity of transposons and integrons is comparatively lower in bvSP.

Specifically, we identified seven types of transposons and four types of

integrons. Transposons were found to be more abundant, comprising a total of

255 in 528 bvSP. Among them, *Tn801* (185/528, 35%) had the highest carriage

rate, followed by *Tn1721* (48/528, 9.1%). Interestingly, L3c is more likely to

carry *Tn801* and *Tn1721* transposons than other lineages. In contrast, a total of

42 integrons were identified among the 528 bvSP, all of which belong to class I:

In498 (n=22), *In1440* (n=14), *In473* (n=4), and *In282* (n=2). The integrons also

exhibited a lineage-preferential distribution, predominantly carried by L3b.

From a geographic distribution perspective, bvSP from northern and southern
China carried the most extensive mobilome, followed by those from eastern
China. Interestingly, bvSP isolated from northern China typically have more
diverse mobilome types (S10 Fig).

Plasmid and transposon guide resistome geo-temporal dissemination
Variations in regional antimicrobial use may result in uneven pressure for
selecting AMR. The mobilome is considered the primary reservoir for spreading
resistome, and a consistent trend between the resistome and the mobilome has
been observed across different lineages, from L1-L3c. We observed an overall
gradual rise in the resistome quantity carried by bvSP across various lineages,
correlating with the total mobilome content (S11 Fig). Furthermore, we
investigated the interplay between particular mobile elements and resistome
types in bvSP.

In bvSP, the predominant resistome is mainly associated with certain types of
mobilome (Fig 5a). Specifically, regarding bla_{TEM_1B}, $tet(A)$, and $sul1$, which are
highly prevalent among bvSP, we found most bla_{TEM_1B} were carried by the
transposon $Tn801$ and $Tn1721$, the plasmid $IncX1$, and the prophage SJ46.
Additionally, $tet(A)$ was primarily associated with the transposon $Tn1721$ and
the plasmid IncX1. Integrons also facilitated the dissemination of the resistome.
For instance, $In498$ and $In1440$ facilitate the dissemination of $sul1$, leading to
quinolone resistance. Our results indicate that plasmids and transposons are
the predominant reservoirs for the resistome in bvSP, with prophages and
integrons following closely behind (Fig 5b). Interestingly, we observed that the
primary reservoir for the resistome shows regional variations. In the southern
and northern regions of China, the primary reservoirs for bla_{TEM_1B} are $Tn801$,
$Tn1721$, $IncX1$, and prophage SJ46. However, $Tn1721$ was less frequently

found in bvSP, which carries bla_{TEM-1B} from the eastern region. Similarly, the

reservoir for *tet(A)* exhibits greater diversity in the east region but less diversity

in the southern region (Fig 5c-e).

Horizontal transfer of resistome occurs widely in localized bvSP

Horizontal transfer of the resistome facilitates the acquisition of AMR among

bacteria, which may record the distinct acquisition event in the bacterial

genome. To compare these events in a geographic manner, we further

investigated the HGT frequency of each ARG carried by bvSP isolated from

China and explored the HGT frequency of resistome between three defined

regions. Potentially horizontally transferred ARGs were defined as those with

perfect identity (100% identity and 100% coverage) and were located on MGEs

across different strains (Fig. 6a). We first categorized a total of 621 ARGs

carried by 436 bvSP isolates collected in China and found that 415 ARGs were

located on MGEs (Supplementary Table 9). After excluding the ARGs not

associated with MGEs, our findings reveal that horizontal gene transfer of

ARGs is widespread among Chinese bvSP isolates, with an overall transfer

rate of 92%. Specifically, 50% of the ARGs exhibited an HGT frequency of

100%, indicating that these ARGs might underwent extensive frequent

horizontal transfer events (Fig. 6b). It is noteworthy that certain resistance

genes, such as *tet(A)*, *aph(3'')-Ib*, and *aph(6)-Id*, appear to be less susceptible

to horizontal transfer.

However, different regions generally exhibited a considerable difference in

resistome HGT frequency. Overall, bvSP from the southern areas in China

showed the highest HGT frequency (HGT frequency=95%). The HGT

frequencies for bvSP within the eastern and northern regions of China are

lower, at 92% and 91%, respectively (Fig. 6c). For specific ARG type, we

found *tet(A)* is more prone to horizontal transfer in the southern region, and this

proportion was considerably lower in the eastern region. Interestingly, certain

ARGs such as *aph(6)-Id*, undergo horizontal transfer only within the eastern

and northern regions of China (Fig. 6d). Notably, as a localized transmission

pathogen, resistome carried by bvSP exhibited a dynamic potential among

inter-regional and local demographic transmission, especially from northern

region to southern region (HGT frequency=93%) (Fig. 6e, Supplementary

Table 7).

Discussion

Over the past decades, the emergence of AMR has garnered significant

attention from both public health agencies and the poultry industry (25).

Region-associated pressures shaped the dissemination patterns of bacterial

pathogens, transitioning from pandemic to endemic, and led to dynamic

changes in the resistome. Here, utilizing *S.* Gallinarum as a prototype for an

endemic pathogen that once spread globally, and assembling the most

comprehensive *S.* Gallinarum WGS database, we have discovered that

region-associated pressures can influence the AMR risks of *S.* Gallinarum

within lineage- and region-specific patterns. These pressures might also

influence lineage-distinct evolutionary structures and transmission histories of

S. Gallinarum.

The biovar types of *S.* Gallinarum have been well-defined as bvSP, bvSG, and

bvSD historically (26). Among these, bvSP can be further subdivided into five

lineages (L1, L2a, L2b, L3b, and L3c) using hierarchical Bayesian analysis.

