

Real-time sequencing of TB: are we there yet?  
Commissioned manuscript with tracking number JCM00358-17

1 Real-time sequencing of *Mycobacterium tuberculosis*: are we there yet?

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13 Running head: Real-time sequencing of TB: are we there yet?

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16 Word count: Abstract 103, text 2671

17 **Abstract**

18 Whole genome sequencing has taken a leading role in epidemiologic studies of tuberculosis,  
19 but thus far, the real-time, clinical utility has been low, in part due to the requirement for  
20 culture. In Votintseva *et al.* (J Clin Microbiol <https://doi.org/10.1128/JCM.02483-16>),  
21 authors present a new method for extracting *Mycobacterium tuberculosis* DNA directly from  
22 smear-positive respiratory samples, generating drug resistance predictions and phylogenetic  
23 trees in 44 hours using the Illumina MiSeq. Authors also illustrate the potential for < 24 hour  
24 turnaround time from DNA extraction to clinically-relevant results, with Illumina MiniSeq  
25 and Oxford Nanopore Technologies' MinION. We comment on the promise and limitations  
26 of these approaches.

27

28

29 Whole genome sequencing (WGS) is becoming a mainstay in epidemiologic studies of  
30 tuberculosis (TB), with numerous studies demonstrating increased resolution of transmission  
31 compared to classical molecular typing methods (reviewed in (1)). While accurately  
32 resolving transmission networks is important for public health, it should be noted that  
33 countries with the greatest burden of TB often lack resources to conduct such epidemiologic  
34 investigations. In such high-incidence, resource-limited settings, diagnosis and treatment  
35 remain the greatest priority for TB control. A role for WGS in clinical management of TB has  
36 been recently discussed in (2, 3); as comparably inexpensive and rapid tests already available  
37 for detection of *Mycobacterium tuberculosis*, the application to drug resistance prediction is  
38 where WGS may ultimately have the greatest impact. We have therefore focused this  
39 commentary on this facet of care.

40

41 Inadequate access to drug susceptibility testing (DST), and long delays in reporting of drug  
42 resistance pose serious challenges to TB control. Phenotypic DST is currently the “gold  
43 standard” for determining resistance to antibiotics, but can take weeks to months for results  
44 as specimens must be sent to centralized laboratories for culture. Given these delays, rapid  
45 molecular tests such as Xpert MTB/RIF (Cepheid Inc., Sunnyvale, California, USA) and line  
46 probe assays (LPAs) have been endorsed by the World Health Organization, to complement  
47 culture-based DST. Although Xpert MTB/RIF and LPAs have enhanced early identification  
48 of drug resistance, these tests can miss resistance-conferring mutations that are outside of the  
49 locus of interest (e.g., the *rpoB* I491F mutation found to cause rifampin resistance is not  
50 detected by Xpert MTB/RIF (4)). They are also unable to differentiate between non-  
51 synonymous mutations and synonymous mutations (5), the latter of which are not thought to  
52 cause resistance, resulting in false positive predictions of resistance.

53

54 In contrast to these targeted, rapid molecular tests, WGS can theoretically identify all  
55 mutations in genes associated with resistance at once, and predict the functional effect of  
56 these mutations, i.e, whether or not they would likely cause antibiotic resistance. Numerous  
57 mutations in the *pncA* gene and inconsistent, unreliable DST (6) suggest that WGS is the best  
58 approach for diagnosing resistance to the drug pyrazinamide. This drug is not only a  
59 cornerstone of first-line regimens, but also is critical to novel drug combinations (e.g.,  
60 “BPaZM”: bedaquiline, pretomamid, pyrazinamide and moxifloxacin) in late phase trials (7).  
61 In addition, WGS offers the potential to detect novel drug mutations; this is of particular  
62 interest as new antimicrobials become available and are included in treatment regimens for  
63 multi-drug resistant TB (MDR-TB) (7). Using WGS, several resistance-connoting mutations  
64 have already been identified for bedaquiline and delamanid (8). With the potential to  
65 diagnose and predict drug resistance (as well as delineate transmission networks for public  
66 health) all in a single ‘test’, WGS therefore has the potential to revolutionize TB control  
67 programs. To do so, several obstacles must first be overcome in terms of speed (i.e., reducing  
68 the turnaround time (TAT) to results), accuracy of predicting phenotypic resistance, and  
69 resources requirements.

