1 Mining ancient microbiomes using selective enrichment of damaged DNA

2 molecules

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11 Abstract

The identification of bona fide microbial taxa in microbiomes derived from historical 12 13 samples is complicated by the unavoidable mixture between DNA from ante- and 14 post-mortem microbial colonizers. One possibility to distinguish between these sources of microbial DNA is guerying for the presence of age-associated degradation patterns 15 typical of ancient DNA (aDNA). The presence of uracils, resulting from cytosine 16 deamination, has been detected ubiguitously in aDNA retrieved from diverse sources, 17 18 and used as an authentication criterion. Here, we employ a library preparation method 19 that separates molecules that carry uracils from those that do not for a set of samples that includes Neandertal remains, herbaria specimens and archaeological plant 20 21 remains. We show that this method facilitates the discovery of authentic ancient 22 microbial taxa, as it amplifies degradation patterns that would otherwise be difficult to 23 detect in sequences from diverse microbial mixtures.

24 Importance

The utility of DNA from historical specimens is being recognized in a growing number of fields, ranging from human, animal and plant genetics, to microbiology and epidemiology of infectious diseases. Providing positive evidence for the authenticity of such ancient DNA from diverse sources is instrumental for all studies that make use of this resource. This is especially challenging when studying ancient microbes, due to their high genetic diversity and the incompleteness of reference databases. The method

we employ and characterize here aids this process through the selective enrichment of
 molecules that carry signatures of age-associated degradation.

33 Introduction

34 DNA retrieved from historical or ancient samples is a complex mixture of molecules that contains not only endogenous host DNA, but also DNA from 35 36 microorganisms that were present ante-mortem or that colonized the tissue post-mortem (1). Therefore, all ancient DNA (aDNA) shotgun sequencing projects are metagenomic 37 in nature. While earlier aDNA research has mostly focused on the evolution of animals 38 39 and plants (2, 3), a growing number of studies are now centering on the identification and characterization of ancient pathogens and microbiomes (4). Ancient microbes 40 permit the replacement of indirect inferences about the past with direct observations of 41 microbial genomes through time. In the pathogen field, it has been possible to identify 42 43 causal and/or associated agents of historical plant and animal disease outbreaks, as 44 well as their spreading patterns throughout both space and time (e.g. (5, 6)). Another 45 challenging endeavour is the characterization of shifts in composition of microbial 46 communities over time. For example, dental calculus from hominids has been exploited as a source of ancient microbiomes and analyzed in the context of diet and lifestyle 47 changes (7–9), whereas coprolites have been used to investigate ecological interactions 48 49 between animals and microorganisms (10). However, this approach is at its beginnings

and the influence of major selective pressures on microbiome evolution remains to be
 explored.

52 A major challenge for the study of aDNA in general, and ancient microbiomes in 53 particular, is the presence of contaminating exogenous DNA, which makes distinction between bona fide ancient microbiome sequences and those of recent origin crucial. 54 One of the most typical features of aDNA is the presence of uracils (Us) that originate 55 from post-mortem deamination of cytosines (Cs), especially in single-stranded 56 57 overhangs at molecule ends (11). Uracils are read as thymines (Ts) by most DNA polymerases, which generates a characteristic increase in C-to-T substitutions at the 58 end of aDNA sequences ((11), Figure 1D and 2A). The presence of such C-to-T 59 substitutions can be used as evidence for the authenticity of DNA sequences retrieved 60 from historical material (12–14). 61

Recently, a single-stranded library preparation method (U-selection) was 62 63 developed, which allows physical separation of uracil-containing molecules from non-deaminated ones (15). In U-selection all library molecules are initially immobilized 64 on streptavidin beads, to which molecules without uracils remain attached (U-depleted 65 fraction), while uracil-containing molecules (originally deaminated) are released into 66 solution (U-enriched fraction). U-selection was originally developed with the aim of 67 increasing the amount of ancient hominid DNA (e.g. Neandertals) from a background of 68 present-day human and microbial DNA (15). However, the method seems to be 69 70 specially suited to study microbiomes, due to the inherent difficulty to authenticate their

71 ancient origin. This complication arises from the fact that microbes can colonize tissues at different times, resulting in different levels of deamination of microbial DNA in 72 73 historical samples. Although sequences that carry terminal C-to-T substitutions can be 74 selected in silico (16, 17), there are two factors that could hinder this approach. Firstly, 75 low levels of deamination will reduce the number of molecules suitable for selection in 76 silico. Secondly, high sequence divergence between samples and reference genomes 77 can mask age-associated deamination signals thereby hinder authentication.. Consequently, enriching for deaminated molecules during library preparation is 78 79 fundamental to tackling these problems. As a proof-of-principle experiment, we used here U-selection in combination with taxonomic binning of Illumina sequenced reads to 80 81 characterize the microbiomes of Neandertal bones (~39,000 years old), herbaria specimens (between 41 and 279 years old) and plant archaeological remains (~2,000 82 83 years old) (Table 1).

