## Plague Phylodynamics and Phylogeography

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## **Keywords**

- Plague
- Yersinia pestis
- Phylodynamics
- Phylogeography

### Introduction

Plague has an impressively long and expansive history as a human pathogen. The earliest evidence of the plague bacterium *Yersinia pestis* comes from ancient DNA studies, dating its emergence to at least the Neolithic [1,2]. Since then, *Y. pestis* has traveled extensively due to ever-expanding global trade networks and the ability to infect a diverse array of mammalian hosts [3,4]. Few regions of the ancient and modern world remain untouched by this disease, as plague has an established presence on every continent except Oceania [5].

Accompanying this prolific global presence is unnervingly high mortality. The infamous medieval Black Death is estimated to have killed more than half of Europe's population [6]. This virulence can still be observed in the post-antibiotic era, where case fatality rates range from 22-71% [7]. As a result, plague maintains its status as a disease that is of vital importance to current public health initiatives.

This high priority disease status is unsurprising given that *Y. pestis* is a member of the Enterobacteriaceae family. This family includes enteric pathogens such as *Escherichia coli* and *Salmonella typhi* that are commonly transmitted by contaminated food and water. In comparison, the plague bacterium is unique among this family due to a striking difference in host habitat and transmission. *Y. pestis* commonly resides in the blood of its mammalian hosts and can be transmitted to new hosts through an infectious fleabite [8]. In addition to these tissues, the [[Yersinia pestis | plague bacterium]] is also capable of colonizing parts of the mammalian immune system including the lymphatic and reticuloendothelial systems. The large diversity of media in which *Y. pestis* has adapted to colonize is particularly surprising given that it only recently (within the last 20,000 years) diverged as a clone of its parent species Yersinia pseudotuberculosis [9].

Despite a close genetic similarity between *Y. pestis* and *Y. pseudotuberculosis*, in which they share 97% gene identity, they differ widely in their transmission and pathogenecity [10]. Whereas *Y. pseudotuberculosis* causes gastrointestinal disease and is transmitted by the food-borne route, *Y. pestis* is primarily transmitted between mammalian hosts by fleas and causes septicemia, pneumonia, and lymphadenitis. Because of this apparent contradiction of genetic homogeneity and diverse phenotypes, an extensive body of research has formed to address how, when, and where, these epidemiological shifts occurred.

#### TO BE DONE:

- Introduce the genomic composition of Y. pestis and mechanism of evolution.
- Introduce the topics phylodynamics and phylogeography and what is known so far.
- Introduce the problem(s) and our objective(s).

## **Objectives**

- 1. Synthesize and curate publicly available Y. pestis genomes.
- 2. To propose a nuanced phylodynamics model.
- 3. To critique interpretations drawn from phylogeographic approaches?

### **Materials and Methods**

#### **Data Collection**

*Y. pestis* genome sequencing projects were retrieved from the NCBI databases using NCBImeta [11]. 1657 projects were identified and comprised three genomic types: 586 modern assembled, 184 ancient unassembled, and 887 modern unassembled genomes. The 887 modern unassembled genomes were excluded from this project, as the wide variety of laboratory methods and sequencing strategies precluded a standardized workflow. Future work will investigate computationally efficient methods for integrating this data.

Collection location, collection date, and collection host metadata were curated by cross-referencing the original publications. Collection location was transformed to latitude and longitude coordinates using GeoPy and the Nominatim API for OpenStreetMap [12,13,14]. Coordinates were standardized at a sub-country resolution, taking the centroid of the parent province/state. Collection dates were standardized according to their year, and recording uncertainty arising from missing data and radiocarbon estimates. Collection host was the most diverse field with regards to precision, ranging from colloquial nomenclature ("rat") to a genus species taxonomy ("Meriones libycus"). For the purposes of this study, collection host was recorded as Human, Non-Human, or Not Available, given the inability to differentiate non-human mammalian hosts.

Genomes were removed if no associated date or location information could be identified in the literature, or if there was documented evidence of laboratory manipulation.