Different sublineages exhibited preferential geographic distribution, with L2b

and L3b of bvSP being predominant global lineage types with a high risk of

AMR. The historical geographical transmission was verified using a

spatiotemporal Bayesian framework. The result shows that L3b was initially

spread from China to Europe in the 18th-19th century, which may be associated

with the European hen fever event in the mid-19th century (27). L2b, on the

other hand, appears to have spread to Europe via South America, potentially

contributing to the prevalence of bvSP in the United States. Considering the

economic losses caused by bvSP, the United States, Europe, and other

industrialized countries implemented eradication programs in the mid-19th

century, thus eliminating the risk of pullorum disease. Implementing similar

measures is challenging due to China's vast geographical area and various

economic factors, resulting in bvSP becoming an endemic pathogen, mainly in

China.

Additionally, forty-five bvSP was isolated from Taishun and Yueqing, to

explore the transmission pattern of bvSP in China. The cgSNP distances

between bvSP isolates from Taishun and Yueqing, showed significant diversity,

with cgSNP distances of less than 100 isolate pairs observed exclusively

within a single region. The transmission history of 45 global bvSP isolates

revealed that most transmission events occurred within a single region.

Additionally, the number of transmission events decreased rapidly with

increasing distance, confirming that bvSP in China belongs to a highly

localized pathogen.

Humanity appeared powerless against epidemics until the advent of

antimicrobials, which are still the primary agents used to combat *S.* Gallinarum

infection (28, 29). As a manifestation of localization-related forces, the

emergence of AMR allows pathogens to acquire additional fitness benefits (30).

Our findings further indicate that the mobilome facilitates the acquisition of
AMR among bvSP, mainly through plasmids and transposons, such as *IncX1*,
IncQ1, and *IncN* type plasmids, as well as *Tn801*, *Tn6205*, and *Tn721* type
transposons. A survey conducted in China between 1962 and 2010
documented the association of class 1 integrons with AMR among bvSP (31).
Our study further demonstrates that *In498*, *In1440*, *In473*, and *In282*, all
belonging to class 1 integrons, play a significant role in carrying ARGs
associated with sulfonamides, and trimethoprim. However, compared to other
Salmonella serovars (32–34), the level of resistome in bvSP remains lower.

Considering lineage prospects, recently emerged lineages present a higher
risk of AMR, with a lineage-specific distribution of resistome. In L2b and L3c,
the most prevalent groups of ARGs include *bla*$_{TEM-1b}$ and *sul2*. L3c, which
exhibits the highest resistome richness with *sul1*, *tet(A)*, *aadA5*, and *dfrA17*,
has a higher likelihood of MDR. However, the transmission routes of resistome
showed a high degree of consistency across continents. Globally, a small
proportion of resistome carriers belonging to L2b and L3b were observed in
the Americas and Europe, while isolates from Asia exhibited a notable rise in
resistome carriage. Therefore, we speculate that the mobilome in bvSP may
have been acquired locally in Asia, possibly due to increased antimicrobial
usage and an intensive industrialized poultry farming pattern within the
regional poultry industry (35). The L3c, being a local Asian lineage and more
prone to acquiring additional resistome locally, provided further support.

In China, the endemic region of Asia, an unexpected rising trend has been
observed from the 1970s to the 2020s. However, the risk of AMR varies
among distinct regions (36–39). Overall, the average number of ARGs carried
by bvSP isolated from eastern China was the lowest, while it was higher

among bvSP isolated from northern and southern China. This variation may be influenced by differences in the economy, climate, or farming practices among different regions. Interestingly, the widespread horizontal transfer of the resistome across regions seems to indicate that the resistome has a different mode of transmission compared to its pathogenic bacteria. Specifically, as a localized pathogen, our results show a low frequency of cross-regional transmission of bvSP (as shown in S5 Fig), but up to 90% of ARGs are transmitted cross-regionally. This risk has not been previously observed in *S.* Gallinarum, suggesting that potential intermediate hosts may play a key role.

In summary, the findings of this study highlight that *S.* Gallinarum remains a significant concern in developing countries, particularly in China. Compared to other regions, *S.* Gallinarum in China poses a notably higher risk of AMR, necessitating the development of additional therapies, i.e. vaccine, probiotics, bacteriophage therapy in response to the government's policy aimed at reducing antimicrobial use (40). Furthermore, given the dynamic nature of *S.* Gallinarum risks across different regions, it is crucial to prioritize continuous monitoring in key areas, particularly in China's southern regions where the extensive poultry farming is located. Lastly, from a One-Health perspective, controlling AMR in *S.* Gallinarum should not solely focus on local farming environments, with improved overall welfare on poultry and farming style. The breeding pyramid of industrialized poultry production should be targeted on the top, with enhanced and accurate detection techniques (41). More importantly, comprehensive efforts should be made to reduce antimicrobial usage overall and mitigate potential AMR transmission from environmental sources or other hosts (42–44).

However, the current study has some limitations. Firstly, despite assembling the most comprehensive WGS database for *S.* Gallinarum from public and laboratory sources, there are still biases in the examined collection. The majority (438/580) of *S.* Gallinarum samples were collected from China, possibly since the WGS is a technology that only became widely available in the 21st century. This makes it impractical to sequence it on a large scale in the 20th century, when *S.* Gallinarum caused a global pandemic. So, we suspect that human intervention in the development of this epidemic is the main driving force behind the fact that most of the strains in the data set originated in China. In our future work, we aim to actively gather more data to minimize potential biases within our dataset, thereby improving the robustness and generalizability of our findings. Secondly, in silico analysis relies heavily on sophisticated, continuously updated databases. Although we utilized the most up-to-date databases, the exact prevalence of mobilome genetic elements (MGEs) and ARGs may be underestimated. Nevertheless, by using the most extensive global WGS data from *S.* Gallinarum, we elucidate a dynamic resistome evolution framework in a single bacterial pathogen from pandemic to endemic, which is hallmarked by a gradual increase in antimicrobial resistance risks and a stepwise resistome adaptation. The insights gained in this study will be helpful in further prevention and surveillance of the AMR risks of localized pathogens.