70

71 First, to have the greatest clinical utility, and meet the needs outlined by key stakeholders in  
72 the TB community (9), TAT needs to be fast – ideally, with results available the same day as  
73 testing. Due to several technical difficulties with performing WGS on clinical samples  
74 (reviewed in (10)), thus far, WGS has been predominantly been conducted on *M. tuberculosis*  
75 DNA extracted from culture. Using this approach, a head-to-head comparison (11) between  
76 WGS and conventional DST found no difference in TAT when real-life delays (as would be  
77 experienced in clinical practice) were considered; median TAT from positive Mycobacterial  
78 Growth Indicator Tube (MGIT) to WGS-based resistance reports was 31 days (IQR 21-60),

79 while median time from positive MGIT to DST report was 25 days for conventional DST  
80 (interquartile range 14-32). To reduce losses to follow-up, and potentially have an immediate  
81 impact on patient care, WGS would ideally be applied directly to clinical samples, as is done  
82 with Xpert MTB/RIF.

83

84 Thus far, only two studies have been published attempting WGS on such samples (12, 13),  
85 with variable success. In a proof-of-concept study published in 2015, Doughty *et al.* extracted  
86 *M. tuberculosis* DNA directly from clinical samples and sequenced this using Illumina MiSeq  
87 (Illumina, San Diego, USA); however, while TB could be diagnosed, the *M. tuberculosis*  
88 DNA obtained was insufficient for drug resistance prediction due to contamination with  
89 human DNA (12). A second study published the same year by Brown *et al.* (13) performed  
90 targeted enrichment using oligonucleotide baits to capture *M. tuberculosis* DNA prior to  
91 sequencing with Illumina MiSeq; this approach yielded  $\geq 20\times$  depth of coverage (the average  
92 number of sequencing 'reads' that map to a given position in the reference genome) and  
93  $>98\%$  genome coverage (the percent of the reference genome with at least one read mapped  
94 to it) for 20/24 smear-positive, culture-positive samples (83%). These quality control  
95 parameters are in line with early studies that performed WGS on DNA extracted from  
96 culture, and were sufficient for speciation and drug resistance prediction. However, this  
97 method was expensive and had technical requirements potentially beyond the capacity of  
98 most microbiology laboratories.

99

100 In this issue of the *Journal of Clinical Microbiology*, Votintseva *et al.* (14) present a new  
101 approach for extracting *M. tuberculosis* DNA from direct respiratory specimens, without  
102 enrichment. Compared to (13), this method was faster (TAT including MiSeq sequencing  
103 was estimated at 50h for Brown *et al.*, 44h for Votintseva *et al.*) and less expensive (203 vs

104 96 Great Britain Pounds per sample for reagents, extraction and sequencing, respectively).  
105 All direct samples were correctly identified as *Mycobacterium tuberculosis* complex (95%  
106 speciated as *M. tuberculosis*). Overall quality control metrics were lower compared to (13),  
107 with depth of coverage >12x and  $\geq 90\%$  genome coverage for only 21/37 (57%) of the smear-  
108 positive, culture-positive samples. As direct samples were partitioned for routine clinical use  
109 prior to analysis, it is possible this influenced the quantity of *M. tuberculosis* DNA available;  
110 however, sample volume was not associated with yield of DNA in univariate and multivariate  
111 models.

112

113 In addition to the Illumina MiSeq, the authors also tested and timed two other sequencing  
114 technologies: the Illumina MiniSeq and the Oxford Nanopore MinION (Oxford Nanopore  
115 Technologies, Oxford, UK). For these platforms, DNA was extracted from cultured  
116 *Mycobacterium bovis* BCG strain using a previously-reported protocol (15). Pure BCG was  
117 then sequenced as well as BCG DNA that was experimentally spiked at specified  
118 concentrations (ranging from 5%-15%) into smear-negative, culture-negative sputum. A PCR  
119 amplification step, using a new protocol developed by the authors, was performed prior to  
120 sequencing with the MinION. The TAT from DNA extraction to complete resistance  
121 prediction (and phylogenetic placement) was 16h for MiniSeq, while TAT to these results for  
122 a single sample of *M. bovis* BCG DNA spiked at 15% concentration was estimated at 12.5h  
123 with the MinION 9.4 flowcell (actual sequencing was performed over 48 hours in total). An  
124 advantage of the MinION is that data can be analyzed in real-time; *M. bovis* BCG DNA was  
125 detected and correctly identified at 1h, while sequencing was still ongoing. Somewhat  
126 disappointingly, authors did not test their novel extraction protocol for direct sequencing of  
127 *M. tuberculosis* in combination with either the MiniSeq or MinION. While authors estimate  
128 potential TAT for *M. tuberculosis* with the MinION R9.4 flowcell based on the single *M.*

129 *bovis* BCG run, to properly assess the performance of the direct sample DNA extraction  
130 protocol in combination with this technology, future studies are needed that use real (rather  
131 than experimentally-generated) respiratory samples, obtained from patients positive for *M.*  
132 *tuberculosis*. Nonetheless, this study serves as a valuable proof of concept, demonstrating the  
133 potential for WGS as a same-day test.