ID	Country of	Age	Species	Reference*
	origin			
KM177500	UK	171 ⁺	Solanum tuberosum	1
KM177497	UK	170 ⁺	Solanum tuberosum	1
BM000815937	UK	279 ⁺	Solanum lycopersicum	2
BH0000061459	USA	119 ⁺	Arabidopsis thaliana	3
OSU13900	USA	82+	Arabidopsis thaliana	4
NY1365364	USA	127+	Arabidopsis thaliana	5
NY1365375	USA	119 ⁺	Arabidopsis thaliana	5
CS5	USA	1852++	Zea mays	6
CS6	USA	Undated	Zea mays	6
CS20	USA	1881++	Zea mays	6
El Sidrón 1253	Spain	39,000++	Neanderthal	7
Vindija 33.17	Croatia	Undated	Neanderthal	8
Vindija 33.19	Croatia	Undated	Neanderthal	8

Table 1. Provenance of herbarium specimens and archaeological remain

*¹Kew Royal Botanical Gardens; ²Natural History Museum, London; ³Cornell Bailey
 Hortorium; ⁴Ohio State University Herbarium; ⁵New York Botanical Garden;

⁸⁷ ⁶Turkey Pen Shelter, UTAH, USA; ⁷El Sidrón Cave, Spain; ⁸Vindija Cave, Croatia.

⁸⁸ ⁺Calculated from collection dates (in years).

^{**}B.P. (Before present years)

90 **Results and Discussion**

Our experiments were motivated by the previous observation that in some Neandertal samples, e.g. from El Sidrón, Spain, the proportion of Neandertal DNA fragments remains unchanged in both the U-depleted and U-enriched fractions, whereas in others, from Vindija Cave, Croatia, this proportion increased in the Uracil-enriched fraction (15). It was hypothesized that the latter effect could have been

96 due to differences in deamination, and hence in age, between Neandertal- and microbial-derived DNA fragments. To explore this effect further, we re-analyzed the 97 previously generated Neanderthal sequence data from both sites by performing 98 99 taxonomic binning of reads derived from the U-depleted and U-enriched fractions, instead of aligning them only to the human reference genome, as had been done 100 101 previously. Reads aligning to the two most abundant bacterial phyla (Actinobacteria and 102 Proteobacteria) from the Vindija Neandertals were enriched in the U-depleted fraction, 103 while hominid reads were enriched in the U-enriched fraction (Figure 1A). This is in 104 accordance with a previous study that reported absence of DNA damage in Actinobacteria derived from a Neandertal bone from Vindija cave (18). In contrast, in 105 106 reads obtained from the El Sidrón Neandertals, we found enrichment of both hominid and Actinobacteria reads in the U-enriched fraction, whereas Proteobacteria reads were 107 108 enriched in the U-depleted fraction. (Figure 1B). Overall, bacteria-derived reads were 109 dominated by the Actinobacteria Streptosporangium roseum (Figure 1C), which showed almost 50% deamination at the first base in the U-enriched fraction (Figure 1D), 110 111 suggesting its ancient origin. The analysis of reads derived from Neandertal bones illustrates how U-selection permits distinguishing between ancient bacteria enriched in 112 113 the U-enriched fraction and more recent colonizers enriched in the U-depleted fraction.