Material and Methods

Ethics statements

The protocols used for animal studies were approved by the Committee of the Laboratory Animal Center of Zhejiang University (ZJU20190094; ZJU20220295).

Bacterial isolates

All 734 samples of dead chicken embryos aged 19 to 20 days were collected from Taishun and Yueqing in Zhejiang Province, China. After a thorough autopsy, the liver, intestines, and spleen were extracted and added separately into 2 mL centrifuge tubes containing 1 mL PBS. The organs were then homogenized by grinding. In the initial enrichment phase, we utilized Buffered Peptone Water (BPW, Haibo Biotechnology Co, Qingdao, China), employing a 1:9 dilution ratio (sample in PBS: BPW). Subsequently, the composite was incubated at 37°C for 16-18 hours in a rotary incubator set to 180 rpm. For selective enrichment, Tetrathionate Broth Base (TTB, Land Bridge Biotechnology Co, Beijing, China), fortified with iodine solution and brilliant green solution (both from Land Bridge Biotechnology Co, Beijing, China), was employed at a ratio of 1:10 (sample in BPW: TTB). The mixture was subjected to incubation at 42°C for a duration ranging between 22 and 26 hours within a rotary incubator set at 180 rpm. Isolated *Salmonella* colonies from positive samples were obtained by sub-culturing selectively enriched samples on Xylose Lysine Deoxycholate (XLD, Land Bridge Technology Co, Beijing, China) agar, followed by an 18–22-hour incubation at 37°C. Typical and pure colonies were selected after sub-culturing on XLD agar and transferred into Luria-Bertani (LB) broth. Finally, the transferred bacterial culture underwent an additional incubation for 18-22 hours at 37°C in a rotary incubator operating at 180 rpm.

DNA extraction and genomic assembly

The DNA of forty-five isolates was extracted using the Vazyme Fastpure® Bacteria DNA Isolation Mini Kit (Vazyme Biotech Co., Ltd.), which was then quantified using the NanoDrop1000 system (Thermo Fisher Scientific, USA).

Subsequently, DNA libraries were constructed and subjected to sequencing

using the Illumina Novaseq 6000 platform (Beijing Novogene Co., Ltd).

Assembly of genome sequences was performed by SPAdes (45) v3.12.0 with

default parameters.

Global dataset assembly

To provide global context, we expanded the dataset by including 540 additional

WGS genome data, consisting of bvSP (n=483), bvSG (n=50), bvSD (n=2), and

Salmonella enterica serovar Enteritidis (n=5). The inclusion of five strains of

Salmonella enterica serovar Enteritidis enhanced evolutionary analyses.

Notably, in this collection of 540 genomic data from public databases, 325

sequences were previously preserved in our laboratory.

All WGS data passed strict quality control according to criteria set by the

European Reference Laboratory (46). Genomic data exceeding 500 contigs

and having an N50 of less than 30,000 were excluded. Lastly, the bacterial

species associated with each genomic data set was confirmed by utilizing

KmerFinder (47) v3.2. Finally, the assembled database includes a total of 585

high-quality sequences.

Genotyping and phylogenetic analyses

Snippy v.4.4.5 was used to identify cgSNPs, with the complete genome of R51

used as a reference. The optimal model (TVM+F) was determined and

employed to construct the maximum-likelihood phylogenetic tree using

IQ-TREE v.1.6 (48). The final phylogenetic tree was annotated and visualized

using iTOL v.6.0 (49).

The global assembly of *S.* Gallinarum underwent a population structure

analysis using the RhierBAPS v1.1.4 (50), based on the Hierarchical Bayesian

clustering (BHC) algorithm and employing cgSNPs as input. The calculation of

population structure utilized R package v.4.3.1, supplemented by the R

packages "phytools" v.2.0.3, "ape" v.5.7.1, and "rhierbaps" v.1.1.4. Default

parameters were employed for all population structure analysis.

Temporal and phylogeographical analysis

To examine the emergence and geographical transfers, we first utilized

Gubbins v.2.3 (51) to eliminate the recombination regions for each lineags (S12

Fig). Then, TreeTime (52) was used to assess the temporal structure. This was

accomplished through regression analysis of the root-to-tip branch distances

within the maximum likelihood tree, considering the sampling date as a variable.

Then, the strains within the dataset underwent phylogeographical

reconstruction via Bayesian Evolutionary Analysis of Sampled Trees (BEAST)

(53) v.2.5. To determine the optimal model for running BEAST, we tested a total

of six combinations in the initial phase of our study. These combinations

included the strict clock, relaxed lognormal clock, and three population models

(Bayesian SkyGrid, Bayesian Skyline, and Constant Size). Before conducting

the full BEAST analysis, we evaluated each combination using a Markov Chain

Monte Carlo (MCMC) analysis with a total chain length of 100 million and

sampling every 10,000 iterations. We then summarized the results using

NSLogAnalyser and determined the optimal model based on the marginal

likelihood value for each combination. The results indicated that the model

incorporating the Bayesian Skyline, and the relaxed lognormal clock yielded the

highest marginal likelihood value in our sample. Consequently, we proceeded

to perform a time-calibrated Bayesian phylogenetic inference analysis for each

lineage. The following settings were configured: the "GTR" substitution model,

"4 gamma categories", the "Relaxed Clock Log Normal" model, the "Coalescent

Bayesian Skyline" tree prior. Convergence was assessed using Tracer, with all

parameter effective sampling sizes (ESS) exceeding 200. Maximum clade

credibility trees were generated using TreeAnnotator v.2.6.7. Finally, key

divergence time points (with 95% credible intervals) were estimated and the

phylogenetic tree was visualized using Figtree v.1.4.3.

