

# <sup>1</sup> Tiled Amplicon Sequencing Enables Culture-free Whole-Genome Sequencing of Pathogenic Bacteria From Clinical Specimens

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## <sup>19</sup> Abstract

<sup>20</sup> Pathogen sequencing is an important tool for disease surveillance and demonstrated its high value during

<sup>21</sup> the COVID-19 pandemic. Viral sequencing during the pandemic allowed us to track disease spread,

<sup>22</sup> quickly identify new variants, and guide the development of vaccines. Tiled amplicon sequencing, in

<sup>23</sup> which a panel of primers is used for multiplex amplification of fragments across an entire genome, was

<sup>24</sup> the cornerstone of SARS-CoV-2 sequencing. The speed, reliability, and cost-effectiveness of this method

<sup>25</sup> led to its implementation in academic and public health laboratories across the world and adaptation to a

<sup>26</sup> broad range of viral pathogens. However, similar methods are not available for larger bacterial genomes,

<sup>27</sup> for which whole-genome sequencing typically requires *in vitro* culture. This increases costs, error rates

<sup>28</sup> and turnaround times. The need to culture poses particular problems for medically important bacteria such

29 as *Mycobacterium tuberculosis*, which are slow to grow and challenging to culture. As a proof of concept,  
30 we developed two novel amplicon panels for *Streptococcus pneumoniae* and *Mycobacterium tuberculosis*,  
31 which enabled recovery of whole bacterial genomes without culturing. Applying our amplicon panels to  
32 clinical samples, we show the ability to classify pathogen subgroups and to reliably identify markers of  
33 drug resistance. Development of this work in clinical settings has the potential to tailor disease  
34 interventions and treatment regimes for these high priority pathogens.

35

## 36 **Introduction**

37 In recent years, whole-genome amplicon sequencing has been adopted as a standard technique for  
38 genomic surveillance of infectious disease. Initially developed for genomic surveillance of the 2016 Zika  
39 epidemic [1], where low viraemia had precluded direct sequencing of clinical samples even where  
40 infection had been confirmed. Amplicon sequencing uses multiplex PCR of tiled overlapping regions of a  
41 target genome to recover whole genomes from samples of low concentration or complex backgrounds.  
42 This has proven particularly useful for sequencing remnant samples from diagnostic tests, and its use in  
43 the ‘artic’ protocol for sequencing SARS-CoV-2 [2] has led to it being deployed in thousands of public  
44 health laboratories around the world, facilitating true global surveillance of viral dynamics [3]. The ease,  
45 reliability and low cost of this approach has seen its adaptation to a broad range of viral pathogens both in  
46 respiratory disease [4,5] and beyond [6,7]. However, viral infections are by no means the only cause of  
47 global morbidity; more than half of all infectious disease-related deaths are caused by 33 bacterial  
48 pathogens, with *Streptococcus pneumoniae* alone estimated to be responsible for more than 800,000  
49 deaths per year [8].

50 As with viral pathogens, sequencing of bacteria can enable reconstruction of transmission chains for  
51 targeting interventions and routine surveillance of pathogen diversity for vaccine design or monoclonal  
52 antibody targeting. It can also be uniquely informative for the detection of drug resistance in bacteria,  
53 allowing insights into the horizontal transfer of antimicrobial resistance genes and potentially enabling

54 more effective tailored treatment regimes [9]. However, the relatively laborious process of isolating and  
55 culturing patient samples means this is typically only performed where a small number of samples can be  
56 highly informative, such as in outbreaks of food-borne pathogens [10–12], or drug-resistant nosocomial  
57 infections [13–15]. This kind of approach is far less appropriate for a bacterium with high rates of  
58 commensal and asymptomatic transmission such as *S. pneumoniae*, and it would be entirely  
59 cost-prohibitive in the 24 low- and middle-income countries (LMICs) in which it is the leading cause of  
60 death [8]. A disease such as tuberculosis (TB) has similarly high burden and low detection rates. In  
61 addition, *Mycobacterium tuberculosis* has low genomic diversity, making traditional approaches, such as  
62 restriction fragment length polymorphism (RFLP), spoligotyping, and variable number tandem repeat  
63 (VNTR) less precise [16]. A notoriously slow growth rate means it can take weeks to detect, let alone  
64 sequence, a TB infection [16]. Though whole genome sequencing (WGS) of *M. tuberculosis* has  
65 demonstrated clear application to detecting superspreading [17], or distinguishing recrudescence from  
66 reinfection [18], the difficulties of culturing *M. tuberculosis* means these studies have typically been  
67 performed retrospectively. The use of WGS to inform outbreak investigations as they occur has been  
68 limited to high-resource settings such as the United Kingdom [19]. Adapting the techniques used for viral  
69 pathogens to bacterial pathogens, enabling rapid culture-free sequencing from minimal input volumes and  
70 the use of remnant tests and other passive surveillance techniques, could be transformative for bacterial  
71 genomic epidemiology.

72

73 We present here the first use of amplicon-based WGS for the sequencing of two bacterial pathogens of  
74 major public health importance. We have designed tiling amplicon schemes for *S. pneumoniae* serotype 3  
75 and *M. tuberculosis*. These assays are able to generate complete genome coverage from samples with  
76 minimal input concentrations without any requirement for bacterial culturing. We show recovery of  
77 genomic data from a broad range of sample types, including saliva, sputum, nasopharyngeal swabs and  
78 remnant diagnostic tests, and further show that this genomic data can reliably perform *in-silico* lineage or  
79 serotype assignment to enable the surveillance of bacterial transmission dynamics. We show that our TB

80 amplicon panel can be applied directly to sputum samples to identify clinically relevant phenotypes such  
81 as antimicrobial susceptibility within days of sample collection, and can detect resistance loci that were  
82 not found by rapid diagnostics. We hope that this work will not only generate opportunities for future  
83 genomic epidemiology of *S. pneumoniae* and *M. tuberculosis*, but will also provide a roadmap for the  
84 development of amplicon sequencing for other clinically important bacterial pathogens.