134

135 In addition to TAT, accuracy of WGS-based prediction is also a current barrier to  
136 implementation; to be financially feasible for most TB control programmes, WGS would  
137 optimally serve as a replacement test - eliminating the need for culture altogether. To do so,  
138 we need sufficient WGS quality to accurately identify ('call') single nucleotide  
139 polymorphisms (SNP; a single change in a base-pair, compared to a reference). The  
140 frequency and type of sequencing errors vary across platforms, due to differences in the  
141 underlying sequencing chemistry (e.g., Illumina MiSeq has an error rate of 0.8%, PacBio RS  
142 12.9%, MinION 5-30%; Dr. David Dolinger, Foundation for Innovative New Diagnostics,  
143 personal communication). Such errors must be accounted for in the analysis, with only high-  
144 quality base calls retained. In the case of the MinION, which Votintseva *et al.* applied - for  
145 the first time ever reported - to *Mycobacterium* (14), authors identified a systematic A→G  
146 error bias in 1D reads, which must be accounted for in the bioinformatics analysis.

147

148 In addition to such platform-specific considerations, overall quality control parameters such  
149 as depth of coverage and genome coverage also matter when assessing sequencing results.  
150 For such parameters, the minimum thresholds necessary for clinical use still need to be  
151 determined. With lower depth of coverage (and/or lower genome coverage), resistance-  
152 connoting mutations are more likely to be missed. Using lower thresholds for depth may also  
153 reduce our power to detect populations with mixed resistance profiles, or rule out false

154 positive SNPs due to sequencing or mapping errors. Previous studies performing WGS on  
155 DNA from cultured *M. tuberculosis* have required a minimum of 8 to 20x depth of coverage  
156 at the specific locus of interest to confidently call a SNP (with a minimum average of 40-50x  
157 across the genome currently recommended). Given the technical difficulty of sequencing  
158 from direct samples, these thresholds may not be feasible for this approach. In Votintseva *et*  
159 *al.* (14), authors required > 3x depth to allow resistance predictions to be made, a threshold  
160 only met by 24/37 (65%) of the direct samples of *M. tuberculosis* sequenced by Illumina  
161 MiSeq. Of the 96 possible predictions made for first-line drugs, 92 were concordant with  
162 phenotypic DST (96%); the 4 discordant predictions (for rifampin and pyrazinamide) were  
163 made on samples from a single patient, which had varying phenotypes for these drugs. When  
164 experimentally-spiked *M. bovis* DNA was sequenced using the MinION, the expected H57D  
165 *pncA* mutation was confidently called for all but the lowest concentration of BCG DNA  
166 (5%). However, while this resistance mutation was correctly classified using a 3x threshold,  
167 it is important to note that deeper sequencing was required to confidently exclude SNPs at  
168 174 other loci known to be associated with resistance. Given the consequences of both false  
169 positive and false negative resistance calls, before WGS can be implemented into the  
170 diagnostic workflow, quality control thresholds used must therefore undergo thorough  
171 validation, and ideally, be standardized for reproducible clinical use (16, 17).

172

173 In addition to being influenced by WGS data quality, our ability to accurately predict  
174 phenotypic resistance is also currently limited by our knowledge of resistance-connoting  
175 mutations. The potential for interaction between different mutations, both compensatory and  
176 resistance-connoting, as well as strain genetic background (i.e., lineage) (18), further  
177 complicates this understanding. While several databases have curated mutations from the  
178 literature (e.g., TBDR (19), TBDR (16)), it is difficult to evaluate causality of such



179 mutations, and assess for potential interactions absent complete WGS and phenotypic data at  
180 the isolate level. Studies using WGS to predict drug resistance ((20-22)) have shown that  
181 sensitivity and specificity of this approach varies substantially by drug, with accuracy lowest  
182 for second-line anti-TB drugs. As phenotypic DST for second-line drugs is typically only  
183 performed if first-line resistance is detected, the numbers of isolates with both genotypic and  
184 phenotypic data for these drugs is small. To improve our knowledge of genotypic-phenotypic  
185 concordance, large WGS databases with corresponding DST are therefore needed,  
186 incorporating both drug-sensitive and drug-resistant isolates (16). In response to this, several  
187 multi-national programs have been initiated (e.g., the CRyPTIC Project at  
188 <http://modmedmicro.nsms.ox.ac.uk/cryptic/> and ReSeqTB (23) at  
189 <https://platform.reseqtb.org>). These programs are facilitating the collection and curation of  
190 WGS data with corresponding phenotypic DST, as well as critical clinical outcome data,  
191 from both public and private sectors, which will enhance our understanding of causal  
192 resistance-connoting mutations. Until such time, if WGS is used to predict resistance,  
193 phenotypic DST should be performed in parallel - a prospect that is likely cost-prohibitive for  
194 most TB control programs.

