



Microevolution of *Mycobacterium tuberculosis* Subpopulations and Heteroresistance in a Patient Receiving 27 Years of Tuberculosis Treatment in Germany

 Lindsay Sonnenkalb,^a Gerald Strohe,^b Viola Dreyer,^a Sönke Andres,^c Doris Hillemann,^c  Florian P. Maurer,^{c,d} Stefan Niemann,^{a,e} Matthias Merker^a

^aMolecular and Experimental Mycobacteriology, Research Centre Borstel, Borstel, Germany

^bLandratsamt Karlsruhe, Gesundheitsamt, Karlsruhe, Germany

^cNational and Supranational Reference Centre for Mycobacteria, Research Centre Borstel, Borstel, Germany

^dInstitute of Medical Microbiology, Virology and Hygiene, University Medical Centre Hamburg-Eppendorf, Hamburg, Germany

^eGerman Centre for Infection Research (DZIF), Partner site Hamburg-Lübeck-Borstel, Borstel, Germany

Stefan Niemann and Matthias Merker contributed equally.

ABSTRACT Preexisting and newly emerging resistant pathogen subpopulations (heteroresistance) are potential risk factors for treatment failure of multi/extensively drug resistant (MDR/XDR) tuberculosis (TB). Inpatient evolutionary dynamics of *Mycobacterium tuberculosis* complex (Mtb) strains and their implications on treatment outcomes are still not completely understood. To elucidate how Mtb strains escape therapy, we analyzed 13 serial isolates from a German patient by whole-genome sequencing. Sequencing data were compared with phenotypic drug susceptibility profiles and the patient's collective 27-year treatment history to further elucidate factors fostering inpatient resistance evolution. The patient endured five distinct TB episodes, ending in resistance to 16 drugs and a nearly untreatable XDR-TB infection. The first isolate obtained, during the patient's 5th TB episode, presented fixed resistance mutations to 7 anti-TB drugs, including isoniazid, rifampin, streptomycin, pyrazinamide, prothionamide, para-aminosalicylic acid, and cycloserine-terizidone. Over the next 13 years, a dynamic evolution with coexisting, heterogeneous subpopulations was observed in 6 out of 13 sequential bacterial isolates. The emergence of drug-resistant subpopulations coincided with frequent changes in treatment regimens, which often included two or fewer active compounds. This evolutionary arms race between competing subpopulations ultimately resulted in the fixation of a single XDR variant. Our data demonstrate the complex inpatient microevolution of Mtb subpopulations during failing MDR/XDR-TB treatment. Designing effective treatment regimens based on rapid detection of (hetero) resistance is key to avoid resistance development and treatment failure.

KEYWORDS *Mycobacterium tuberculosis*, antibiotic resistance, antibiotics, evolution, microevolution, multidrug resistance, patient treatment, relapse, treatment failure, tuberculosis

With an estimated 10 million new cases in 2018 and half a million new multidrug-resistant (MDR) cases, tuberculosis (TB) caused by bacteria of the *Mycobacterium tuberculosis* complex (Mtb) continues to be the most devastating disease caused by a single infectious agent (1). Due to transmission and treatment failures, MDR-TB (defined as resistance to isoniazid [H] and rifampicin [R]) and extensively drug resistant (XDR) TB (includes additional resistance to a fluoroquinolone [FQ] and second-line injectable drug) cases continue to rise. Worldwide, TB treatment success rates are

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Address correspondence to Matthias Merker, mmerker@fz-borstel.de.

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about 82% for susceptible infections, decreasing to as low as 55% in MDR-TB cases (1). In approximately 90% of failed treatment cases, relapse occurs within 12 months of completed treatment (2, 3). Not all possible causes of treatment failure are well defined, but patient noncompliance and inappropriate drug regimens are most important (4, 5). Additionally, patient metabolism, pharmacokinetics, and pharmacodynamics also play a role in resistance acquisition and treatment failure (6). In efforts to overcome these treatment limitations, strategies such as drug administration programs, centralized treatment methods, comprehensive and rapid drug susceptibility testing, and precision medicine have been implemented (7, 8).