85

## 86 Results

### 87 *In silico* predictions indicate broad applicability of amplicon schemes across clades

88 In order to design primer schemes with efficient amplification of diverse target sequences, we  
89 downloaded a selection of whole-genome sequences available on public repositories for both *M.*  
90 *tuberculosis* (n=489, **Supplemental file 1a**) and *S. pneumoniae* (n=490, **Supplemental file 1b**). For *S.*  
91 *pneumoniae*, we assembled these sequences into a ‘metaconsensus’ sequence, a reference-guided core  
92 genome with SNPs and indels replaced with ‘N’. PrimalScheme was run on the output of this, in order to  
93 design primers which cover the core *S. pneumoniae* genome and avoid variant sites. Because *M.*  
94 *tuberculosis* has very little within-species diversity outside of the repetitive hypervariable PE/PPE/PGRS  
95 regions, PrimalScheme was run directly on the H37Rv reference genome after masking PE/PPE/PGRS  
96 regions and sites with known resistance-related polymorphisms to avoid primers being designed at these  
97 loci.

98 For both pathogens, we selected a small number of genetically diverse sequences from the larger set of  
99 publicly-available sequences to predict coverage beyond the sequences used for primer panel design. As  
100 expected, predicted amplicon coverage was highest against the strains used for panel design (*Sp*:CC180:  
101 98.93%; *Mt*/H37Rv: 94.31%) - *M. tuberculosis* coverage is reduced due to omission of PE/PPE regions  
102 from the design, which account for 8-10% of the genome [20]. However, predicted coverage in *M.*  
103 *tuberculosis* remained high across all 7 lineages (>= 94.23%) and in the *M. canetti* outgroup (89.44%),  
104 while *S. pneumoniae* fell sharply across the clade (>=89.44) and in the *S. mitis* outgroup (32.18%)

105 (Figure 1). Average nucleotide identity fell less across the clade for *M. tuberculosis* (99.98-99.27%) than  
106 *S. pneumoniae* (98.76-92.51%), however pangenome size and similarity was markedly different between  
107 the two species, with *M. tuberculosis* having a smaller relative pangenome (4,335 total genes and a mean  
108 genome size of 4,067 genes, a ratio of approximately 1.07, compared to 3,942 total genes and a mean  
109 genome size of 2,071 genes for *S. pneumoniae*, a ratio of approximately 1.90) and, as a result, far more  
110 sharing of genes with the reference strain (4,002-4,050) than *S. pneumoniae* (1,705-1,793). Our findings  
111 suggest panel applicability is largely affected by genome rearrangement rather than increases in genetic  
112 distance.

113

114 **Amplicon sequencing enables recovery of whole genomes from diverse and minimal-input bacteria**  
115 **samples**

116 To determine our ability to enrich target genomes from within diverse sample sources, for *S. pneumoniae*,  
117 we sequenced DNA from cultured isolates, nasopharyngeal (NP) swabs, saliva samples, and  
118 culture-enriched saliva and NP swabs using our amplicon panel and standard metagenomic sequencing.  
119 For *M. tuberculosis*, we sequenced DNA from cultured isolates and sputum samples using the same  
120 amplicon sequencing workflow, with and without adding amplicon panel primers, as well as standard  
121 metagenomic sequencing.

122 For *S. pneumoniae*-positive samples, despite high species diversity in each sample type, increases in the  
123 proportion of reads mapping to the target genome were seen for both saliva and nasopharyngeal swabs  
124 compared to standard metagenomic sequencing, alongside an additional increase in related *Streptococcus*  
125 species (Figure 2a-b); this is particularly noticeable within saliva samples, which are expected to have  
126 high complements of *S. oralis* and *S. mitis*. *S. pneumoniae* read recruitment was high among cultured  
127 isolates regardless of amplification protocol.

128 Comparisons of amplified and unamplified *M. tuberculosis*-positive sputum samples demonstrated  
129 dramatic increases in coverage for amplified samples as compared to the same samples without  
130 amplification (Figure 2c-d). While only 2/10 of unamplified samples achieved more than 75% coverage,

131 9/10 of the amplified samples sequenced above this threshold, with 7 of those generating more than 95%  
132 coverage. The remaining sample achieved 33% coverage amplified, and negligible coverage unamplified.  
133 Metagenomic sequencing indicated lower overall species diversity for TB sputum samples, yet successful  
134 amplification from samples containing both commensal and pathogenic bacteria including *Streptococcus*,  
135 *Pseudomonas*, *Actinomycetes* and *Schaalia* spp.

136 We assessed the limits of detection for each amplicon panel by sequencing serial 10-fold dilutions of 6  
137 cultured samples of each bacteria using both amplified and unamplified sequencing approaches. For *M.*  
138 *tuberculosis*, high genome coverage (>95%) was observed in all amplified samples above 100 genome  
139 copies per microlitre (GC/µL), compared to 10,000 GC/µL for unamplified samples (**Fig S1**).