195

196 In addition to issues with speed and accuracy, the substantial resources required for WGS are  
197 other obstacles to implementation. A large capital investment is necessary to establish the  
198 infrastructure for WGS, which may be particularly difficult for developing countries.  
199 Sequencing platforms themselves are expensive; as of January 2017, “desktop sequencers”  
200 Illumina MiniSeq and MiniSeq cost approximately \$50,000 USD and \$100,000 USD,  
201 respectively (excluding consumables). Platforms with even higher throughput such as the  
202 PacBio RS P6-C4 (Pacific Biosciences, Menlo Park, California, USA) or Illumina HiSeq X  
203 cost \$675,000 USD and \$1 million USD, respectively (Dr. David Dolinger, Foundation for

204 Innovative New Diagnostics, Geneva, personal communication). Aside from the platforms  
205 themselves, laboratories must also have allocated space, continuous, uninterrupted power  
206 supplies and highly trained personnel to calibrate, operate and support these platforms (24,  
207 25). Oxford Nanopore Technology's MinION, recently used to sequence Ebola strains in  
208 Guinea (24), has been proposed as a potential solution to some of these implementation  
209 issues. Cost-wise, the Oxford Nanopore MinION platform itself is only \$1,000 (price as of  
210 February 3<sup>rd</sup>, 2017, <https://store.nanoporetech.com/minion/sets/>, including 2 flowcells and a  
211 starter kit of reagents). However, a maximum of 12 samples can be sequenced at once using  
212 the MinION 9.4 flowcell, and, as they are not reusable, a new flowcell must be used for each  
213 additional use (at a cost of ~\$10 USD a piece, when purchased in bulk). In terms of  
214 laboratory requirements, the MinION is more easily transportable and requires less lab space  
215 (24, 25), but still requires human intervention for sample preparation. Thus, regardless of the  
216 next-generation sequencing platform used, most of the required resources are currently only  
217 be available in a reference laboratory setting, prohibiting the decentralization of WGS for TB  
218 diagnostics (26).

219  
220 In addition to these infrastructure requirements, the bioinformatics analysis of WGS data has  
221 been a major bottleneck in the widespread use of this method. Recently, substantial advances  
222 in this area have been made, with the development of rapid, user-friendly tools, such as  
223 PhyResSE (27), TBProfiler (22), TGS-TB (28) (which uses KvarQ (29)) and Mykrobe  
224 Predictor (20), to facilitate resistance prediction. These tools can take raw sequencing files  
225 from cultured *M. tuberculosis* DNA all the way through to resistance profiles (though one  
226 needs some understanding of bioinformatics concepts for interpretation). Additionally, both  
227 Mykrobe and KvarQ can be run offline, though, to our knowledge, Mykrobe is the only rapid  
228 resistance tool compatible with the readily-transportable MinION; this is ideal for use in

229 regions with limited internet access (though periodic updates to include novel resistance-  
230 connoting mutations from the literature would be needed). To utilize these tools with the  
231 direct sequencing protocol proposed in (14), human DNA must be removed *a priori*; as this  
232 involved mapping to a human reference genome, future software iterations will need to  
233 automate this step for easier use by frontline technicians without bioinformatics expertise.

234

235 In conclusion, WGS has the potential to become the future of TB DST (30) - provided these  
236 key considerations in terms of speed, accuracy and resource requirements are addressed.

237 Studies like (14) provide essential methodological advances and proof-of-concept needed to  
238 help move WGS to the clinical arena, and ultimately make real-time sequencing of *M.*

239 *tuberculosis* a reality. To fully evaluate the different DNA extraction protocols for direct

240 sequencing, in combination with different sequencing platforms, future studies are needed

241 employing these methods on the same samples, in a head-to-head comparison. Additional

242 studies are also needed to validate bioinformatics thresholds for clinical use, and assess the

243 performance of direct sequencing on different specimen types (e.g., smear-negative, culture-

244 positive samples, and non-respiratory samples). Given that countries with the highest burden

245 of TB are those with more limited resources, lower-cost, decentralized, simpler platforms are

246 essential for any meaningful scale-up.

247

248 **Acknowledgements.** Authors would like to thank Dr. Marcel A. Behr, McGill University, for

249 his critical review of this manuscript, and Dr. David Dolinger from the Foundation for New

250 Innovative Diagnostics (FIND) for providing current costing data for next-generation

251 sequencing platforms, and discussing issues in comparing these technologies. This research

252 received no specific grant from any funding agency in the public, commercial, or not-for-

253 profit sectors.

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