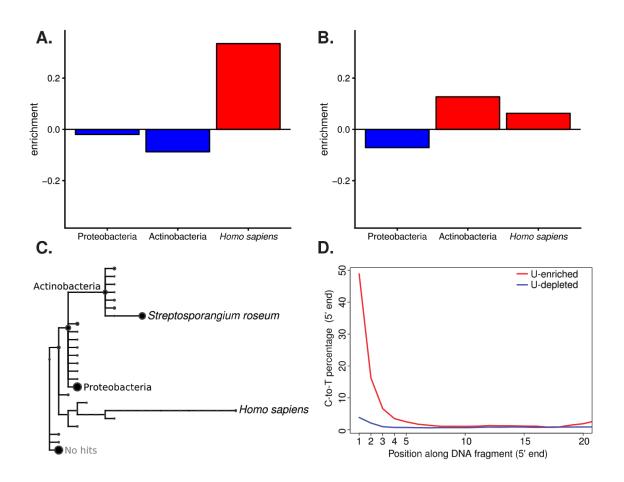


Figure 1. Relative enrichment, taxonomic assignment and substitution profiles of 114 Neandertal-derived U-selected libraries. A. Relative enrichment (number of reads) 115 in the U-enriched relative to the U-depleted fraction from Vindija Neandertal 116 117 assigned to the phyla Actinobacteria and Proteobacteria, as well as to Homo sapiens. B. Relative enrichment (number of reads) in the U-enriched relative to 118 the U-depleted fraction from Sidrón Neandertal assigned to the phyla 119 Actinobacteria and Proteobacteria, as well as to Homo sapiens. C. Taxonomic tree 120 of reads from Sidrón Neandertal assigned to different taxonomic levels. The size 121 122 of the circle represents the amount of reads assigned to a particular part or the taxonomy. Assignments to the phyla Actinobacteria and Proteobacteria, as well as 123 the species Streptosporangium roseum and Homo sapiens are named in the 124 taxonomic tree. D. Cytosine to Thymine substitutions at the 5' end of reads aligned 125 to S. roseum from the Sidrón Neandertal U-selected library (U-enriched and 126 127 U-depleted fractions).

128 In order to further evaluate the performance of U-selection in characterizing microbial communities, we selected a set of plant samples (both herbaria specimens 129 130 and archaeological remains) with low levels of deamination. We extracted DNA from plant samples and generated libraries using both a regular double-stranded (ds) 131 approach (19), and U-selection (15). Sequences from the dsDNA libraries were then 132 133 used as a baseline to evaluate depletion and enrichment of uracil-containing molecules (Figure 2A). U-selection successfully enriched for deaminated molecules in all plant 134 samples, as it is manifest in the much higher levels of deamination present in the 135 136 U-enriched fraction compared with the dsDNA libraries and the U-depleted fraction (Figure 2A-B). The plant samples showed substantial variation in the content of 137 138 endogenous DNA (2.8-91%), which was very similar between the U-depleted and U-enriched fractions, indicating similar levels of deamination between host- and microbe 139 derived reads (Figure S1A). Assuming that plant- and microbial-derived DNA deaminate 140 141 at a similar rate (20), this observation indicates that microbes found in plant tissue were present at the time of collection or colonized the tissue shortly thereafter. The 142 143 percentage of reads (including host-derived reads) that could be taxonomically binned varied depending on the sample (Figure S1B) and, since the host genome was included 144 in the nucleotide database, positively correlated with the percentage of host 145 146 endogenous DNA (Figure S1C). The inability to taxonomically assign the vast majority of 147 reads from samples with low endogenous DNA reflects the incompleteness of the 148 reference database compared to the diversity of the microbiomes in those samples.

149 Additionally, single stranded DNA library preparation methods as employed during Uracil enrichment generate shorter reads (21, 22), which are more difficult to map to a 150 151 reference genome and to assign taxonomically to a nucleotide database. This is 152 reflected in the higher percentage of reads mapped and assigned from the dsDNA library compared with shorter reads derived from both the U-depleted and U-enriched 153 154 fraction (Figure S2A-B). Originally, it was reported that the U-enriched fraction shows a mild increase in GC-content (15), however in the plant libraries analyzed here we did not 155 156 find a significant difference in GC-content between the U-depleted and U-enriched 157 fractions (Figure S2C). In theory, since Us originate from Cs, the U-enriched fraction would be enriched for GC-rich species and GC-rich genomic regions within a given 158 159 genome. However, as the enrichment would depend on the diversity of taxa present and their relative age difference, and hence difference in deamination, GC-biases, if any, are 160 161 expected to be highly sample-dependent.