140

141 **Amplicon derived data enables phylogenetic classification and population delineation of *M.***

142 ***tuberculosis***

143 Lineages of *M. tuberculosis* were called with the Mykrobe package [21], which assigned all samples to  
144 lineages 2 (sublineage 2.2.1) or 4. Mykrobe performed equally well in high coverage samples, regardless  
145 of whether these were derived from cell culture or sputum. We derived maximum likelihood phylogenies  
146 using IQ-tree [22] including all sequenced specimens and the broad reference set of *M. tuberculosis*  
147 sequences used for primer design (**Supplemental Figure 2**; **Supplemental file 1a**). In all cases the  
148 primary lineage predicted by Mykrobe aligned with lineages from a maximum-likelihood tree, though in  
149 some cases secondary lineages were predicted based on minor variants which did not concord with the  
150 ML tree.

151 Lineage calling for *S. pneumoniae* was largely unsuccessful. Culture-derived samples could be assigned  
152 to serotype using either PneumoKITy [23] or PopPunk [24], however none of the lineage callers  
153 (PneumoKITy [23], PopPunk [24], SRST2 [25]) were able to assign lineages to any of the direct clinical  
154 samples from saliva or nasopharyngeal swabs. While our data did indicate coverage of the 7 major *S.*  
155 *pneumoniae* housekeeping genes (*aroE*, *gdh*, *gki*, *recP*, *spi*, *xpt*, *ddl*) lineage predictions may have been  
156 impaired by the high concentration of commensal bacteria in the enriched samples.

**158 Direct sputum sequencing for TB detects markers of antimicrobial resistance to first-line therapies**

159 For *M. tuberculosis*, sequencing data was high-quality enough to produce a prediction for all template

160 dilutions from cultured isolates with at least 10 GE/µL starting quantity (**Figure 3, Supplemental Figure**

161 **3**). While we do not have access to phenotypic susceptibility results for these isolates, predictions were

162 internally consistent for all template dilutions above 100 GE/µL (though there was some variability

163 between partial vs full resistance calling) with the exception of streptomycin. DNA was extracted directly,

164 without culturing, from 60 sputum specimens with a range of acid-fast bacilli semi-quantitative

165 measurements (e.g., 1+ to 3+); sequencing data was high-quality enough to produce a drug susceptibility

166 prediction for 53/60 sputum specimens. Of the 7 specimens which failed, (Yale-TB121, Yale-TB123,

167 Yale-TB149, Yale-TB150) had starting quantities (following extraction) below 10 GE/uL. None of the

168 other 3 (Yale-TB126, Yale-TB139, Yale-TB148) have GenXpert results available as comparison. Several

169 different extraction methods were used (detailed in **Supplemental table S1b**) as it was not clear what

170 method would perform best; all 20 specimens extracted with the final protocol, which included a

171 NALC-NaOH treatment to deplete non-mycobacterial DNA, had adequate data to predict resistance.

172

173 *In-silico* antibiotic resistance screening in *S. pneumoniae* identified resistance to several second-line and

174 broad-spectrum antibiotics, including Lincosamides, Macrolides, and Fluoroquinolones in 9/9 culture

175 isolate samples, 7/9 culture enriched samples, 14/15 saliva samples, and 0/3 nasopharyngeal samples.

176 Percent coverage and percent identity toward resistance genes (*RlmAII*, *patA*, *patB*, and *pmrA*) ranged

177 from 76.33% to 100% (mean = 94.85%) and 99.24% to 100% (mean 93.87%) for cultured isolates,

178 75.04% to 100% (mean 90.6%) and 78.43% to 88.93% (mean 83.72%) for culture-enriched samples, and

179 75.21% to 100% (mean 93.91%) and 82.84% to 99.38% (mean 92.57%) for saliva samples, respectively.

180 On average, samples contained at least 3 resistance genes (**Supplemental Table 3**). We identified several

181 virulence factors including capsular polysaccharides, many of which are associated with TIGR4 (Serotype

182 4) and *Streptococcus pyogenes*, suggesting prior capsular switching and horizontal gene transfer events,

183 highlighting the ability of amplicon sequencing to pick up on the genetic diversity and evolutionary  
184 adaptability of *S. pneumoniae*.

185

## 186 Discussion

187 Tiled amplicon sequencing of pathogens has proven extremely useful for reconstructing disease spread  
188 and gaining insight into transmission patterns for a variety of viruses [26]. The 2020 SARS-CoV-2  
189 pandemic stimulated a global effort to adopt these methods and use genomics to track and monitor the  
190 virus; however, it has not previously been applied to the significantly larger and often more complex  
191 genomes of bacteria. Our work here, in which we have successfully used a tiled amplicon approach to  
192 sequence two pathogenic bacteria from specimens with minimal input DNA and demonstrated the ability  
193 to identify clades and markers of drug resistance, could have a major impact on disease control for these  
194 two species.

195 Both *S. pneumoniae* and *M. tuberculosis* are pathogens of prime public health importance. *S. pneumoniae*  
196 is responsible for more than 800,000 deaths per year, with the majority of these resulting from respiratory  
197 tract infections in children under five [8], and vaccine design is guided by ongoing genomic sequencing  
198 [27]. Prior to the Covid-19 pandemic, TB was the world's leading cause of death from a single infectious  
199 agent, causing more than a million deaths per year [28]. Despite the availability of vaccines, treatment,  
200 and significant funding [29], we continue to miss WHO targets for reductions in TB incidence and death  
201 by wide margins, indeed cases have risen worldwide over the past 2 years [30].