SNP distance-based *S*. Gallinarum geographical tracing

To understand potential transmission events, we calculated the cgSNPs

distance between distinct bacterial strains employing the SNP-dists. We

estimated the overall evolutionary rate of the *S*. Gallinarum using BEAST. We

applied the methodology described previously (54). The numbers of SNPs per

year were determined by multiplying the evolutionary rates estimated with

BEAST by the number of core SNP sites identified in the alignments. We

hypothesize that a slower evolutionary rate in bacteria typically requires a lower

SNP threshold when tracing transmission events using SNP distance analysis.

Previous research found an average evolutionary rate of 1.97 SNPs per year

(95% HPD, 0.48 to 4.61) across 22 different *Salmonella* serotypes. Our

updated BEAST estimation for the evolutionary rate of *S*. Gallinarum suggests

it is approximately 0.74 SNPs per year (95% HPD, 0.42 to 1.06). Based on

these findings, as well as our previous experience with similar studies (55), we

set a threshold of 5 SNPs. Additionally, the geographic location was considered

to enhance the precision of speculation regarding potential transmission sites

and transmission direction.

MGE detecting

Four types of MGEs were detected: plasmids, transposons, integrons, and

prophages. Plasmids were detected using Abricate v.1.0.1 with the

PlasmidFinder (56) database. Only plasmids with a similarity of more than 90%

and a coverage over 95% were identified. BacAnt (57) was used to detect

integrons and transposons in the genomes. Only integrons or transposons with

a similarity of more than 60% and a coverage greater than 60% will be identified.

The prophages were detected using the Phaster pipeline. The genomic data

were split into two temporary databases based on the number of contigs, one

dataset containing single contig files and the other containing multiple contig

files. The two databases were imported into the Phaster pipeline separately

with default parameters.

Determination of the potential MGE carried ARG and horizontal ARG

transfer frequency

We have devised a pipeline to discern the coexistence of MGEs and ARGs. For

plasmids and prophages, only ARGs situated within MGE regions are deemed

MGE-carried. Regarding transposons and integrons, the search region was

expanded to 5 kilobases upstream and downstream of ARGs, accounting for

potential splicing errors. In this study, HGT events involving ARGs were defined

as instances where ARGs exhibited perfect identity (100% coverage and 100%

identity) and were located on MGEs across different strains. In a previous study

(58), horizontal transfer of ARGs was identified based on gene regions with 100%

coverage and over 99% identity. We believe that these stricter criteria in our

pipeline more accurately reflect the transfer of ARGs within the *S.* Gallinarum

population. HGT frequency was utilized to evaluate the extent of horizontal

transfer of ARGs as follows:

$$HGT\ frequency = \frac{\text{Horizontal ARG transfer Count}}{\text{Total dissemination count}}$$

The pipeline developed in this study accepts Resfinder or Resistance Gene

Identifier (RGI) results as input. Specific code and examples are uploaded to:

https://github.com/tjiaa/Cal_HGT_Frequency/tree/main.

Genomic data analysis

The serovars of all WGS genome data were confirmed through SISTR (59)

v.1.1 and Seqsero2 (60). Multilocus Sequence Typing was carried out using

MLST v.2.22 with the senterica_achtman_2 scheme. ARGs were detected by

ResFinder (61). Both similarity and coverage were set to a minimum value of 90.

The invasiveness index was calculated using methods previously reported (62).

Specifically, *Salmonella*'s ability to cause intestinal or extraintestinal infections

in hosts is related to the degree of genome degradation. We evaluated the

potential for extraintestinal infection by 45 newly isolated *S*. Gallinarum strains

(L2b and L3b) using a model that quantitatively assesses genome degradation.

We analyzed each sample using the 196 top predictor genes for measuring the

invasiveness of *S*. Gallinarum, employing a machine-learning approach that

utilizes a random forest classifier and delta-bitscore functional variant-calling.

This method evaluated the invasiveness of *S*. Gallinarum towards the host, and

the distribution of invasiveness index values for each region was statistically

tested using unpaired *t*-test. The code used for calculating the invasiveness

index is available at https://github.com/Gardner-BinfLab/invasive_salmonella.

Data Availability

For the newly isolated 45 strains of *Salmonella* Gallinarum, genome data have

been deposited in NCBI Sequence Read Archive (SRA) database. The "SRA

Accession" for each strain are listed in Supplementary Table 1. Additionally, the

genome data for the 540 publicly available genomes have been uploaded to

Figureshare and can be accessed at:

https://figshare.com/articles/journal_contribution/Genomic_data/26028592.

Code availability

The open-source software used in this study includes:

BacAnt (https://github.com/xthua/bacant),

Abricate (https://github.com/tseemann/abricate),

Snippy (https://github.com/tseemann/snippy),

SISTR (https://github.com/phac-nml/sistr_cmd),

MLST (https://cge.food.dtu.dk/services/MLST/)

SeqSero2 (https://github.com/denglab/SeqSero2),

IQtree (https://github.com/iqtree/iqtree2).

BEAST (https://www.beast2.org)

SNP-dists (https://github.com/tseemann/snp-dists)

Fast BAPS (https://github.com/gtonkinhill/fastbaps)

R (https://www.r-project.org)

Phaster (http://phaster.ca)

KmerFinder (https://bitbucket.org/genomicepidemiology/kmerfinder)

Treetime (https://github.com/neherlab/treetime)

Author contributions

M.Y., C.J. designed the study; finalized the manuscript;

C.J. collected isolates and clinical data, integrated the data, performed

metadata and genomic analysis, prepared tables and figures; and wrote the

draft manuscript;

C.H. performed metadata and genomic analysis; interpreted the data; finalized

the manuscript;

X.Z., X.K., Q.C., Y.H. and Z.W. contributed to the data analysis;

C.H., M.Y. prepared tables and figures;

M.Y., C.H., A.S., Y.L., F.H., and C.J. finalized the manuscript;

M.Y. managed the project;

All the authors contributed to the article and approved the submitted version.