A number of studies, which have utilized sequencing techniques on serial isolates from the same patient, have found that resistance variants arise and are selected in failing treatment regimens from heterogeneous populations (9–14). These heterogeneous populations can comprise of several resistant subpopulations, also known as heteroresistance. Continued drug exposure on these populations ultimately selects and fixes a single resistance-mediating mutation. *In vitro* studies have further demonstrated that individual mutations can lead to a significant (and variable) reduction or increase in bacterial fitness, which likely explains the selection and loss of certain mutations during therapy (15, 16).

Understanding the mode and conditions under which drug resistance-associated mutations arise and are selected is paramount when considering diagnostic procedures and treatment regimens. For this understanding, we need rapid and sensitive diagnostics like next-generation sequencing (NGS) amplicon sequencing, which allows for the detection of genotypic drug resistance and heteroresistance populations in patient sputum samples at low frequencies (17). Such tools could guide better treatment regimens, as they have been shown to detect resistance and diverse populations as well as, or better than, phenotypic drug susceptibility testing (pDST) (18, 19).

In this study, we investigated intrapatient *Mtbc* microevolution within a single patient who suffered 5 distinct TB episodes resulting in a TB treatment of 27 years. We aligned whole-genome sequencing (WGS) data of 13 serial isolates collected during the final infectious episode, with pDST data and the patient's treatment history, to explore the connection between treatment regimens and evolutionary dynamics.

RESULTS

Case history. The patient, of German descent, was first diagnosed with pulmonary TB in Western Germany in the late 1950s at 4 years old. After 4 months of treatment with H, para-aminosalicylic acid (PAS), and dihydrothentat injectable (a derivative of streptomycin), therapy was concluded and the patient was considered cured. Over 39 years, the patient endured an additional 4 relapse events, with treatment lasting 4 months, 41 months, 88 months, and finally 216 months until the patient died of the infection (see Table S1 in the supplemental material). It is not clear whether each episode was attributed to reinfection, reactivation, or both, as only bacterial isolates from the last TB episode were recovered.

The patient disclosed treatment noncompliance to clinicians, stated as “not properly administering his medication” during some previous treatment periods. Patient records did not indicate that an HIV coinfection was present; however, immunosuppressive activity, such as alcohol abuse and smoking cigarettes, was noted (20, 21).

Initial pDST was conducted during the patient's 4th TB episode that revealed the *Mtbc* strain was already resistant to five drugs. Several months later, additional testing confirmed MDR-TB. Over the final 18-year treatment course, the antibiotic regimen included a maximum of 3 active drugs but most often 2 or 1 (Table S1; Fig. 1). During this final TB episode, sequential bacterial isolates were recovered spanning 13 years of the treatment period. WGS revealed initial resistance to seven drugs (nine when considering low frequency populations) and indicated the acquisition of resistance to five additional drugs over this time. Although pDST designated similar resistance profiles