Two additional datasets were required for downstream analyses. First, *Y. pestis* strain CO92 (GCA\_000009065.1) was used as the reference genome for sequence alignment and variant annotation. Second, *Yersinia pseudotuberculosis* strains NCTC10275 (GCA\_900637475.1) and IP32953 (GCA\_000834295.1) served as an outgroup to root the maximum likelihood phylogeny.

## **Sequence Quality Criteria**

## **Alignment**

Ancient unassembled genomes were downloaded from the SRA database in FASTQ format using the SRA Toolkit [15]. Pre-processing and alignment to the reference genome was performed using the nf-core/eager pipeline, a reproducible workflow for ancient genome reconstruction [16]. Ancient genomes were removed if the number of sites covered at a minimum depth of 3X was less than 70% of the reference genome.

Modern assembled genomes were aligned to the reference genome using Snippy, a pipeline for core genome alignments [17]. Modern genomes were removed if the number of sites covered at a minimum depth of 10X was less than 70% of the reference genome.

A multiple sequence alignment was constructed using the Snippy Core module of the Snippy pipeline. The output alignment was filtered to only include chromosomal variants and to exclude sites that had more than 5% missing data.

## **Phylogenetic Reconstruction**

Model selection was performed using Modelfinder which identified the K3Pu+F+I model as the optimal choice based on the Bayesian Information Criterion (BIC) [18]. A maximum-likelihood phylogeny was then estimated across 10 independent runs of IQTREE [19]. Branch support was evaluated using 1000 iterations of the ultrafast bootstrap approximation, with a threshold of 95% required for strong support [20].

#### **Modified Datasets**

To investigate the influence of between-clade variation in substitution rates, the multiple alignment was separated into the major clades of *Y. pestis*, which will be referred to as the *Clade* dataset. The subclade associated with the Plague of Justinian (0.ANT4) was considered to be a distinct clade separate from its parent (0.ANT) due to its geographic, temporal, and ecological uniqueness. In total, 12 clades were considered and are described in Table 1.

To improve the performance and convergence of Bayesian analysis, a subsampled dataset was constructed. Clades that contained multiple samples drawn from the same geographic location and the same time period were reduced to one representative sample. The sample with the shortest terminal branch length was prioritized, to diminish the influence of derived mutations on the estimated substitution rate. An interval of 25 years was identified as striking an optimal balance, resulting in 200 representative samples.

## **Phylodynamics**

To investigate the degree of temporal signal present in the data, two tests were formed . The first was a root-to-tip (RTT) regression on collection date. This linear model is a simple approach to explore whether the data follows a strict clock model. Uncertainty in the model parameters, namely the mean substitution rate and tMRCA, were estimated using 1000 iterations of the non-parametric bootstrap on the residuals.

While RTT is a practical approach, it has two main limitations: 1) No rate variation is accounted for, and 2) The data are not independent observations due to shared internal branch lengths. Therefore to complement this approach, a bayesian evaluation of temporal signal (BETS) was performed.

A maximum-likelihood timetree was estimated using a least-squares approach as implemented in LSD2 [21]. Rate variation was modeled using a lognormal relaxed clock using the default parameters for the mean (1.0) and the standard deviation (0.2). The outgroup *Y. pseudotuberculosis* was used to root the tree and then subsequently removed.

A bayesian timetree was estimated using ... as implemented in BEAST.

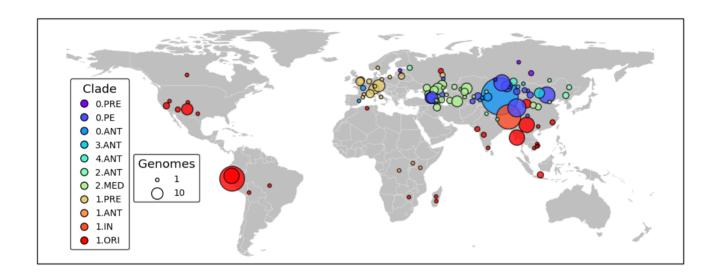
## **Phylogeography**

Geographic location was modeled as a discrete state with transitions following a GTR mugration model as implemented in TreeTime [22].

## **Results**

### **Curated Public Dataset**

After curation, 600 genomes remained, with 539 (90%) being modern in origin and 61 (10%) being ancient. The geographic distribution of samples is shown in Figure  $\underline{1}$ .