202

203 Antimicrobial resistance is a critical issue in treating and controlling TB, due to the prevalence of  
204 resistance to first line drugs and the length, cost, and complexity of treatment regimes [31]. Despite the  
205 introduction of shorter regimens, the time taken to find an effective treatment can be long, and incomplete  
206 treatment remains a problem [32]. For this reason, the WHO now recommends the use of targeted  
207 sequence-based diagnostics for rapid drug susceptibility testing for patients who are at high risk of, or

208 have already experienced treatment failure [33]. However designing such an assay is not simple; more  
209 than 40 separate loci, each containing numerous individual mutations, have been implicated in drug  
210 resistance [34], and uncertainty can be higher for new or second line drugs [35]. Whole-genome  
211 sequencing works around these limitations of targeted amplicon sequencing. As data are being generated  
212 across the entire genome, drug susceptibility predictions can be improved and expanded bioinformatically  
213 as new genetic markers are discovered without updating primers, unlike existing targeted  
214 sequencing-based diagnostics.

215 The required time, infrastructure, and costs for tiled amplicon sequencing are almost identical to targeted  
216 amplicons; the additional data generated through WGS can be used along with phenotypic drug  
217 susceptibility to expand our understanding of the genetic markers of drug resistance, especially for  
218 third-line or novel drugs, increasing the accuracy of predictions over time [20].

219 Whole genome sequencing obviates the need to design a targeted assay and can also return resistance  
220 predictions within days of a positive culture. However, the requirement of most existing WGS approaches  
221 to first grow a culture sample means that the overall sample-to-sequence turnaround time for *M.*  
222 *tuberculosis* is measured in weeks or months [19] and significant biases can be introduced during the  
223 culturing process itself [20]. Direct WGS without culture does not consistently produce data of high  
224 enough quality for resistance prediction or thorough epidemiologic investigation [36,37], is limited to  
225 specimens with a high bacterial load [37], or relies on expensive techniques such as hybrid capture [38].  
226 We have demonstrated tiled amplicon sequencing directly from sputum specimens, without culture, can  
227 be used to make accurate drug susceptibility predictions and lineage assignments for the majority (53/60)  
228 of specimens, unlike prior whole-genome sequencing approaches [19,36–39]. For a notoriously  
229 slow-growing organism such as *M. tuberculosis*, eliminating this step reduces the time from sample  
230 collection to genome from weeks to days. Not only could patients receive appropriate antibiotics sooner,  
231 but genomic epidemiology could be used in real-time to inform outbreak investigations [40,41] and public  
232 health measures to reduce spread [42,43].

233

234 Despite more consistent coverage across *in vitro* and *in silico* predictions, gaps remain in our coverage of  
235 the *M. tuberculosis* genome in the PE/PPE regions. While these are frequently omitted from *M.*  
236 *tuberculosis* analyses, increasing evidence of functions in host cell invasion [44] and importance for  
237 vaccine design [45] suggest inclusion of these regions in future iterations of this amplicon panel would be  
238 a significant improvement.

239

240 Nevertheless the comparison between our results for *S. pneumoniae* and *M. tuberculosis* is instructive.  
241 Neither panel showed high rates of amplicon dropout when faced with targets which had drifted from the  
242 reference strain (a regular issue with viral amplicon panels). However, *in silico* predictions suggest a  
243 weaker applicability of the amplicon panel in species which undergo significant levels of recombination  
244 and horizontal gene transfer, and our inability to reliably recover serotype and resistance loci in *S.*  
245 *pneumoniae* supports this conclusion. Indeed, *S. pneumoniae* exhibits extremely high levels of horizontal  
246 gene transfer, not only within the species, but also with frequently co-occurring commensal bacteria such  
247 as *S. mitis* and *S. oralis* [46,47]. This species may simply not be a suitable target for this approach, where  
248 metagenomic or hybrid capture-based sequencing may be more appropriate.

249

250 Faced with both drift and genomic rearrangement, designing primers that target conserved motifs will rely  
251 upon databases of previously sequenced genomes to allow us to determine circulating genetic diversity.  
252 Rapid improvements in sequencing and assembly technology have generated vast databases of assembled  
253 genomes; while these resources are not comprehensive, their bias towards improved representation of  
254 species of clinical interest [48] suggests this will not be a limiting factor in panel design.

255 An alternative consideration is targeting bacteria which do not undergo significant levels of horizontal  
256 gene transfer, and the ratio of genome to pangenome size is likely to be a key metric for our ability to  
257 design an amplicon panel. This ratio is highly sensitive to the diversity of habitats in which the pathogen

258 is found: free living or commensal species gain particularly large pangenomes to enable adaptation to  
259 diverse environments; intracellular pathogens show strong purifying selection, low effective population  
260 sizes and low genome:pangome ratios [49]. *M. tuberculosis*, an obligate pathogen and intracellular  
261 bacterium which has been extensively sequenced [50], has little horizontal gene transfer, and remains a  
262 major threat to human life [30], may be archetypal, however other intracellular pathogens such as *Yersinia*  
263 *pestis*, *Listeria monocytogenes*, *Legionella pneumophila*, and *Chlamydia trachomatis* are suitable targets.