Declaration of Interests

650 None.

Acknowledgements

This work was supported by the National Program on the Key Research Project
of China (2022YFC2604201), the Zhejiang Provincial Natural Science
Foundation of China (LZ24C180002; LR19C180001), the Hainan Provincial
Joint Project of Sanya Yazhou Bay Science and Technology City
(2021JJLH0083), the Zhejiang Provincial Key R&D Program of China
(2023C03045, 2022C02024) and the Open Project Program of the Jiangsu Key
Laboratory of Zoonosis (R1902). We also thank Dr. Peide Li for providing the
valuable resource for this investigation.

Reference

1. Li Y, Ed-Dra A, Tang B, Kang X, Müller A, Kehrenberg C, Jia C, Pan H, Yang H, Yue M. 2022. Higher tolerance of predominant Salmonella serovars circulating in the antibiotic-free feed farms to environmental stresses. Journal of Hazardous Materials 438:129476.

2. Hou J, Long X, Wang X, Li L, Mao D, Luo Y, Ren H. 2023. Global trend of antimicrobial resistance in common bacterial pathogens in response to antibiotic consumption. Journal of Hazardous Materials 442:130042.

3. Lee K, Raguideau S, Sirén K, Asnicar F, Cumbo F, Hildebrand F, Segata N, Cha C-J, Quince C. 2023. Population-level impacts of antibiotic usage on the human gut microbiome. Nat Commun 14:1191.

4. Rappuoli R, Bloom DE, Black S. 2017. Deploy vaccines to fight superbugs. Nature 552:165–167.

5. Larsson DGJ, Flach C-F. 2022. Antibiotic resistance in the environment. Nat Rev Microbiol 20:257–269.

6. Tóth AG, Judge MF, Nagy SÁ, Papp M, Solymosi N. 2023. A survey on

antimicrobial resistance genes of frequently used probiotic bacteria, 1901

to 2022. Eurosurveillance 28.

7. Jia C, Wang Z, Huang C, Teng L, Zhou H, An H, Liao S, Liu Y, Huang L,

Tang B, Yue M. 2023. Mobilome-driven partitions of the resistome in

Salmonella. mSystems 8:e00883-23.

8. Carr VR, Witherden EA, Lee S, Shoaie S, Mullany P, Proctor GB,

Gomez-Cabrero D, Moyes DL. 2020. Abundance and diversity of

resistomes differ between healthy human oral cavities and gut. Nat

Commun 11:693.

9. Wang Z, Huang C, Liu Y, Chen J, Yin R, Jia C, Kang X, Zhou X, Liao S, Jin

X, Feng M, Jiang Z, Song Y, Zhou H, Yao Y, Teng L, Wang B, Li Y, Yue M.

2024. Salmonellosis outbreak archive in China: data collection and

assembly. Sci Data 11:244.

10. Jia C, Cao Q, Wang Z, Van Den Dool A, Yue M. 2024. Climate change

affects the spread of typhoid pathogens. Microbial Biotechnology

17:e14417.

11. Petersen E, Petrosillo N, Koopmans M, Beeching N, Di Caro A,

Gkrania-Klotsas E, Kantele A, Kohlmann R, Koopmans M, Lim P-L,

Markotic A, López-Vélez R, Poirel L, Rossen JWA, Stienstra Y, Storgaard

M. 2018. Emerging infections—an increasingly important topic: review by

the Emerging Infections Task Force. Clinical Microbiology and Infection

24:369–375.

12. Wang Z, Zhou H, Liu Y, Huang C, Chen J, Siddique A, Yin R, Jia C, Li Y,

Zhao G, Yue M. 2024. Nationwide trends and features of human

salmonellosis outbreaks in China. Emerging Microbes & Infections

13:2372364.

13. Ke Y, Teng L, Zhu Z, Lu W, Liu W, Zhou H, Yu Q, Ye L, Zhu P, Zhao G,

Yue M. 2024. Genomic investigation and nationwide tracking of pediatric

invasive nontyphoidal *Salmonella* in China. mLife 3:156–160.

14. Baquero F, Martínez JL, F. Lanza V, Rodríguez-Beltrán J, Galán JC, San

Millán A, Cantón R, Coque TM. 2021. Evolutionary Pathways and

Trajectories in Antibiotic Resistance. Clin Microbiol Rev 34:e00050-19.

15. Didelot X, Bowden R, Wilson DJ, Peto TEA, Crook DW. 2012.

Transforming clinical microbiology with bacterial genome sequencing. Nat

Rev Genet 13:601–612.

16. Gossner CM, Le Hello S, De Jong B, Rolfhamre P, Faensen D, Weill F-X,

Johan Giesecke. 2016. Around the World in 1,475 *Salmonella*

Geo-serotypes. Emerg Infect Dis 22:1298–1302.

17. Kang X, Zhou X, Tang Y, Jiang Z, Chen J, Mohsin M, Yue M. 2022.

Characterization of Two-Component System CitB Family in Salmonella

Pullorum. IJMS 23:10201.

18. Zhou X, Kang X, Zhou K, Yue M. 2022. A global dataset for prevalence of

Salmonella Gallinarum between 1945 and 2021. Sci Data 9:495.

19. Zhou X, Kang X, Chen J, Song Y, Jia C, Teng L, Tang Y, Jiang Z, Peng X,

Tao X, Xu Y, Huang L, Xu X, Xu Y, Zhang T, Yu S, Gong J, Wang S, Liu Y,

Zhu G, Kehrenberg C, Weill F-X, Barrow P, Li Y, Zhao G, Yue M. 2023.

Genome degradation promotes *Salmonella* pathoadaptation by

remodeling fimbriae-mediated proinflammatory response. National

Science Review 10:nwad228.

20. De Carli S, Gräf T, Kipper D, Lehmann FKM, Zanetti N, Siqueira FM,

Cibulski S, Fonseca ASK, Ikuta N, Lunge VR. 2017. Molecular and

phylogenetic analyses of Salmonella Gallinarum trace the origin and

diversification of recent outbreaks of fowl typhoid in poultry farms.