**Figure 1:** Geographic distribution of *Yersinia pestis* genomes.

## **Phylogeny**

Divergence-scaled phylogeny of Y. pestis (Figure 2).

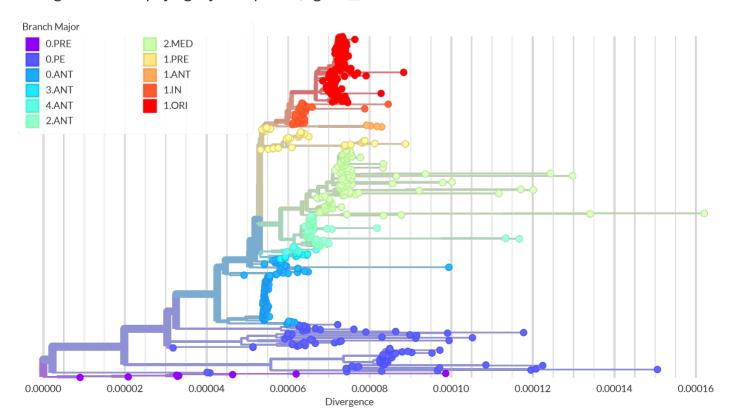


Figure 2: Yersinia pestis phylogeny. (Significant SVG editing required) | 800

## **Phylodynamics**

## **Molecular Clock**

- Y. pestis has extreme rate variation.
- A Root to Tip Regression on collection date confirms this, as the Coefficient of Determination (R<sup>2</sup>) is 0.09, revealing a poor fit to a simple linear model (Table 1).
- To some extent, this variation can be explained by examining the clades in isolation (Figure 3).
- Finding an appropriate evolutionary model is key to estimating historic events, like clade emergence (Figure 4).

Table 1: Temporal signal statistics by clade

Branch	Clade	Origin	R <sup>2</sup>	p-value
all	all	Ancient, Modern	0.09	3.81E-14
0	0.PRE	Ancient	0.91	1.53E-04*
0	0.PE	Modern	0.01	2.25E-01
0	0.ANT4	Ancient	0.66	7.84E-04*
0	0.ANT	Modern	-0.01	7.35E-01
1	1.ANT	Modern	0.45	2.03E-01
1	1.IN	Modern	0.0	3.24E-01
1	1.ORI	Modern	0.04	1.32E-02*
1	1.PRE	Ancient	0.76	1.68E-13*
2	2.ANT	Modern	0.05	5.96E-02
2	2.MED	Modern	0.01	1.86E-01
3	3.ANT	Modern	-0.04	4.39E-01
4	4.ANT	Modern	-0.11	8.80E-01

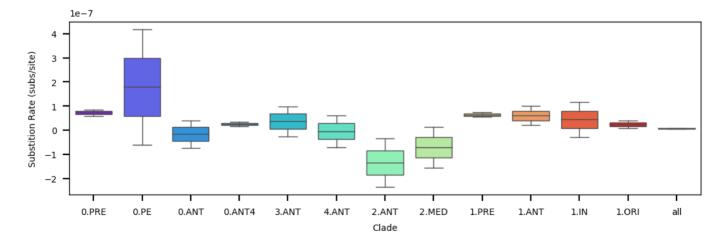


Figure 3: Rate variation by clade.

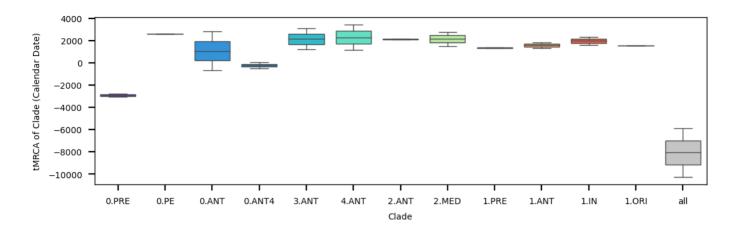
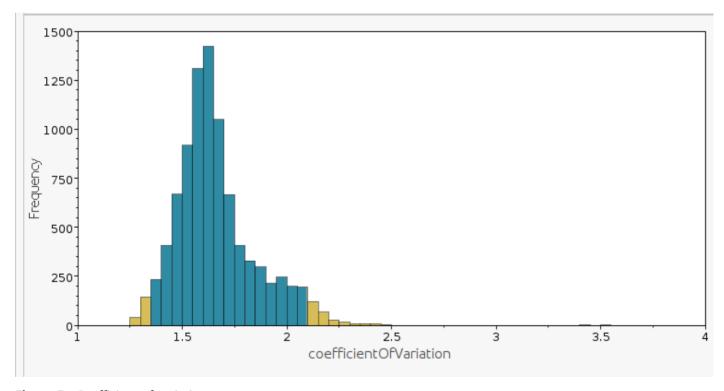