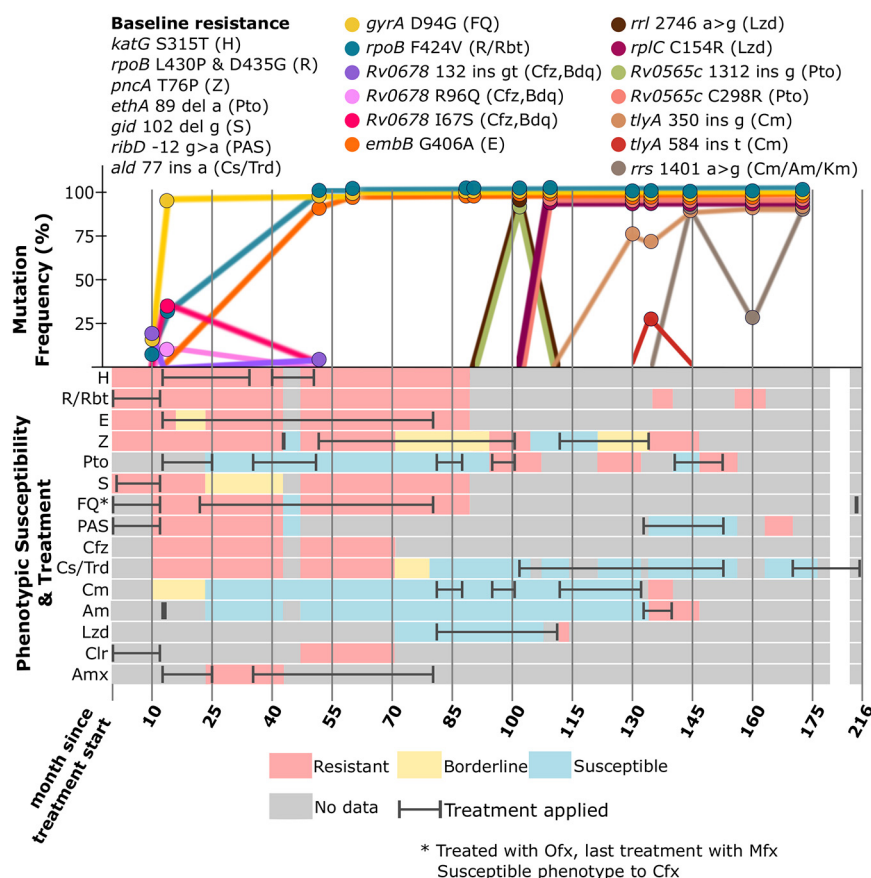


FIG 1 Patient treatment history, bacterial phenotypic drug susceptibility test results, and acquisition of resistance-mediating mutations. Newly emerging mutations implicated in resistance are color coded in the top panel. Mutation frequency (y axis) is inferred from next-generation sequencing (NGS) data, i.e., frequency of the resistance allele, and time points available for NGS analysis are indicated by circles. Lines represent changes of mutation frequencies over time. Phenotypic drug susceptibility test (pDST) results are color coded in the bottom panel. Horizontal bars indicate application of a drug (overlying the pDST). Both pDST and drug regimens are based on the patient record. Recent guidelines for drug susceptibility testing of *M. tuberculosis* complex isolates do not support the critical test concentration for the antibiotics Clr, Amx, Cs/Trd, and PAS. Am, amikacin; Amx, amoxicillin+clavulanic acid; Bdq, bedaquiline; Cfx, ciprofloxacin; Cfz, clofazimine; Clr, clarithromycin; Cm, capreomycin; Cs, cycloserine; E, ethambutol; FQ, fluoroquinolone; H, isoniazid; Km, kanamycin; Lzd, linezolid; Mfx, moxifloxacin; OP, outpatient; PAS, para-aminosalicylic acid; Pto, prothionamide; R, rifampicin; Rbt, rifabutin; S, streptomycin; Trd, terizidone; Z, pyrazinamide.

as the genotype predicted, inconsistent results for some antibiotics were recurrent, alternating between resistant and susceptible.

Preexisting and treatment-selected resistance-mediating mutations. Overall, WGS analysis was performed on 13 serial isolates to predict drug resistance and drug susceptibility (termed genotypic drug susceptibility testing [gDST]) (Fig. 1; see Table S2 in the supplemental material). All bacterial isolates obtained belonged to Mtbc lineage 4.7 and had a maximum distance between sequential isolates of less than five alleles, while showing a strictly clonal evolution, thus excluding a reinfection (Fig. 2).

The first isolate available for WGS was collected 10 months after beginning treatment, which revealed resistance-mediating mutations to 7 different antibiotics, i.e., *katG* S315T for H, *rpoB* L430P in combination with D435G for R, *pncA* T76P for pyrazinamide (Z), *ribD* at -12 g > a for para-aminosalicylic acid (PAS), *ethA* 89 del a for prothionamide (Pto), *gidB* 102 del g for streptomycin (S), and *ald* 77 ins a for cycloserine-terizidone (Cs/Trd) (where del is deletion and ins is insertion). Antibiotics which had been included in previous regimens were H, R, ethambutol (E), Z, S, PAS, capreomycin