Veterinary Microbiology 212:80–86.

21.   Barrow PA, Neto OCF. 2011. Pullorum disease and fowl typhoid—new

thoughts on old diseases: a review. Avian Pathology 40:1–13.

22.   Nhung NT, Chansiripornchai N, Carrique-Mas JJ. 2017. Antimicrobial

Resistance in Bacterial Poultry Pathogens: A Review. Front Vet Sci 4:126.

23.   Penha Filho RAC, Ferreira JC, Kanashiro AMI, Darini AL da C, Berchieri

Junior A. 2016. Antimicrobial susceptibility of Salmonella Gallinarum and

Salmonella Pullorum isolated from ill poultry in Brazil. Ciência Rural

46:513–518.

24.   Jia C, Huang L, Zhou H, Cao Q, Wang Z, He F, Li Y, Yue M. 2024. A global

genome dataset for Salmonella Gallinarum recovered between 1920 and

2024. Sci Data 11:1094.

25.   Huang L, Zhou H, Chen J, Jia C, Siddique A, Wu B, Wang H, Tang B, He F,

Zhao G, Yue M. 2024. Impact of COVID-19-related nonpharmaceutical

interventions on diarrheal diseases and zoonotic Salmonella. hLife

S294992832400021X.

26.   Kisiela D, Kuczkowski M, Kiczak L, Wieliczko A, Ugorski M. 2005.

Differentiation of *Salmonella* Gallinarum biovar Gallinarum from

Salmonella Gallinarum biovar Pullorum by PCR-RFLP of the *fimH* gene.

Journal of Veterinary Medicine, Series B 52:214–218.

27. Burnham GP. 1855. The history of the hen fever: a humorous record.

James French.

28. Worboys M. 2000. Spreading germs: disease theories and medical

practice in Britain, 1865-1900. Cambridge University Press.

29. Porter R, Crosby AW. 1998. The Greatest Benefit to Mankind: A Medical

History of Humanity from Antiquity to the Present. Nature 391:241–241.

30. Helekal D, Keeling M, Grad YH, Didelot X. 2023. Estimating the fitness

cost and benefit of antimicrobial resistance from pathogen genomic data. J

R Soc Interface 20:20230074.

31. Gong J, Xu M, Zhu C, Miao J, Liu X, Xu B, Zhang J, Yu Y, Jia X. 2013.

Antimicrobial resistance, presence of integrons and biofilm formation of

Salmonella Pullorum isolates from eastern China (1962–2010). Avian

Pathology 42:290–294.

32. Li Y, Teng L, Xu X, Li X, Peng X, Zhou X, Du J, Tang Y, Jiang Z, Wang Z,

Jia C, Müller A, Kehrenberg C, Wang H, Wu B, Weill F, Yue M. 2022. A

nontyphoidal *Salmonella* serovar domestication accompanying enhanced

niche adaptation. EMBO Mol Med 14:e16366.

33. Teng L, Liao S, Zhou X, Jia C, Feng M, Pan H, Ma Z, Yue M. 2022.

Prevalence and Genomic Investigation of Multidrug-Resistant Salmonella

Isolates from Companion Animals in Hangzhou, China. Antibiotics 11:625.

34. Pan H, Jia C, Paudyal N, Li F, Mao J, Liu X, Dong C, Zhou K, Liao X, Gong

J, Fang W, Li X, Kehrenberg C, Yue M. 2022. Comprehensive Assessment

of Subtyping Methods for Improved Surveillance of Foodborne Salmonella.

Microbiol Spectr 10:e02479-22.

35. Kumar D, Pornsukarom S, Thakur S. 2019. Antibiotic usage in poultry

production and antimicrobial-resistant Salmonella in poultry. Food safety

in poultry meat production 47–66.

36. Jin C, Jia C, Hu W, Xu H, Shen Y, Yue M. 2024. Predicting antimicrobial

resistance in E. coli with discriminative position fused deep learning

classifier. Computational and Structural Biotechnology Journal

23:559–565.

37. Elbediwi M, Tang Y, Shi D, Ramadan H, Xu Y, Xu S, Li Y, Yue M. 2021.

Genomic Investigation of Antimicrobial-Resistant Salmonella enterica

Isolates From Dead Chick Embryos in China. Front Microbiol 12:684400.

38. Li Y, Kang X, Ed-Dra A, Zhou X, Jia C, Müller A, Liu Y, Kehrenberg C, Yue

M. 2022. Genome-Based Assessment of Antimicrobial Resistance and

Virulence Potential of Isolates of Non-Pullorum/Gallinarum Salmonella

Serovars Recovered from Dead Poultry in China. Microbiol Spectr

10:e00965-22.

39. Xu Y, Zhou X, Jiang Z, Qi Y, Ed-dra A, Yue M. 2020. Epidemiological

Investigation and Antimicrobial Resistance Profiles of Salmonella Isolated

From Breeder Chicken Hatcheries in Henan, China. Front Cell Infect

Microbiol 10:497.

40. Golkar Z, Bagasra O, Pace DG. 2014. Bacteriophage therapy: a potential

solution for the antibiotic resistance crisis. J Infect Dev Ctries 8:129–136.

41. Kang X, An H, Wang B, Huang L, Huang C, Huang Y, Wang Z, He F, Li Y,

Yue M. 2024. Integrated OMICs approach reveals energy metabolism

pathway is vital for *Salmonella* Pullorum survival within the egg white.

mSphere 9:e00362-24.

42. Peng X, Ed-Dra A, Song Y, Elbediwi M, Nambiar RB, Zhou X, Yue M. 2022.

Lacticaseibacillus rhamnosus alleviates intestinal inflammation and

promotes microbiota-mediated protection against Salmonella fatal

infections. Front Immunol 13:973224.