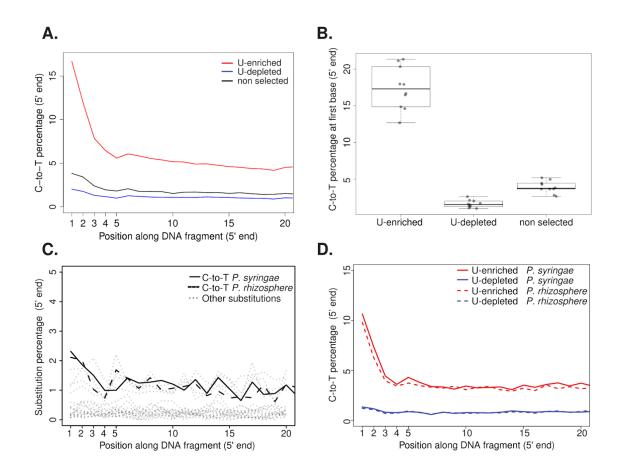


Figure 2. Patterns of cytosine to thymine (C-to-T) substitutions at the 5' end of 162 plant- and Pseudomonas-derived reads. A. C-to-T substitutions at the 5' end of 163 Solanum tuberosum sample KM177500 for a non-selected and U-selected library 164 165 (U-enriched and U-depleted fractions). B. Distributions of C-to-T substitution percentage at first base (5' end) for non-selected and U-selected libraries 166 (U-enriched and U-depleted fractions). Median values are denoted as black lines 167 and points show the original value for each individual sample. C. Substitution 168 patterns at the 5' end of Pseudomonas syringae and Pseudomonas rhizosphere 169 170 mapped reads from a non-selected library from a Solanum tuberosum sample 171 KM177500. D. Cytosine to Thymine substitutions at the 5' end of P. syringae and P. rhizosphere mapped reads from a U-selected library (U-enriched and 172 U-depleted fractions) from a Solanum tuberosum sample KM177500. 173

Given the low taxonomic diversity of microorganisms in the samples included in 174 175 proof-of-principle experiment, instead of centering our analyses on the our 176 compositional assessment of microbial communities, we investigated in detail samples in which a specific microbe or group of microbes were more prevalent based on read 177 178 abundance. We identified a large number of reads that were assigned to the bacterium Pantoea vagans in a potato (Solanum tuberosum) and a maize (Zea mays) sample 179 180 (Figure S3). In both samples we found patterns of C-to-T substitutions that suggest the 181 historical nature of the sequenced reads (Figure 3A). Since *P. vagans* is a plant epiphyte (23), it is not entirely surprising to find it in two different plant species. We compared the 182 potato and maize *P. vagans* with publicly available genomes using single nucleotide 183 polymorphisms (SNPs) ascertained in these modern samples. Our analysis linked the 184 185 two historical strains to a distinct cluster of modern strains based on genetic similarity 186 (Figure 3B). Based on a set of 432,891 SNPs, the two historical isolates showed 95% 187 SNP identity between them, and an average of 92% SNP identity between historical and modern strains of the same cluster. Conversely, comparisons between historical strains 188 189 and any modern strain of a different cluster showed only an average of 59% identity at 190 variable positions.

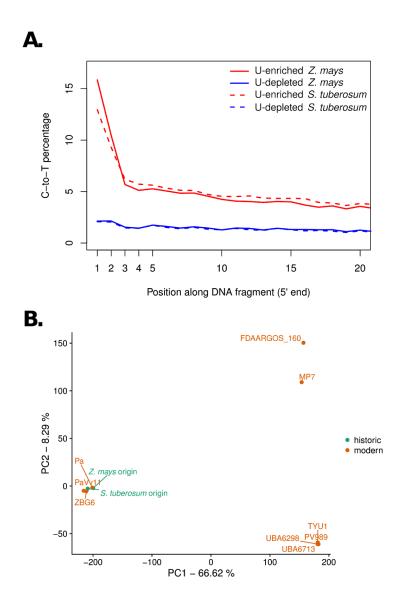


Figure 3. Substitution patterns and genetic distances of the bacterium Pantoea 191 vagans identified from Zea mays and Solanum tuberosum samples (same 192 samples as in Figure S3). A. Cytosine to Thymine substitutions at the 5' end of P. 193 vagans for U-selected libraries (U-enriched and U-depleted fractions) from Z. 194 195 mays and S. tuberosum. B. Principal component analysis of P. vagans from Z. mays and S. tuberosum samples, as well as nine publicly available genomes, 196 based on single nucleotide polymorphisms. Numbers in axis labels indicate the 197 percentage of the variance explained by each principal component (PC). 198