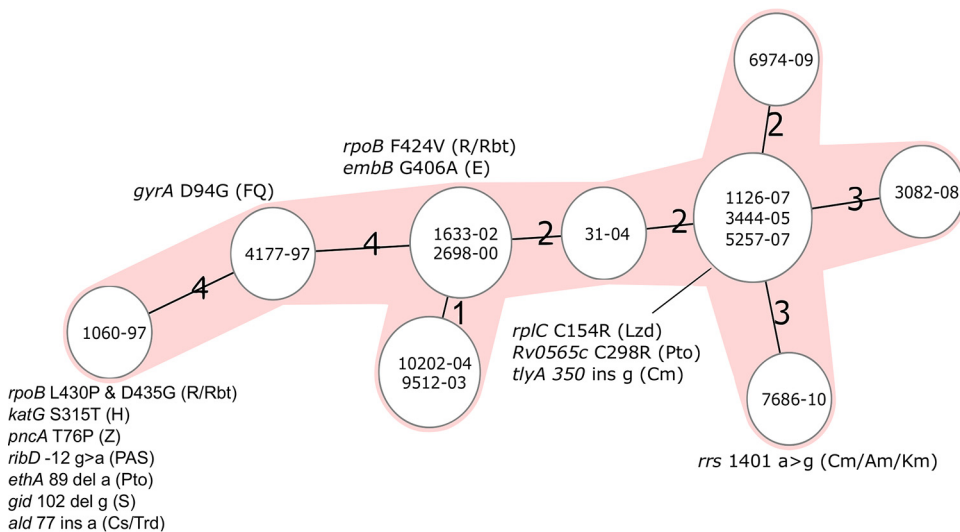


FIG 2 Inpatient microevolution of a multidrug resistant (MDR) *Mycobacterium tuberculosis* complex (Mtb) strain. Minimum spanning tree based on a core genome multilocus sequence type (cgMLST) analysis of 13 serial patient isolates. The number of allele differences are indicated on connecting lines. Resistance- mediating mutation (and associated drug) are noted next to the isolate in which the mutation became fixed in the genome. Am, amikacin; Cm, capreomycin; Cs, cycloserine; E, ethambutol; FQ, fluoroquinolone; H, isoniazid; Km, kanamycin; Lzd, linezolid; PAS, para-aminosalicylic acid; Pto, prothionamide; R, rifampicin; Rbt, rifabutin; S, streptomycin; Z, pyrazinamide.

(Cm), clarithromycin (Clr), Pto, and Trd for a minimum of 3 consecutive months, although most were applied for over 10 months (Table S1).

The isolate analyzed from month 13 indicated fluoroquinolone (FQ) resistance, mediated by the mutation *gyrA* D94G, after about 1 year of ofloxacin (Ofx) treatment. No further isolate was collected until month 47, which then revealed resistance to E mediated by the mutation *embB* G406A. However, phenotypic resistance to E was detected already in month 22, while E was included in the treatment since month 13. Additionally, a third R-associated mutation (*rpoB* F424V) emerged, while R was still included in the treatment regimen (Fig. 1).

In month 118, after 26 months of linezolid (Lzd) treatment, the resistance-mediating mutation *rplC* C154R emerged to fixation. From the isolate obtained in month 130, we identified the mutation *tlyA* 350 ins g likely mediating resistance against the second-line injectable drug Cm, observed after 18 consecutive months of Cm treatment. Additionally, in month 144, after 8 months of amikacin (Am) treatment, the mutation *rrs* 1401 a > g arose conferring cross-resistance against all second-line injectable drugs (Cm, Am, and kanamycin [Km]).

Insufficient treatment regimens. The treatment of this patient was highly complex, with regimens frequently changing. However, the treatment regimens were not always congruent with pDST results (Fig. 1). This resulted in suboptimal regimens consisting of fewer than 3 active drugs for virtually the last 18 years of treatment. Moreover, only two, one, or even no active drugs were administered at several time points (Table S1; Fig. 1).

In the final TB treatment episode, when considering only gDST, we observed only one instance in which a maximum of three active drugs (E, Am, and amoxicillin + clavulanic acid [Amx]) were included in treatment, in months 13 and 14. On the other hand, Amx may not be considered, as it is not typically prescribed for TB, but it has presented synergistic effects during MDR-TB treatment (22, 23). From month 1 through 12, Clr appeared to be the only active drug applied but has since shown to be trivial for TB treatment due to intrinsic resistance by Mtb strains (24, 25). However, a study in 1996 indicated that Clr might have a synergistic effect when included in combination treatments with R, H, and/or E *in vitro* (26).