Figure 4: tMRCA by clade.

## **Relaxing the Clock**

Relaxed clock MCMC runs produce a high Coefficient of Variation indicating a relaxed model is
favored over a strict model (Figure 5). However, these runs do not converge, suggesting there is too
much rate variation to confidently estimate key parameters such as the mean Substitution Rate or
tMRCA.



**Figure 5:** Coefficient of variation.

• A strict clock and relaxed clock have overlapping distributions with similar peaks for the Tree Height (blue: strict, green: relaxed) (Figure 6).

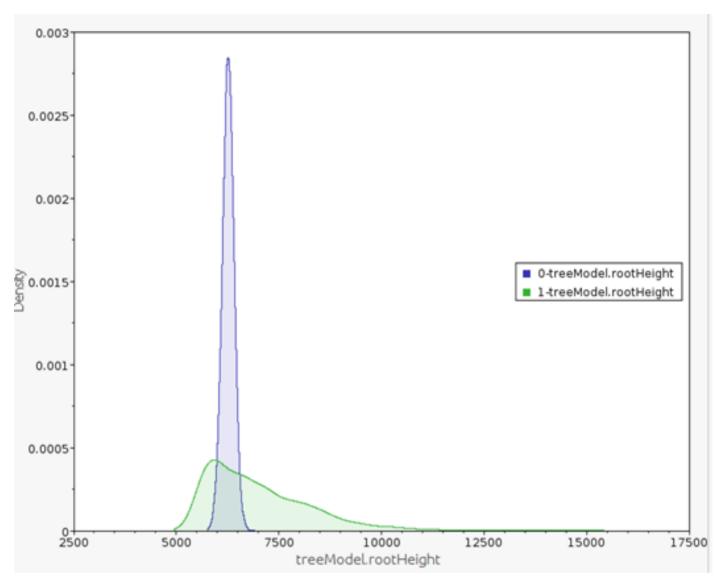


Figure 6: Tree height comparison.

• When estimating a Substitution Rate for all of *Y. pestis*, a [[Clock Model|strict clock]] and relaxed clock produce different estimates (green: strict, orange: relaxed) (Figure 7).

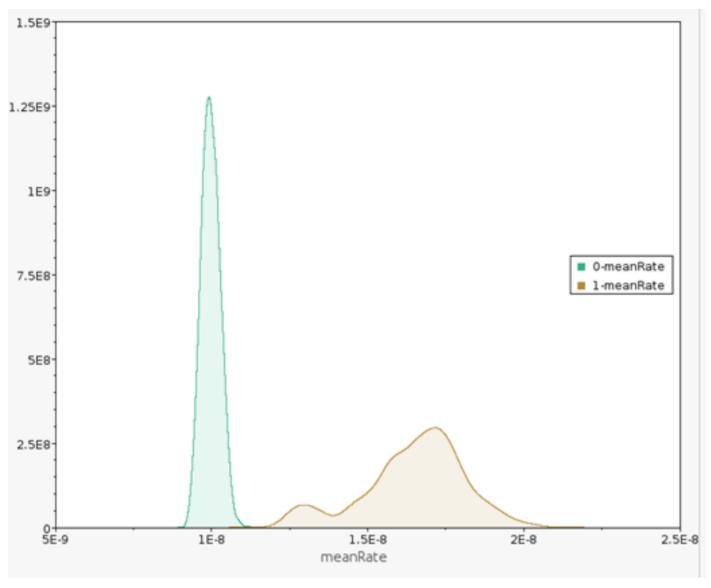
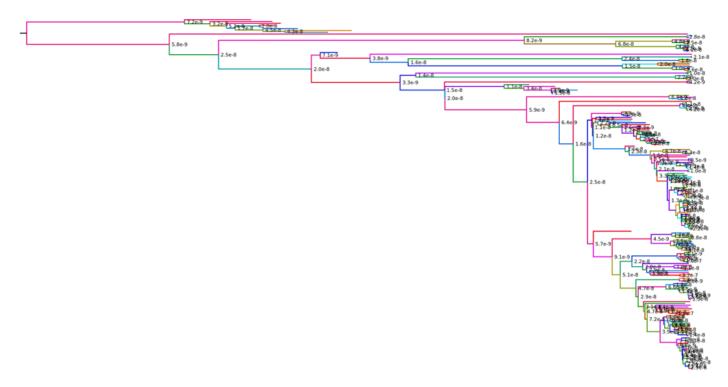