264

265 The widespread use of tiled amplicon sequencing for pathogen genomics during the Covid-19 pandemic  
266 has ensured that this method is trusted, understood, and easily implemented in academic and public health  
267 laboratories worldwide. As the focus now turns to adapting this capacity to other public health threats [3],  
268 it is important to prioritize the development of tools for global priority pathogens that can be implemented  
269 in the regions suffering the greatest burden. Genomic surveillance of TB has demonstrated capacity to  
270 guide TB interventions in high income countries [17,18]; the reductions in cost and turnaround time  
271 afforded by tiled amplicon sequencing could enable this to be implemented in LMICs with high TB  
272 burden. Just four countries (India, Bangladesh, Indonesia, Democratic Republic of the Congo) account for  
273 over half of all TB deaths; all have seen prior in-country amplicon sequencing of SARS-CoV-2 [51–54]  
274 suggesting a ready capacity for tiling amplicon sequencing of *M. tuberculosis*. Extensive use of  
275 alternative sequencing methods such as Oxford Nanopore in these regions [51,53,55] suggest adaptation  
276 to cheaper and more portable sequencing platforms may further increase surveillance capacity.  
277 Barriers to clinical application are necessarily higher [56]; if diagnostics and resistance prediction are to  
278 be used to tailor treatment regimes it is vital that they can be shown to work reliably in a range of likely  
279 scenarios: paucibacillary infections; mixed *M. tuberculosis* strains; mixed *M. tuberculosis* and  
280 non-tuberculosis mycobacteria; partial and incomplete resistance. Despite this complex landscape, the  
281 capacity of culture-free *M. tuberculosis* sequencing to allow early diagnosis and resistance detection could  
282 be transformative. Not only by increasing completion rates of TB treatment at the patient level [57], but  
283 by preventing the further transmission of MDR-TB at the population level [58,59]. A comprehensive

284 evaluation of culture-free sequencing methods in a clinical environment should be a priority for TB  
285 control.

286

## 287 Materials and Methods

### 288 Ethics statement

289 All specimens were discarded and de-identified specimens used previously for diagnostic testing or  
290 IRB-approved human subjects research in accordance with Yale University IRB-exempt protocol  
291 #2000033281. *S. pneumoniae* specimens were remnant specimens collected from study participants  
292 enrolled and sampled in accordance with the Yale University Humans Investigation Committee-approved  
293 protocol #2000027690. *M. tuberculosis* specimens from Moldova were remnant specimens collected from  
294 study participants enrolled and sampled in accordance protocol #2000023071 approved by Yale  
295 University Human Investigations Committee and the Ethics Committee of Research of the  
296 Phthisiopneumology Institute in Moldova. *M. tuberculosis* specimens from Peru were remnant specimens  
297 collected from study participants enrolled in accordance with protocol #204749 approved by the  
298 Institutional Committee on Research Ethics at Cayetano Heredia University, Peru.

### 299 Primer design

300 We downloaded all available *S. pneumoniae* Serotype 3 contigs (n=490; **Supplemental file 1b**) from the  
301 Global Pneumococcal Sequencing (GPS) database [60] on 02FEB2023. We downloaded raw reads for *M.*  
302 *tuberculosis* sequences from a previously described globally representative dataset [50] (n=489;  
303 **Supplemental file 1a**) from the European Nucleotide Archive (ENA) at EMBL-EBI. For both targets, we  
304 downloaded complete reference genomes from the National Center for Biotechnology Information  
305 (NCBI) GenBank (OXC141; accession NC\_017592 and H37Rv; accession NC\_000962.3).  
306 For *M. tuberculosis*, variants were called against the reference using Snippy and time-resolved  
307 maximum-likelihood tree was built using our variant call file, along with sample data generated from  
308 Augur (v.22.4.0), IQ-Tree (v.2.23), and TreeTime (v.0.10.1). Representative sequences (n=6) were

309 selected from across this tree using Parnas (v.0.1.4), to cover >50% of the expected overall diversity. We  
310 used these representatives to create an *M. tuberculosis* core genome assembly using Snippy.

311 For *S. pneumoniae*, consensus genome sequences were generated (n=4) with Snippy (v.4.6.0).

312 Tiled primer schemes (target amplicon size 2kb) were designed for both *S. pneumoniae* and *M.*  
313 *tuberculosis* (excluding PE/PPE and repeat regions) using PrimalScheme [1]. Primers were ordered at  
314 100uM and 200uM in IDTE for *S. pneumoniae* and *M. tuberculosis*, respectively. Primer pools consisted  
315 of an equal volume of each primer and were used for amplification without further dilution.

### 316 Clinical specimens

317 *S. pneumoniae* samples consisted of DNA extracted from raw saliva (15), nasopharyngeal swabs in viral  
318 transport media (VTM) (6), culture-enriched bacteria (16), and cultured pure isolates (9). All saliva  
319 specimens had a paired cultured specimen cultured from the saliva (either culture enriched or cultured  
320 isolate); six also had a paired nasopharyngeal swab collected simultaneously from the same patient, three  
321 of which were sequenced with and without amplification. A full list of *S. pneumoniae* samples and  
322 descriptions can be found in **Supplemental Table S1a**. DNA was extracted from 200uL of each sample  
323 using the MagMAX Ultra viral/pathogen nucleic acid isolation kit (Thermo Fisher Scientific) using a  
324 KingFisher Apex instrument (Thermo Fisher Scientific) and quantified using two qPCR primer/probe  
325 pairs, *lytA* [61] and *piaB* [62] as described previously [63].

326 *M. tuberculosis* samples consisted of DNA extracted from positive solid or liquid cultures from sputum  
327 and DNA extracted directly from sputum specimens. Extracts from culture consisted of remnant  
328 specimens from a prior study in Moldova, where sputum specimens were tested at a number of diagnostic  
329 centers in Moldova by microscopy, Xpert, and culture and positive cultures sent to the National TB  
330 Reference Laboratory in Chisnau for extraction by the cetyltrimethylammonium bromide (CTAB) method  
331 as described previously [42]. Extracts from sputum consisted of specimens collected in Peru after routine  
332 diagnostics had been carried out and TB confirmed. In order to test the efficiency of different methods for  
333 extracting DNA from sputum, each specimen was split into two and processed with two different  
334 protocols. A total of 30 unique sputum specimens were processed with two protocols each, and a total of

335 6 different protocols were tested. A full list of all *M. tuberculosis* samples and the extraction methods  
336 used can be found in **Supplemental Table S1b**, and a detailed description of extraction methods can be  
337 found in **Supplemental Methods 1**. Following extraction, DNA was quantified with a mycobacterium  
338 tuberculosis-complex specific, fluorescence-based real-time PCR assay on the Bio-Rad CFX96  
339 instrument [64].