In most instances when new resistance arose, only 1 or 2 active drugs were prescribed; this was notably seen with Lzd (Table S1). From month 88 through 94, Lzd was the only active drug included in treatment, and the patient gained Lzd resistance by the

fixation of mutation *rrl* 2746 a > g in month 104 of treatment, followed by *rplC* C154R in month 108, when the isolate was still indicated as phenotypically susceptible (Fig. 1).

Heteroresistance and microevolution. To resolve discrepancies between pDST and gDST and to investigate the evolutionary trajectories of emerging resistant subpopulations, i.e., heteroresistance, we investigated the presence of mutations down to a frequency of 1% and included only statistically verified variants in our report (Table S2; see Table S3 in the supplemental material). Overall, we found 10 instances of low frequency resistance-mediating mutations (below 75% frequency), which either emerged months before reaching fixation, fluctuated at different frequencies over the course of the therapy, and/or disappeared from the bacterial population (Fig. 1; Table S2).

For instance, FQ resistance mediated by *gyrA* D94G was already detected in month 10 at a frequency of 21%. Also, a third *rpoB* F424V mutation (with unclear phenotypic effect) could be observed in month 10 at 5% and month 13 at 30%, and it reached fixation in the subsequent isolate collected in month 47 at 96% frequency.

Heteroresistance could also be observed with regard to Cm (Fig. 1; Table S2). In month 130, one *tlyA* insertion was detected, namely, 350 ins g at 79% frequency, 18 months after Cm was introduced into the treatment regimen. In the subsequent isolate collected 5 months later, the additional frameshift mutation *tlyA* 584 ins t was detected at 28% and the 350 ins g mutation decreased slightly to 71%. Finally, 9 months later, *tlyA* 350 ins g reached 97% frequency, while the second subpopulation *tlyA* 584 ins t was not again detected.

Cross-resistance to all second-line injectable drugs mediated by *rrs* 1401 a > g was then found at 100% frequency in isolate 3082-08 collected in month 144, decreased to 29% frequency in the subsequent isolate, and then increased again to 99% thereafter.

In the first collected isolates, we discovered a low frequency mutation (*Rv0678* 132 ins gt at 22%) potentially conferring cross-resistance to clofazimine (Cfz) and/or bedaquiline (Bdq). This frameshift mutation was lost in the following isolate, correlating with the emerging mutations *Rv0678* I67S at 36% and *Rv0678* R96Q at 10% frequencies. Both mutations correlated with Cfz resistance pDST, although there is no mention of Cfz inclusion in the drug regimens throughout all the patient's treatments. Again, *Rv0678* 132 ins gt was detected in month 47 at 3.6%. All *Rv0678* mutations eventually disappeared from the population by month 71 of treatment.

Putative compensatory and tolerance effects. Finally, two high frequency mutations without a direct effect on resistance were detected. A mutation in the gene *prpR* (*Rv1129c*) F334L involved in drug tolerance (27) was first found in month 92, was not observed in month 104 isolate, but was then fixed in the population after month 108 (Table S2).

Two mutations arose in monooxygenase *Rv0565c*, a gene with a possible compensatory mechanism which overcomes fitness defects brought on by mutations in the monooxygenase *ethA* gene (activating the drugs ethionamide [Eto] and Pto) (28). These mutations in *Rv0565c* developed in isolates after the mutation *ethA* 89 del a. First the mutation *Rv0565c* 1312 ins g arose to 97% frequency in month 104 but was lost in all following isolates. In the subsequent isolate from month 108, a second mutation *Rv0565c* C298R was detected at 99% and remained fixed in the population.

DISCUSSION

Our analysis of Mtbc microevolution over the complex treatment of one patient revealed that fixation of resistance mutations in the bacterial population is a dynamic process. The emergence and extinction of different subpopulations are likely triggered by therapy changes and suboptimal treatment regimens. The rapid detection of heteroresistance by NGS techniques could offer new opportunities for intervention measures and a more effective treatment regimen.