Figure 7: Substitution rate comparison.

• There doesn't appear to be clustering of rates. Branches with high rates are next to those with low rates (Figure §).



**Figure 8:** Time tree colored by rate.

## **Discussion**

## Conclusion

### References

#### 1. The Stone Age Plague and Its Persistence in Eurasia

Aida Andrades Valtueña, Alissa Mittnik, Felix M. Key, Wolfgang Haak, Raili Allmäe, Andrej Belinskij, Mantas Daubaras, Michal Feldman, Rimantas Jankauskas, Ivor Janković, ... Johannes Krause *Current Biology* (2017-12-04)

DOI: 10.1016/j.cub.2017.10.025 · PMID: 29174893

#### 2. Emergence and spread of basal lineages of Yersinia pestis during the Neolithic Decline

Nicolás Rascovan, Karl-Göran Sjögren, Kristian Kristiansen, Rasmus Nielsen, Eske Willerslev, Christelle Desnues, Simon Rasmussen

Cell (2019-01-10) https://www.cell.com/cell/abstract/S0092-8674(18)31464-8

DOI: 10.1016/j.cell.2018.11.005 · PMID: 30528431

#### 3. Trade routes and plague transmission in pre-industrial Europe

Ricci P. H. Yue, Harry F. Lee, Connor Y. H. Wu

Scientific Reports (2017-10-11) https://www.ncbi.nlm.nih.gov/pmc/articles/PMC5636801/

DOI: 10.1038/s41598-017-13481-2 · PMID: 29021541 · PMCID: PMC5636801

#### 4. Yersinia pestis-etiologic agent of plague

R. D. Perry, J. D. Fetherston

Clinical Microbiology Reviews (1997-01)

PMID: 8993858 · PMCID: PMC172914

#### 5. Plague

World Health Organization

(2017-10-31) <a href="https://www.who.int/news-room/fact-sheets/detail/plague">https://www.who.int/news-room/fact-sheets/detail/plague</a>

#### 6. The Black Death, 1346-1353: The Complete History

O. J. Benedictow

Boydell Press (2004)

ISBN: <u>0-85115-943-5</u>

#### 7. Plague around the world in 2019

**Eric Bertherat** 

Weekly Epidemiological Record (2019-06-21) <a href="https://apps.who.int/iris/bitstream/handle/">https://apps.who.int/iris/bitstream/handle/</a> 10665/325481/WER9425-en-fr.pdf

#### 8. Recent trends in plague ecology

K Gage, M Kosoy

(2006) http://reviverestore.org/wp-content/uploads/2015/02/Gage-and-Kosoy\_USGS-Blk-footed-ferret-symp\_2006-copy.pdf

# 9. Yersinia pestis, the cause of plague, is a recently emerged clone of Yersinia pseudotuberculosis

M. Achtman, K. Zurth, G. Morelli, G. Torrea, A. Guiyoule, E. Carniel *Proceedings of the National Academy of Sciences of the United States of America* (1999-11-23)

DOI: <u>10.1073/pnas.96.24.14043</u> · PMID: <u>10570195</u> · PMCID: <u>PMC24187</u>