### 340 **Metagenomic sequencing**

341 For *S. pneumoniae*, 1-3ng of each sample (up to 4uL for samples which were undetectable) and a negative  
342 template control (4uL H<sub>2</sub>O) underwent tagmentation for 5 minutes followed by a magnetic bead cleanup.  
343 Then, samples were amplified with Nextera dual-index adapters followed by a second magnetic bead  
344 cleanup. Each sample was quantified with a Qubit fluorometer and 5ng of each library were pooled  
345 together (up to 4uL for undetectable samples). The pooled libraries underwent a final 0.7X bead clean up,  
346 then were quantified on a Qubit fluorimeter and quality and fragment distribution verified using an  
347 Agilent Bioanalyzer. For *M. tuberculosis*, samples were prepared as described for amplicon sequencing,  
348 but with the addition of sterile water in place of PCR primer pools for both amplification reactions.

### 349 **Amplicon sequencing**

350 Amplicon DNA was prepared using the Illumina COVIDSeq DNA prep kit with primer pools for either *S.*  
351 *pneumoniae* or *M. tuberculosis* alongside a negative template control as performed previously [65].  
352 Template DNA from each specimen was amplified in two separate PCRs, one reaction for each primer  
353 pool. For each sample, equal amounts of each PCR product were combined and the 2kb amplification  
354 products underwent tagmentation for 3 minutes followed by a bead cleanup and library amplification with  
355 Illumina index adapters. Equal volumes of the fragmented and indexed library for each sample was  
356 pooled, followed by size-selective bead cleanup for DNA fragments between 300-600 bp. The final  
357 pooled library was quantified with a Qubit fluorometer and dsDNA High-Sensitivity Assay kit, and the  
358 fragment distribution verified on an Agilent Bioanalyzer and high-sensitivity DNA kit. Pooled libraries  
359 were sequenced on an Illumina NovaSeq (paired-end 150) with an average of 10 million reads per library.

**360 Alignments & Calling**

361 Reads were aligned to the appropriate reference (*S. pneumoniae*: CC180 (Serotype 3); *M. tuberculosis*:  
362 H37Rv) using BWA-MEM (v.2.2.1) [66] and SAMtools (v1.15.1) [67]. Amplicon sequencing data were  
363 filtered (using defaults; Q>20 over a sliding window of 4, minimum read length 50% of the average  
364 length). TB primer sequences were trimmed using iVar (v.1.4.2) [68]. Metagenomic sequences were  
365 trimmed and filtered for quality and length (<100bp), using Trim Galore (v.0.6.10) [69]. Variants were  
366 called and filtered (Phred score Q>10 and read depth >10) using BCFtools [70]. Read subsampling, depth,  
367 and coverage was calculated using SAMtools [67]. Raw reads were directly submitted to the CZID  
368 mNGS Illumina pipeline [71] for microbial composition characterization within samples. Further data  
369 analyses and visualizations were carried out in Rstudio (v.2024.04.2+764) [72] using the tidyverse suite  
370 (v.2.0.0) [73].

**371 Off-target amplification prediction**

372 For each amplicon panel, off-target amplification was assessed *in-silico* against a set of related genomes.  
373 For each species we compiled a genome cluster consisting of the reference genome, 12 representative  
374 near-neighbor genomes, and an outgroup (**Supplemental Table 2**). The pangenome for each cluster was  
375 calculated using Roary (v.3.13.0) [74] and assembled a maximum-likelihood (ML) phylogeny using  
376 FastTree (v.2.1.11) [75]. Average nucleotide distance was calculated between out references and all other  
377 genomes in the cluster using FastANI (v.1.34) [76]. Off-target amplification was inferred by primer  
378 alignment using Bowtie (v.1.3.1) [77]; amplicons were predicted for any properly oriented amplicon pairs  
379 within 2,200 bp.