Through the application of WGS, we could show that resistance evolution was influenced by long periods of ineffective treatment regimens, with several periods of only one or two active drugs applied. Suboptimal therapy design was partially due to minimal options of active drugs, especially in the last 18 years of treatment. In fact, there

were several time points in which drugs were still applied, despite resistant pDST. For example, between month 48 and 81, the patient was treated with E, Ofx, Z, and Amx, despite already presenting previous phenotypic resistant to E, Ofx, and Z. Genotypic tests were not performed at that time, but WGS analysis retrospectively confirmed resistance to E (*embB* G406A), Ofx (*gyrA* D94G), and Z (*pncA* T76P). Finally, frequent regimen changes and poor treatment design may have been fostered by changes in clinic (treated in 12 clinics within Germany throughout the patient's life), several laboratories reporting pDST results, and also lack of remaining effective drugs.

During our longitudinal analysis, we observed at several time points heterogeneous subpopulations, indicating emerging resistance during the treatment course. As multiple resistant subpopulations can arise and coexist during MDR-TB treatment, diagnostic approaches need to be employed, such as amplicon sequencing of sputum, which enables the rapid and sensitive detection of resistance and low frequency resistance subpopulations (29). Such information could then be used to rapidly change treatment regimens and help avoid treatment failure. In order to enact this approach in the future, one also needs to consider bacterial subpopulations can reside at different sites of infection, each following their own microevolution (30). In order to rapidly respond to changes of mutation frequencies, sampling should be performed regularly to capture the entire inpatient strain diversity.

The limitation of this study is the retrospective character and that most of the patient isolates were not available to repeat pDST according to current standards. Of note, any cultivation step prior to pDST and DNA isolation for NGS can potentially influence the mutation frequencies reported in our study. Furthermore, individual sputum specimens may only comprise a fraction of the overall Mtb diversity within the patient. Additionally, distant anatomical lesions can contain different bacterial populations with distinct resistance mutation profiles (31). As mentioned, patient records were compiled from 12 different clinics and external pDST results were sometimes contradicting the data acquired at National Reference Centre Borstel, but as presented, they should represent the information given to the attending clinician at that time. Rationales for the design of individual therapy episodes could not be retrieved.

Additionally, inpatient competition of subpopulations may also induce the emergence of putative compensatory and tolerance mutations described as *Rv0565c* C298R, *Rv0565c* 1312 ins g (Pto), and *pprR* F334L (27, 28). Drug tolerance-associated mutations and compensatory mutations are discussed for their potential clinical relevance, as they may affect bacterial growth and mutation rates. These types of mutations and mechanisms are generally not detected in typical pDST. In the case for the emergence of *Rv0565c* mutations, in an *ethA*-deficient genetic background, and coinciding with a Pto-resistant phenotype, we cannot exclude that *Rv0565c* is also implicated in resistance against Pto itself.

We demonstrated that resistance development in a failing MDR-TB therapy involved an arms race of coexisting bacterial subpopulations. Continued treatment with less than four active drugs likely selected for the most resistant clone over time. This study also highlights the benefits of genomic resistance testing, which can improve treatment with drugs lacking recommendations of critical test concentrations (PAS and Cs) or drugs with poorly reproducibly pDST, such as Z and E (32, 33).

MATERIALS AND METHODS

Phenotyping and DNA isolation of serial patient isolates. Sputum samples were collected from a single patient in Germany and were received by the National Reference Centre of Research Centre Borstel Leibniz Lung Centre (NRC-Borstel). The bacteria recovered from the sputum samples were stored at NRC-Borstel and were regrown on Löwenstein-Jensen slants for DNA isolation for this study. The bacteria were cultured at 37°C until colonies were visible (about 3 weeks); and colonies were then scraped from the medium using a sterile loop, transferred to 400- μ l Tris-EDTA (TE) buffer, and heat killed in an 80°C water bath for 20 minutes. Isolation of genomic DNA was conducted using the standard cetyltrimethylammonium bromide (CTAB) method as described previously (34). Drug susceptibility testing conducted at NRC-Borstel used *Mycobacterium* growth indicator tubes (MGIT; Becton, Dickinson Microbiology Systems, Sparks, MD) or agar dilution on Middlebrook 7H10 (or 7H11) plates with the recommended critical concentrations at the time of pDST; pDST methods conducted by other labs could