199 In a potato sample, in which the pathogenic oomycete *Phytophthora infestans* was previously identified (6), we found a large portion of reads assigned to the bacterial 200 genus *Pseudomonas*. Reads were assigned in particular to the species *Pseudomonas* 201 syringae and Pseudomonas rhizosphere in different proportions (Figure S4). We 202 performed de novo assembly using reads assigned to the genus Pseudomonas and 203 204 aligned the contigs to the reference genomes of *P. syringae* and *P. rhizosphere* covering about 80% of both reference genomes (Figure S5). We subsequently filtered for contigs 205 that aligned uniquely to either P. syringae and P. rhizosphere genomes and found 206 207 different k-mer coverage distributions in contigs aligning uniquely to each genome (Figure S6), an observation that reinforced our confidence in the presence of the two 208 209 *Pseudomonas* species in this sample. Due to the high level of sequence divergence 210 between the *Pseudomonas* in our sample and the reference genomes present in the 211 database, it is difficult to assess typical deamination patterns in the dsDNA library 212 (Figure 2C). However, we were able to examine damage patterns in both *Pseudomonas* species using the U-enriched fraction (Figure 2D), since the C-to-T signal is amplified 213 214 and is much higher than the basal level of substitutions.

In summary, we showed here that the U-selection method selectively enriches for authentic microbial aDNA molecules in samples from plant and animal tissues with a wide-distribution of ages and deamination levels. For instance, in *P. vagans*, U-selection increases the fraction of molecules carrying a terminal C-to-T substitution at the 5'-end 2-3 fold over the library without enrichment, relative to the total number of molecules

220 sequenced. We think that the application of U-selection for ancient microbiome research will be particularly useful in both samples with minute levels of deamination, where the 221 nucleotide divergence between samples and reference genomes will obscure the 222 223 identification of the C-to-T pattern typical of aDNA, as well as in moderately or heavily deaminated samples which carry modern contaminants, since in those samples ancient 224 225 taxa would be efficiently enriched. Since it is extremely difficult to differentiate between ante-mortem and early post-mortem colonizers based only on deamination patterns, it is 226 227 fundamental to also evaluate the biological relevance of detected taxa by comparing 228 them with reference modern microbiomes.

229 Materials and Methods

230 Plant Samples

- 231 We used herbarium specimens from three different plant species (*Arabidopsis thaliana*,
- 232 Solanum tuberosum, and Solanum lycopersicum) with ages ranging from 41 to 279
- 233 years (Table 1). S. *tuberosum* herbarium specimens were documented to be infected by
- 234 *Phytophthora infestans* (6). We used also *Zea mays* archaeological remains excavated
- in the Turkey Penn shelter in Utah, USA (24). The Zea mays samples were dated using
- accelerator mass spectrometry and have ages ranging between 1852 and 1881 years
- BP (Before Present) (Table 1). The sequencing data for these samples is available on
 the European Nucleotide Archive under study number PRJEB30666.

239 Neanderthal samples

We used Neanderthal samples (Table 1) prepared by (15), which were sequenced deeper for this study.

242 **DNA extraction and library preparation**

- DNA from all herbarium specimens and plant archaeological remains was performed as
 previously described (6).
- For each plant sample two libraries were produced, one using a double-stranded library preparation (19, 25) and the second using the single-stranded U-selection protocol (15)

without enzymatic removal of uracils (26).

248 Sequencing and initial data processing

Since the length of aDNA molecules is in most of the cases shorter than the read length of the sequencing platform, it is possible that a fragment of the aDNA molecule is sequenced by both the forward and reverse read, and also that a part of the adaptor is sequenced (27). Therefore, it is recommended to merge sequences based on the overlapping fraction sequenced by both forward and reverse reads (27). We remove adaptors and merged sequences using the software leeHom with the "--ancientdna" option (28). Putative chimeric sequences were flagged as failing quality.

256 Mapping of sequenced reads to their host genome

Merged reads were mapped as single-ended reads to their respective or most closely relative genome: *Zea mays* (29), *Arabidopsis thaliana* (30, 31), *Solanum tuberosum* (32), *Solanum lycopersicum* (33), *Homo sapiens* (Genome Reference Consortium Human Build 37). The mapping was performed using BWA-MEM (version 0.7.10) with default parameters, which includes a minimum length cutoff of 30 bp (34).

262 Metagenomics assignment of sequenced reads

Reads were aligned to the full non-redundant NCBI nucleotide collection (nt) database (downloaded January 2015) using MALT (version 0.0.12, (35)) in BlastN mode. The resulting RMA files were analyzed using MEGAN (version 5.11.3, (36)). The reads were assigned to the NCBI taxonomy using a lowest common ancestor algorithm (36).