43. Siddique A, Wang Z, Zhou H, Huang L, Jia C, Wang B, Ed-Dra A, Teng L,

Li Y, Yue M. 2024. The Evolution of Vaccines Development across

Salmonella Serovars among Animal Hosts: A Systematic Review.

Vaccines 12:1067.

44. Jiang Z, Kang X, Song Y, Zhou X, Yue M. 2023. Identification and

Evaluation of Novel Antigen Candidates against Salmonella Pullorum

Infection Using Reverse Vaccinology. Vaccines 11:865.

45. Bankevich A, Nurk S, Antipov D, Gurevich AA, Dvorkin M, Kulikov AS,

Lesin VM, Nikolenko SI, Pham S, Prjibelski AD, Pyshkin AV, Sirotkin AV,

Vyahhi N, Tesler G, Alekseyev MA, Pevzner PA. 2012. SPAdes: A New

Genome Assembly Algorithm and Its Applications to Single-Cell

Sequencing. Journal of Computational Biology 19:455–477.

46. Ellington MJ, Ekelund O, Aarestrup FM, Canton R, Doumith M, Giske C,

Grundman H, Hasman H, Holden MTG, Hopkins KL, Iredell J, Kahlmeter G,

Köser CU, MacGowan A, Mevius D, Mulvey M, Naas T, Peto T, Rolain J-M,

Samuelsen Ø, Woodford N. 2017. The role of whole genome sequencing

in antimicrobial susceptibility testing of bacteria: report from the EUCAST

Subcommittee. Clinical Microbiology and Infection 23:2–22.

47. Hasman H, Saputra D, Sicheritz-Ponten T, Lund O, Svendsen CA,

Frimodt-Møller N, Aarestrup FM. 2014. Rapid Whole-Genome Sequencing

for Detection and Characterization of Microorganisms Directly from

Clinical Samples. J Clin Microbiol 52:139–146.

48. Nguyen L-T, Schmidt HA, Von Haeseler A, Minh BQ. 2015. IQ-TREE: A

Fast and Effective Stochastic Algorithm for Estimating

Maximum-Likelihood Phylogenies. Molecular Biology and Evolution

32:268–274.

49. Letunic I, Bork P. 2021. Interactive Tree Of Life (iTOL) v5: an online tool for

phylogenetic tree display and annotation. Nucleic Acids Research

49:W293–W296.

50. Cheng L, Connor TR, Siren J, Aanensen DM, Corander J. 2013.

Hierarchical and Spatially Explicit Clustering of DNA Sequences with

BAPS Software. Molecular Biology and Evolution 30:1224–1228.

51. Croucher NJ, Page AJ, Connor TR, Delaney AJ, Keane JA, Bentley SD,

Parkhill J, Harris SR. 2015. Rapid phylogenetic analysis of large samples

of recombinant bacterial whole genome sequences using Gubbins.

Nucleic Acids Research 43:e15–e15.

52. Sagulenko P, Puller V, Neher RA. 2018. TreeTime: Maximum-likelihood

phylodynamic analysis. Virus Evolution 4.

53. Bouckaert R, Vaughan TG, Barido-Sottani J, Duchêne S, Fourment M,

Gavryushkina A, Heled J, Jones G, Kühnert D, De Maio N, Matschiner M,

Mendes FK, Müller NF, Ogilvie HA, Du Plessis L, Popinga A, Rambaut A,

Rasmussen D, Siveroni I, Suchard MA, Wu C-H, Xie D, Zhang C, Stadler T,

Drummond AJ. 2019. BEAST 2.5: An advanced software platform for

Bayesian evolutionary analysis. PLoS Comput Biol 15:e1006650.

54. Pightling AW, Rand H, Pettengill J. 2022. Using Evolutionary Analyses to

Refine Whole-Genome Sequence Match Criteria. Front Microbiol

13:797997.

55. Feng Y, Pan H, Zheng B, Li F, Teng L, Jiang Z, Feng M, Zhou X, Peng X,

Xu X, Wang H, Wu B, Xiao Y, Baker S, Zhao G, Yue M. 2023. An

integrated nationwide genomics study reveals transmission modes of

typhoid fever in China. mBio 14:e01333-23.

56. Carattoli A, Zankari E, García-Fernández A, Voldby Larsen M, Lund O,

Villa L, Møller Aarestrup F, Hasman H. 2014. *In Silico* Detection and

Typing of Plasmids using PlasmidFinder and Plasmid Multilocus

Sequence Typing. Antimicrob Agents Chemother 58:3895–3903.

57. Hua X, Liang Q, Deng M, He J, Wang M, Hong W, Wu J, Lu B, Leptihn S,

Yu Y, Chen H. 2021. BacAnt: A Combination Annotation Server for

Bacterial DNA Sequences to Identify Antibiotic Resistance Genes,

Integrons, and Transposable Elements. Front Microbiol 12:649969.

58. Wang X, Zhang H, Yu S, Li D, Gillings MR, Ren H, Mao D, Guo J, Luo Y.

2024. Inter-plasmid transfer of antibiotic resistance genes accelerates

antibiotic resistance in bacterial pathogens. The ISME Journal

18:wrad032.

59. Yoshida CE, Kruczkiewicz P, Laing CR, Lingohr EJ, Gannon VPJ, Nash

JHE, Taboada EN. 2016. The Salmonella In Silico Typing Resource

(SISTR): An Open Web-Accessible Tool for Rapidly Typing and Subtyping

Draft Salmonella Genome Assemblies. PLoS ONE 11:e0147101.

60. Zhang S, Den Bakker HC, Li S, Chen J, Dinsmore BA, Lane C, Lauer AC,

Fields PI, Deng X. 2019. SeqSero2: Rapid and Improved *Salmonella*

Serotype Determination Using Whole-Genome Sequencing Data. Appl

Environ Microbiol 85:e01746-19.

61. Zankari E, Hasman H, Cosentino S, Vestergaard M, Rasmussen S, Lund

O, Aarestrup FM, Larsen MV. 2012. Identification of acquired antimicrobial

resistance genes. Journal of Antimicrobial Chemotherapy 67:2640–2644.