not be retrieved. Patient treatment history was collected and provided by the Rastatt Department of Health and the state health office of Stuttgart in Germany. The pDST results presented were also compiled by the Rastatt Department of Health. Although there are not clearly defined critical concentrations for Clr and Amx, these drugs were included in pDST in the patient treatment history; therefore, we presented the resistance profile for these drugs as it was documented.

Next-generation sequencing and resistance prediction of serial patient isolates. Genomic DNA was sequenced using Illumina NextSeq 500 technology and Nextera XT library preparation kits according to manufacturer's guidelines. Mutations (single-nucleotide polymorphisms, small insertions, and deletions) were detected with the MTBSeq pipeline adjusting the minimum coverage (of two reads) to distinguish mutations at low frequencies (35). Variant calls were first filtered in the MTBSeq low frequency output with a threshold of at least 1 read in both forward and reverse orientation and a minimum of 1% frequency. All variant calls with a frequency below 75% were statistically verified using the binoSNP variant detection tool and calculated by number of calls with a minimum phred base quality of 20 (36). All low frequency variants with a *P* value of ≤ 0.05 were included (Table S3). Genes in highly repetitive regions, e.g., proline-glutamate (PE), proline-proline-glutamate (PPE) genes, and genes with polymorphic-GC-rich sequences (PGRS), were not considered. For the genotypic prediction of drug resistance and drug susceptibility, mutations in 92 genes implicated in resistance to 21 different anti-TB drugs were screened; phylogenetically informative mutations were not considered (37). In the absence of a known resistance mutation, the isolate was considered drug susceptible.

Of note, false-negative results could not be excluded, as many of the historic specimens were no longer available for repeated pDST. A minimum spanning tree was calculated with a core genome multilocus sequence type (cgMLST) approach with SeqSphere v5.9 (Ridom, Münster, Germany) as described previously and by pairwise ignoring missing values (38).

Data availability. The bacterial DNA sequencing data supporting the conclusion of the manuscript are available at the European Nucleotide Archive (<https://www.ebi.ac.uk/ena/browser/home>). Accession numbers are ERS4932039 (patient isolate identification no. 1060-97), ERS4932040 (4177-97), ERS4932041 (2698-00), ERS4932042 (1633-02), ERS4932043 (9512-03), ERS4932044 (31-04), ERS4932045 (10202-04), ERS4932046 (3444-05), ERS4932047 (1126-07), ERS4932048 (5257-07), ERS4932049 (3082-08), ERS4932050 (6974-09), and ERS4932051 (7686-10).

SUPPLEMENTAL MATERIAL

Supplemental material is available online only.

SUPPLEMENTAL FILE 1, PDF file, 0.2 MB.