267 Mapping of sequenced reads to microbial genomes

Libraries were mapped to microbial reference genomes of interest, after the presence of 268 269 certain taxa was detected during metagenomic assignment. Specifically, the references 270 of Streptosporangium roseum (37), Pseudomonas syringae pv. syringae B728a (38), Pseudomonas rhizosphaerae (39) and Pantoea vagans (23) were used. Since mapping 271 272 metagenomic libraries to bacterial reference genomes is very prone to false alignments, 273 we used a different mapping strategy for these genomes. The mappings were 274 performed with bowtie2 (version 2.2.4, (40)), with the settings "--score-min 'L,-0.3,-0.3' 275 --sensitive --end-to-end" to increase stringency.

Assessment of nucleotide substitution patterns

All types of nucleotide substitutions relative to the reference genome were calculated per library using mapDamage 2.0 (v. 2.0.2–12, (41)). The percentages of C-to-T substitutions at the 5' end were extracted from the output file 5pCtoT_freq.txt produced by mapDamage.

281 *Pantoea vagans* genomic variation

In order to reduce the effect of aDNA-associated C-to-T substitutions on variant discovery, we used exclusively the U-depleted fraction of libraries where *P. vagans* was detected in the metagenomic screening. The libraries were mapped to the *P. vagans* 285 reference genome using BWA-MEM, to reduce reference bias and increase SNP 286 discovery. False alignments from the metagenomic libraries posed a lesser problem here, as variants were ascertained based on modern material. Variants for historical 287 samples were called for both libraries together using the bcftools (version 1.8, (42)) 288 utilities mpileup ("bcftools mpileup -q 1 -I -Ou -f \$REF \$IN1 \$IN2") and call ("bcftools call 289 290 --ploidy 1 -m -O -z"). Additionally, 11 assemblies of different contiguity were downloaded from NCBI (https://www.ncbi.nlm.nih.gov/genome/genomes/2707). These assemblies 291 were aligned to the reference genome using minimap2 (version 2.10-r764, (43)) and its 292 293 "asm20" parameter preset. Only strains with at least 80% reference coverage were kept for subsequent analysis (9/11, average reference coverage: 91%). The paftools utility, 294 295 which is distributed with minimap2, was used to call variants from these alignments, with the parameter set "-I 2000 -L 5000". All resulting VCF files from modern samples were 296 merged using bcftools' merge utility with the parameter "--missing-to-ref", assuming that 297 298 those positions not called by paftools in any one sample were indeed reference calls. The merged VCF from modern material was then merged with the VCF from the two 299 300 historical samples using bcftools (version 1.8, (42)), and filtered to include only full information, biallelic SNPs. This approach discovers sites, which are segregating in 301 modern material, and have read data (be it reference, alternative or segregating sites) in 302 303 both historical samples. The resulting VCF file was loaded into R using vcfR (version 304 1.7.0, (44)), and a PCA was produced by converting the information into a genlight 305 object using adegenet (version 2.0.1, (45)) in R (version 3.3.3, (46)).

306 *Pseudomonas* spp. assembly and evaluation

To evaluate the presence of *Pseudomonas* spp. strains in a *Solanum tuberosum* historic 307 308 herbarium sample, we extracted from this library all reads that were taxonomically 309 assigned to the *Pseudomonas* genus or to inferior taxonomic levels within it. These reads were then assembled using SPAdes (version 3.5.0) with default parameters (47). 310 The resulting contigs were filtered for a minimum length of 2Kb, which yielded 3,314 311 contigs with a total length of 16Mb. We used the lastz (version 1.03.66, (48)) and Circos 312 313 (version 0.64, (49)) interface of AliTV (50) to align these contigs to either the P. syringae 314 or *P. rhizosphaerae* reference genome. We were able to align 72% of contigs to either one or both of these reference genomes in alignments of at least 1Kb. We then 315 316 extracted all contigs which had alignments of at least 10Kb in length and were unique to one of the reference genomes. These sets of contigs were again aligned to their 317 318 corresponding reference using AliTV as described above. Additionally, we used these 319 uniquely aligning contigs to assess their average kmer coverage during the assembly, 320 as reported by SPAdes.