# 10. Insights into the evolution of Yersinia pestis through whole-genome comparison with Yersinia pseudotuberculosis

P. S. G. Chain, E. Carniel, F. W. Larimer, J. Lamerdin, P. O. Stoutland, W. M. Regala, A. M. Georgescu, L. M. Vergez, M. L. Land, V. L. Motin, ... E. Garcia

*Proceedings of the National Academy of Sciences* (2004-09-21) <a href="http://www.pnas.org/cgi/doi/10.1073/">http://www.pnas.org/cgi/doi/10.1073/</a> <a href="pnas.0404012101">pnas.0404012101</a>

DOI: 10.1073/pnas.0404012101

#### 11. NCBImeta

Katherine Eaton

NCBImeta (2019) https://github.com/ktmeaton/NCBImeta

#### 12. GeoPy: A Python client for several popular geocoding web services.

Kostya Esmukov

(2020-12) <a href="https://github.com/geopy/geopy">https://github.com/geopy/geopy</a>

#### 13. Nominatim: A tool to search OpenStreetMap data.

Sarah Hoffman

(2020-12) <a href="https://github.com/osm-search/Nominatim">https://github.com/osm-search/Nominatim</a>

#### 14. Planet dump retrieved from https://planet.osm.org

OpenStreetMap Contributors

(2017) https://www.openstreetmap.org

#### 15. ncbi/sra-tools

NCBI - National Center for Biotechnology Information/NLM/NIH (2021-05-18) <a href="https://github.com/ncbi/sra-tools">https://github.com/ncbi/sra-tools</a>

#### 16. Reproducible, portable, and efficient ancient genome reconstruction with nf-core/eager

James A. Fellows Yates, Thiseas C. Lamnidis, Maxime Borry, Aida Andrades Valtueña, Zandra Fagernäs, Stephen Clayton, Maxime U. Garcia, Judith Neukamm, Alexander Peltzer *PeerJ* (2021-03-16) https://peerj.com/articles/10947

DOI: 10.7717/peerj.10947

#### 17. Snippy: Rapid haploid variant calling and core genome alignment.

Torsten Seemann

(2020-03-08) https://github.com/tseemann/snippy

### 18. ModelFinder: fast model selection for accurate phylogenetic estimates

Subha Kalyaanamoorthy, Bui Quang Minh, Thomas K. F. Wong, Arndt von Haeseler, Lars S. Jermiin *Nature Methods* (2017-06) <a href="http://www.nature.com/articles/nmeth.4285">http://www.nature.com/articles/nmeth.4285</a>

DOI: 10.1038/nmeth.4285

#### 19. IQ-TREE 2: New Models and Efficient Methods for Phylogenetic Inference in the Genomic Era

Bui Quang Minh, Heiko A. Schmidt, Olga Chernomor, Dominik Schrempf, Michael D. Woodhams, Arndt von Haeseler, Robert Lanfear

Molecular Biology and Evolution (2020-05-01) https://academic.oup.com/mbe/article/

37/5/1530/5721363

DOI: 10.1093/molbev/msaa015

#### 20. UFBoot2: Improving the Ultrafast Bootstrap Approximation

Diep Thi Hoang, Olga Chernomor, Arndt von Haeseler, Bui Quang Minh, Le Sy Vinh *Molecular Biology and Evolution* (2018-02-01) <a href="https://academic.oup.com/mbe/article/35/2/518/4565479">https://academic.oup.com/mbe/article/35/2/518/4565479</a>

DOI: 10.1093/molbev/msx281

#### 21. Fast Dating Using Least-Squares Criteria and Algorithms

Thu-Hien To, Matthieu Jung, Samantha Lycett, Olivier Gascuel

Systematic Biology (2016-01) <a href="https://academic.oup.com/sysbio/article-lookup/doi/10.1093/sysbio/syv068">https://academic.oup.com/sysbio/article-lookup/doi/10.1093/sysbio/syv068</a>

DOI: 10.1093/sysbio/syv068

#### 22. TreeTime: Maximum-likelihood phylodynamic analysis

Pavel Sagulenko, Vadim Puller, Richard A Neher

Virus Evolution (2018-01-08) https://www.ncbi.nlm.nih.gov/pmc/articles/PMC5758920/

DOI: 10.1093/ve/vex042 · PMID: 29340210 · PMCID: PMC5758920