62. Van Puyvelde S, Pickard D, Vandelannoote K, Heinz E, Barbé B, De Block

T, Clare S, Coomber EL, Harcourt K, Sridhar S, Lees EA, Wheeler NE,

Klemm EJ, Kuijpers L, Mbuyi Kalonji L, Phoba M-F, Falay D, Ngbonda D,

Lunguya O, Jacobs J, Dougan G, Deborggraeve S. 2019. An African

Salmonella Typhimurium ST313 sublineage with extensive

drug-resistance and signatures of host adaptation. Nat Commun 10:4280.

Figure Legends

Figure 1. Genetic diversity of *S*. Gallinarum biovar Pullorum (bvSP) by geography and time. Different colours were used to represent the various lineages of bvSP: fuchsia for L1, orange for L2a, pink for L2b, red for L3b, and green for L3c. (a) The bvSP in the dataset is classified into five continents based on their isolation regions. The bar graph illustrates the distribution of lineage-specific bvSP across continents, depicted as percentages. The total sample size for each bar is indicated at the top. (b) The bar graph shows the distribution of bvSP isolated in China by lineage per decade. The total sample size for each bar is also indicated on the right side. (c) Geographic distribution of bvSP isolated from China. Doughnut charts in the map show the proportion of lineage types of bvSP collected in the corresponding region, with the total number of isolates in the center.

Figure 2. Phylogenetic tree of *S.* Gallinarum L2b and L3b based on a spatiotemporal Bayesian framework. The phylogenetic tree on the left was constructed using a reference-mapped multiple core-genome SNPs sequence alignment, with recombination regions detected and removed by Gubbins. The spatiotemporal Bayesian framework was configured with the "GTR" substitution model, 4 Gamma Category Count, "Relaxed Clock Log Normal" model, "Coalescent Bayesian Skyline" tree prior model, and a Markov Chain Monte Carlo (MCMC) chain length of 100,000,000, with sampling every 10,000 iterations. Convergence was assessed using Tracer, ensuring all parameter effective sampling sizes (ESS) exceeded 200. Evolutionary time is represented by the length of the branches. The heatmap on the right displays, respectively, the sequence type (ST), region of isolation, and the number (Num.) of antimicrobial resistance genes (ARGs) carried by the corresponding *Salmonella* Gallinarum.

Figure 3. Genomic characteristics of 45 newly isolated *S.* Gallinarum biovar Pullorum (bvSP) strains from Taishun and Yueqing, Zhejiang Province, China. (a) The phylogenetic tree, constructed using cgSNPs, categorizes the 45 bvSP strains into two distinct lineages (L3b and L2b). On the left side of the heatmap, information on isolation regions, sequence types (ST), and sampling times are displayed, with various colors indicating different categories as specified on the left. The right section of the heatmap presents a detailed matrix showing plasmids, ARGs, and cgSNP distances. The presence of a plasmid or ARG in an isolate is denoted by gray shading, while absence is indicated by white. cgSNPs below five were used as the threshold, with darker circles signifying a higher probability of transmission between isolate pairs. (b) The average cgSNP distance between isolates from Taishun and Yueqing. E404 led to an increase in the mean cgSNP distance of bvSP from Taishun. (c) Invasiveness index of bvSP in Taishun and Yueqing. The results show a higher invasiveness index for bvSP isolated from Taishun, indicating that bvSP isolated from Taishun might have greater invasive capabilities among vulnerable hens.

Figure 4. The ARGs carried by *S.* Gallinarum. (a) The phylogenetic tree was constructed using cgSNPs, revealing three *S.* Gallinarum biovars: *S.* Gallinarum biovar Pullorum (bvSP), *S.* Gallinarum biovar Duisburg (bvSD), and *S.* Gallinarum biovar Gallinarum (bvSG). Additionally, *Salmonella serovar* Enteritidis (SE.) is represented by a gray line. Further, bvSP can be subdivided into five lineages: fuchsia (L1), orange (L2a), pink (L2b), red (L3b), and green (L3c). The heatmap on the right indicates the resistome carried by the corresponding *Salmonella*. (b) The dominant resistome types in different regions of China. The y-axis represents the percentage of each dominant

resistome. The total sample size for each bar is indicated at the top. (c) The
average number of resistome carried by bvSP is from different regions of
China.

Figure 5. The primary source of resistome is carried by distinct
mobilome. Different font colours denote various mobilome types. Specifically,
orange signifies Integrons, red denotes transposons, blue represents plasmids,
and green indicates prophages. A black font is utilized to distinguish the
categories of resistome. The connecting line between the resistome and the
mobilome represents the potential carrying relationship. (a) The
mobilome-carried ARGs among *S*. Gallinarum biovar Pullorum (bvSP). (b) The
average number of ARGs carried by the four MGEs in bvSP. The unpaired
t-test was used to compare the differences between isolates, with *P* < 0.05
considered statistically significant. (c-e) The mobilome-carried ARGs among
bvSP isolated from the eastern, southern, and northern regions of China.

Figure 6. The HGT frequency of the resistome among *S*. Gallinarum
biovar Pullorum (bvSP) isolated from China. (a) Workflow for identification
of horizontally transferred ARGs in *Salmonella*. (b) The x-axis represents the
resistome of bvSP, while the y-axis represents the corresponding levels of
HGT frequency. (c) The overall resistome HGT frequency level of bvSP
isolated from the various regions of China. (d) The HGT frequency level of
specific ARGs carried by bvSP isolated from various regions of China. Deeper
colours mean higher HGT frequency. (e) The frequency of horizontal
retransmission of the resistome between different regions of China. A higher
value indicates more frequent transfer events of resistome between two
regions. The colors on the map represent the number of bvSP isolates in each
region. Darker colors indicate a higher number of bvSP isolates in that area.