321 Acknowledgements

322	We thank Verena Schuenemann for help in the laboratory; Patricia Lang, Claudia S.
323	Burbano, members of the Research Group for Ancient Genomics and Evolution, and
324	specially Talia Karasov for input on data analysis and comments on the manuscript;
325	Janet Kelso for help with processing the raw sequencing data; Ivan Gusic, Zeljko Kucan,
326	Carles Lalueza-Fox, Marco de la Rasilla, Antonio Rosas, Pavao Rudan, Sandra Knapp,
327	Bruce Benz, Michael Blake, R.G. Matson, Bryn Dentinger, Anna Stalter, Robert Capers,
328	John Peter for providing samples. This work was funded by the Max Planck Society and
329	its Presidential Innovation Fund.

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507 Supplementary Figures

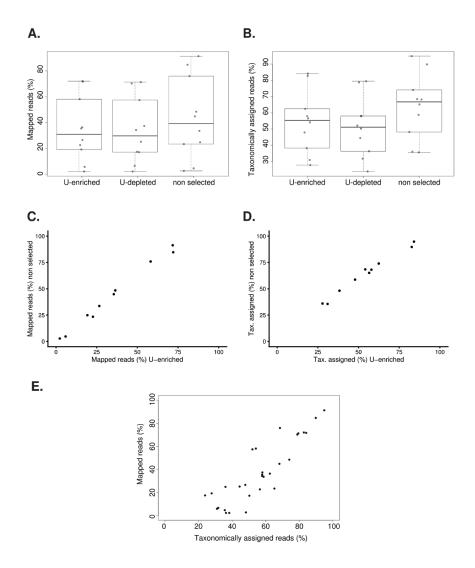


Figure S1. Mapped and taxonomically assigned reads of plant historical specimens. A. 508 Distributions of percentage of mapped reads for non-selected and U-selected libraries 509 (U-enriched and U-depleted fractions). **B.** Distributions of percentage of taxonomically 510 assigned reads for non-selected and U-selected libraries (U-enriched and U-depleted 511 512 fractions). C. Correlation of the percentage of mapped reads between the U-enriched and the non-selected library **D**. Correlation of the percentage of taxonomically assigned 513 reads between the U-enriched and the non-selected library E. Relation between 514 percentages of mapped and taxonomically assigned reads from U-selected libraries 515 (U-enriched fraction). 516

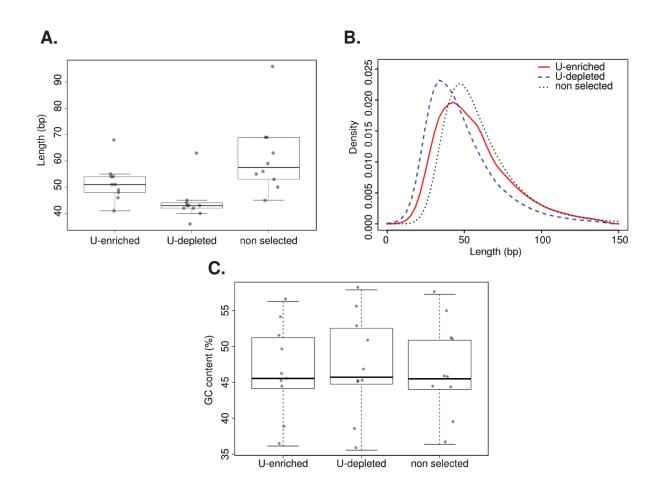


Figure S2. Length and GC content of plant historical specimens. **A.** Distributions of mean length for non-selected and U-selected libraries (U-enriched and U-depleted fractions). Median values are denoted as black lines and points show the original value for each individual sample. **B.** Length distribution of *Arabidopsis thaliana* sample NY1365375 for a non-selected and U-selected library (U-enriched and U-depleted fractions). **C.** Distributions of mean GC content for non-selected and U-selected libraries (U-enriched and U-depleted fractions).

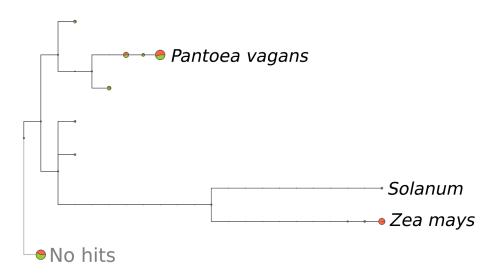


Figure S3. Taxonomic tree of reads from *Solanum tuberosum* and *Zea mays* assigned to different taxonomic levels. The size of the circle represents the amount of reads assigned to a particular part or the taxonomy. *S. tuberosum-* and *Z. mays-*derived reads are shown in green and orange, respectively.