**380 Serotyping, lineage assignment, and resistance prediction**

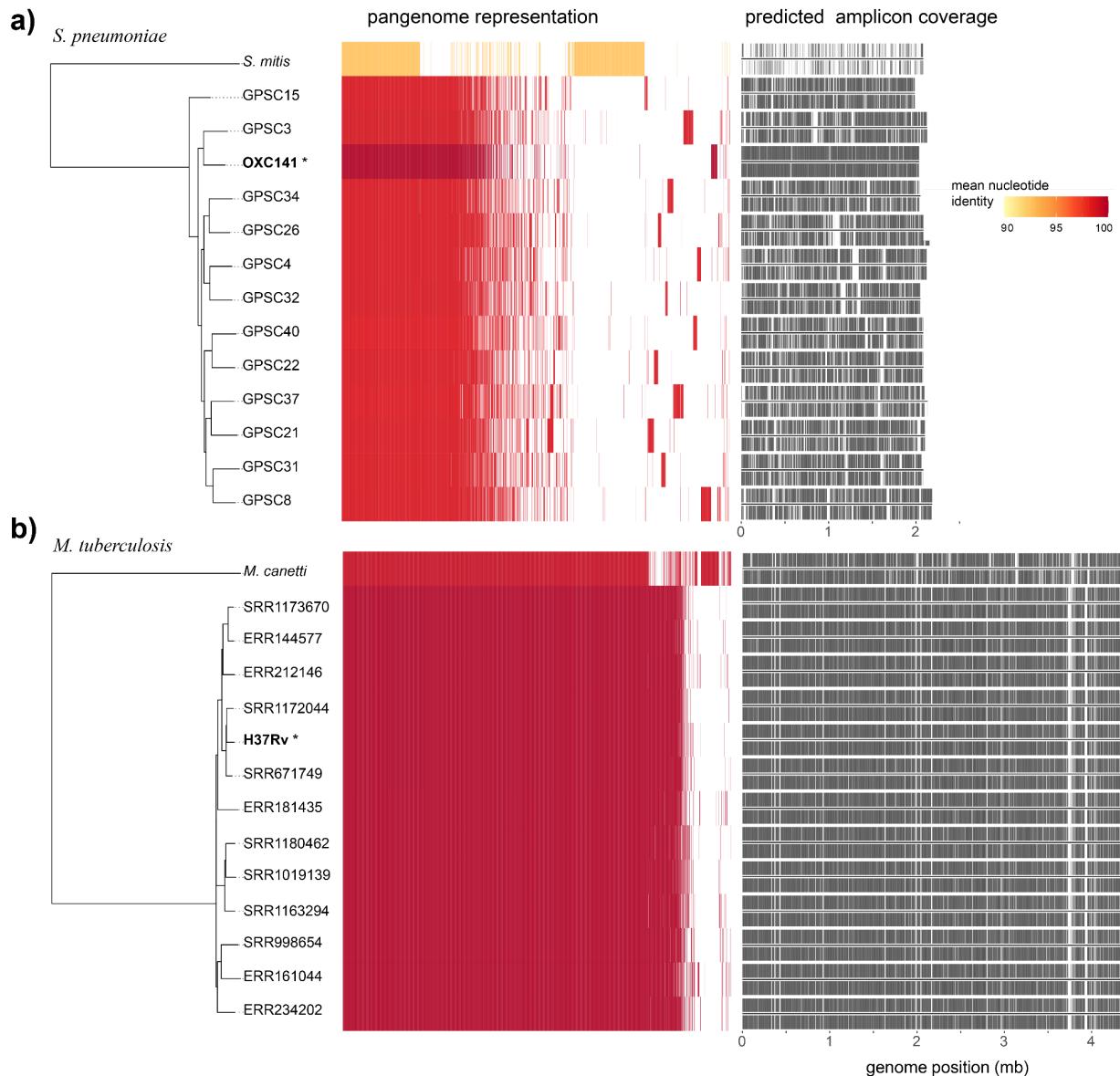
381 *S. pneumoniae* isolates were *de novo* assembled with Shovill (v.1.1.0) [78]. *In-silico* multi-locus  
382 sequencing types (MLST) were assigned with mlst (v.2.23.0) [79]. Global pneumococcal sequencing  
383 clusters (GPSCs) were assigned with poppuk (v.2.7.0) [24]. *In-silico* screening of contigs for *S.*  
384 *pneumoniae* antimicrobial and virulence genes was done using ABRicate (v1.0.1) [80] and appropriate  
385 AMR databases [81–84]. For *M. tuberculosis*, Mykrobe [21] was used to both assign lineages and predict

386 resistance using the built-in panel “202309” for tuberculosis [85]. As a comparison, a time-resolved  
387 maximum-likelihood tree was built using our variant call file, along with sample data generated from  
388 Augur (v.22.4.0), IQ-Tree (v.2.23), and TreeTime (v.0.10.1). Tree visualisations were done using Auspice  
389 (v2.57.0).

390

391 **Figures**

392 **Figure 1: *In silico* modeling indicates broad applicability across diverse TB clades**



394 Pangenome representation of (A) *S. pneumoniae* whole genome sequences (n=13) and *S. mitis* outgroup

395 (Accession: AP023349) and (B) *M. tuberculosis* whole genome sequences (n=13) and *M. canetti* outgroup

396 (Accession: NC\_019950). Starred phylogenetic tree tips mark the reference sequences used for primer

397 design. Shaded bar graphs (middle) denote genes shared amongst clades, color denotes average nucleotide

398 identity. Predicted amplicon coverage (right) is shown in grey with forward and reverse amplicon pairs