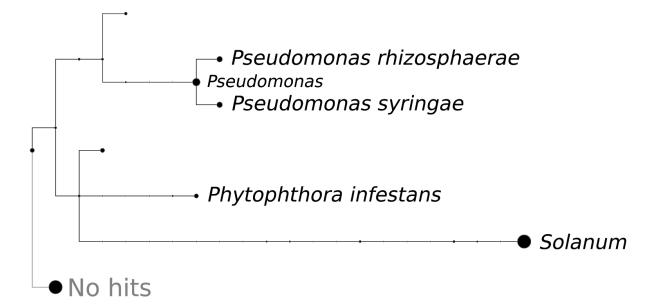


Figure S4. Taxonomic tree of reads from a *Solanum tuberosum* library assigned to different taxonomic levels. The size of the circle represents the amount of reads assigned to a particular part of the taxonomy. Reads assigned to some species *Phytophthora infestans, Pseudomonas syringae and Pseudomonas rhizosphere*, as well as the genera Pseudomonas and Solanum are named in the taxonomic tree.

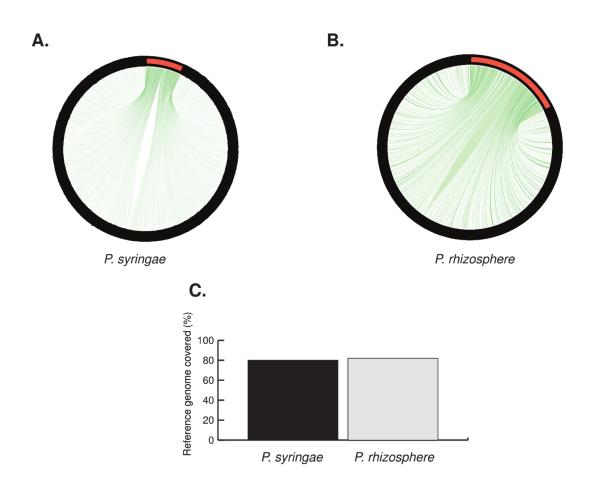
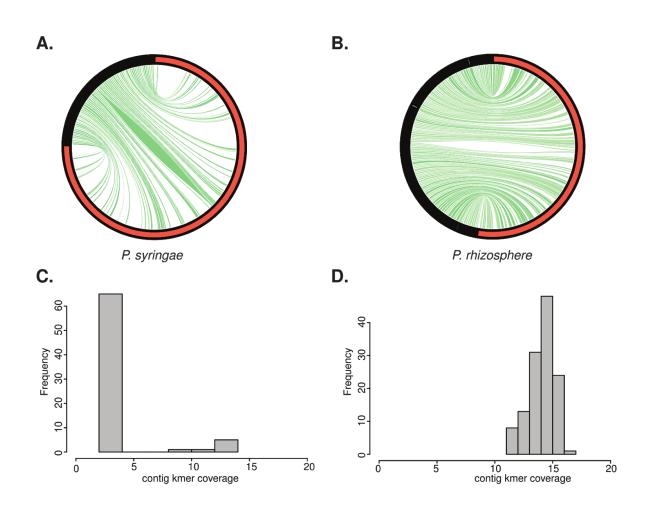


Figure S5. De novo assembly and genomic coverage of *Pseudomonas syringae* and *Pseudomonas rhizosphere* from a *Solanum tuberosum* sample. **A.** Alignments (represented as green lines) between all de novo assembly contigs (black) and *P. syringae* reference genome (red). **B.** Alignments (represented as green lines) between all de novo assembly contigs (black) and *P. rhizosphere* reference genome (red). **C.** Percentage of reference genome of *P. syringae* and *P. rhizosphere* covered by de novo assembled contigs from A. and B., respectively.



540 Figure S6. De novo assembly (uniquely mapped contigs) and contig k-mer coverage of Pseudomonas syringae and Pseudomonas rhizosphere from a Solanum tuberosum 541 sample. A. Alignments (represented as green lines) between de novo assembly contigs 542 (black) uniquely mapped to P. syringae reference genome (red). B. Alignments 543 (represented as green lines) between de novo assembly contigs (black) uniquely 544 mapped to P. rhizosphere reference genome (red). C. Histogram of contig k-mer 545 coverage from de novo assembled contigs uniquely mapping to P. syringae. D. 546 Histogram of contig k-mer coverage from de novo assembled contigs uniquely mapping 547 to P. rhizosphere. 548