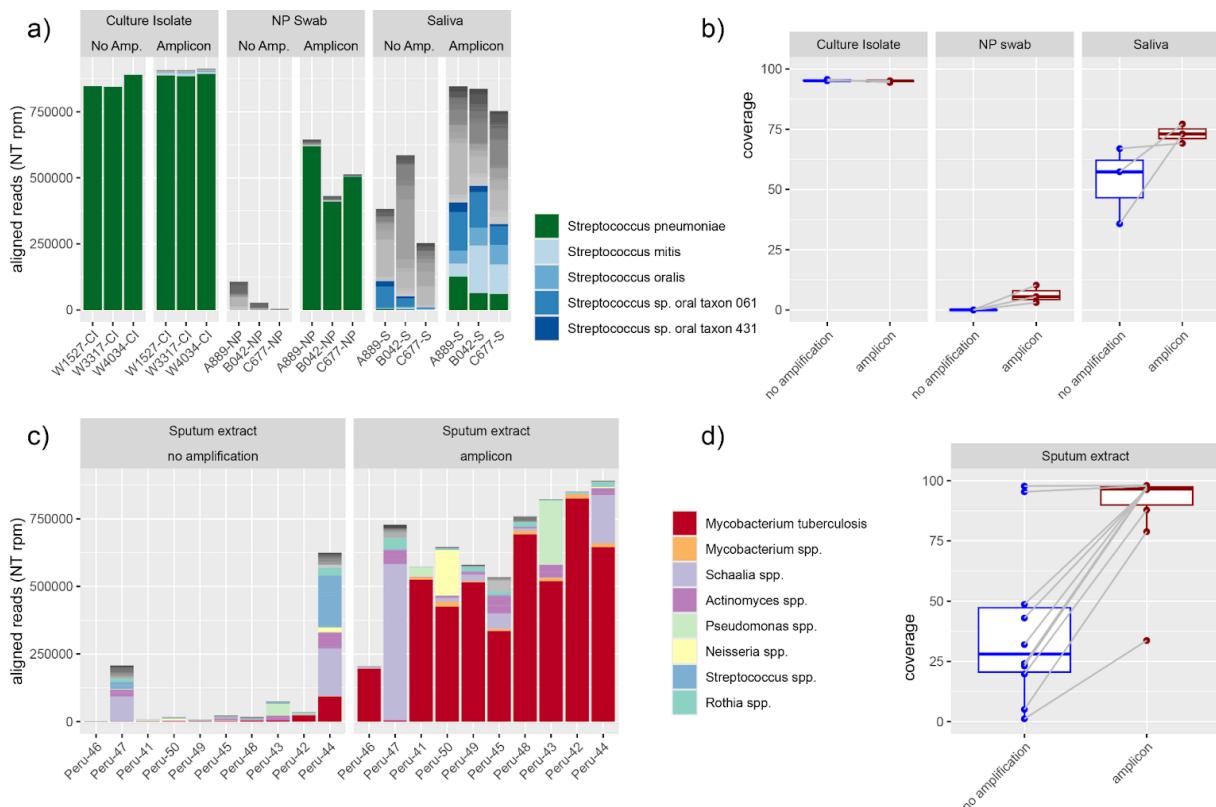
399 displayed above and below the line. A table of the sequences used in this analysis can be found at

400 **Supplemental Table 2.**

401

402

403 **Figure 2: Tiled amplicon sequencing enables recovery of whole genome sequences from TB sputum**



404

405 Comparisons between amplified and unamplified clinical samples were made for both species with regard

406 to metagenomics (a,c), via the CZID metagenomics pipeline, and overall genome coverage (b,d). *S.*

407 *pneumoniae* samples from multiple sample types from matched patients (a-b) showed increases in

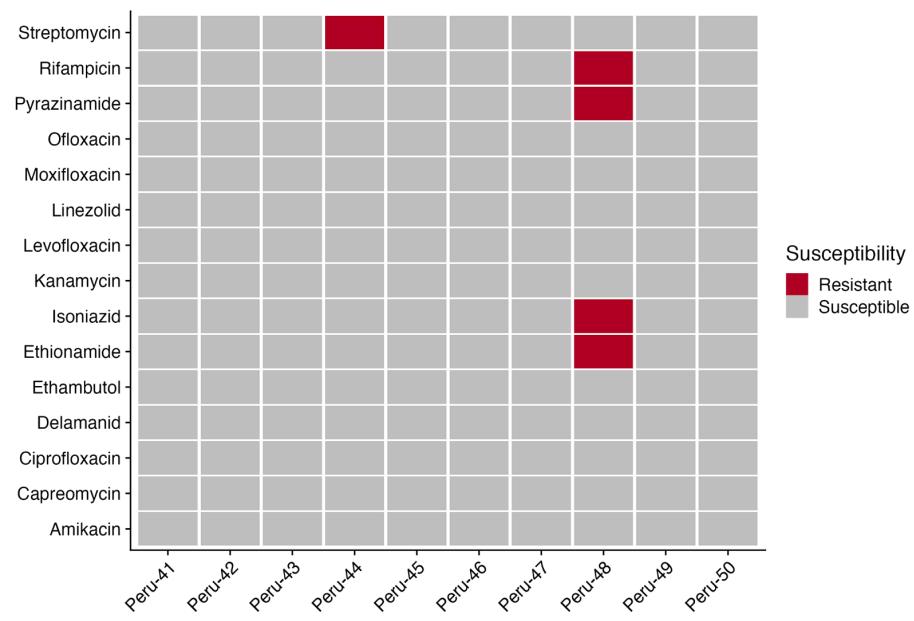
408 genome coverage and depth for all sample types, despite simultaneous amplification of closely related

409 taxa. *M. tuberculosis* samples taken from direct sputum sequencing (c-d) show dramatic increases in

410 genome coverage, with 8/10 samples generating more than 80% coverage after amplification with our

411 protocol, and a ninth sample generating 78% coverage despite a significant infection with *Schaalia*

412 *odontolytica*.

414 **Figure 3: Amplicon sequencing predicts TB antimicrobial resistance *in-silico*.**

416 Predicted susceptibility to 15 anti-TB drugs by amplicon sequencing for DNA extracted from sputum

417 without prior culture using our optimised extraction protocol, showing detection of Streptomycin and

418 combined Rifampicin / Isoniazid resistance.

421 **Funding statement**

422 This publication was made possible by the New England Pathogen Genomics Center of Excellence (US

423 CDC NU50CK000629); The National Heart, Lung, and Blood Institute of the National Institutes of

424 Health and the Richard K. Gershon Endowed Medical Student Fellowship at Yale University School of

425 Medicine.

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