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REFERENCES

1. WHO. 2020. Global tuberculosis report 2019. WHO, Geneva, Switzerland.
2. Luzzo H, Johnson DF, Dickman K, Mayanja-Kizza H, Okwera A, Eisenach K, Cave MD, Whalen CC, Johnson JL, Boom WH, Joloba M. 2013. Relapse more common than reinfection in recurrent tuberculosis 1–2 years post treatment in urban Uganda. *Int J Tuberc Lung Dis* 17:361–367. <https://doi.org/10.5588/ijtld.11.0692>.
3. Johnson JL, Thiel BA. 2012. Time until relapse in tuberculosis treatment trials: implication for phase 3 trial design. *Am J Respir Crit Care Med* 186:464. <https://doi.org/10.1164/ajrccm.186.5.464>.
4. Dheda K, Gumbo T, Maartens G, Dooley KE, McNerney R, Murray M, Furin J, Nardell EA, London L, Lessem E, Theron G, van Helden P, Niemann S, Merker M, Dowdy D, Van Rie A, Siu GKH, Pasipanodya JG, Rodrigues C, Clark TG, Sirgel FA, Esmail A, Lin HH, Atre SR, Schaaf HS, Chang KC, Lange C, Nahid P, Udawadia ZF, Horsburgh CR, Churchyard GJ, Menzies D, Hesselring AC, Nuermberger E, McIlleron H, Fennelly KP, Goemaere E, Jaramillo E, Low M, Jara CM, Padayatchi N, Warren RM. 2017. The epidemiology, pathogenesis, transmission, diagnosis, and management of multidrug-resistant, extensively drug-resistant, and incurable tuberculosis. *Lancet Respir Med* 15:S2213–2600(17)30079–6. [https://doi.org/10.1016/S2213-2600\(17\)30079-6](https://doi.org/10.1016/S2213-2600(17)30079-6).
5. Burman WJ, Cohn DL, Rietmeijer CA, Judson FN, Sbarbaro JA, Reves RR. 1997. Noncompliance with directly observed therapy for tuberculosis: epidemiology and effect on the outcome of treatment. *Chest* 111:1168–1173. <https://doi.org/10.1378/chest.111.5.1168>.
6. Srivastava S, Pasipanodya JG, Meek C, Leff R, Gumbo T. 2011. Multidrug-resistant tuberculosis not due to noncompliance but to between-patient pharmacokinetic variability. *J Infect Dis* 204:1951–1959. <https://doi.org/10.1093/infdis/jir658>.
7. Miotto P, Zhang Y, Cirillo DM, Yam WC. 2018. Drug resistance mechanisms and drug susceptibility testing for tuberculosis. *Respirology* 23:1098–1113. <https://doi.org/10.1111/resp.13393>.
8. Gröschel MI, Walker TM, van der Werf TS, Lange C, Niemann S, Merker M. 2018. Pathogen-based precision medicine for drug-resistant tuberculosis. *PLoS Pathog* 14:e1007297. <https://doi.org/10.1371/journal.ppat.1007297>.
9. Sun G, Luo T, Yang C, Dong X, Li J, Zhu Y, Zheng H, Tian W, Wang S, Barry CE, Mei J, Gao Q. 2012. Dynamic population changes in mycobacterium tuberculosis during acquisition and fixation of drug resistance in patients. *J Infect Dis* 206:1724–1733. <https://doi.org/10.1093/infdis/jis601>.
10. Merker M, Kohl TA, Roetzer A, Truebe L, Richter E, Rüsche-Gerdes S, Fattorini L, Oggioni MR, Cox H, Varaine F, Niemann S. 2013. Whole genome sequencing reveals complex evolution patterns of multidrug-resistant *Mycobacterium tuberculosis* Beijing strains in patients. *PLoS One* 8:e82551. <https://doi.org/10.1371/journal.pone.0082551>.
11. Eldholm V, Norheim G, von der Lippe B, Kinander W, Dahle UR, Caugant DA, Mannsåker T, Orun Mengshoel AT, Dyrhol-Riise AM, Balloux F. 2014. Evolution of extensively drug-resistant *Mycobacterium tuberculosis* from a susceptible ancestor in a single patient. *Genome Biol* 15:490. <https://doi.org/10.1186/s13059-014-0490-3>.
12. Liu Q, Via LE, Luo T, Liang L, Liu X, Wu S, Shen Q, Wei W, Ruan X, Yuan X, Zhang G, Barry CE, Gao Q. 2015. Within patient microevolution of *Mycobacterium tuberculosis* correlates with heterogeneous responses to treatment. *Sci Rep* 5:17507. <https://doi.org/10.1038/srep17507>.

13. Bloemberg GV, Keller PM, Stucki D, Trauner A, Borrell S, Latshang T, Coscolla M, Rothe T, Hömke R, Ritter C, Feldmann J, Schulthess B, Gagneux S, Böttger EC. 2015. Acquired resistance to bedaquiline and delamanid in therapy for tuberculosis. *N Engl J Med* 373:1986–1988. <https://doi.org/10.1056/NEJMc1505196>.
14. Müller B, Borrell S, Rose G, Gagneux S. 2013. The heterogeneous evolution of multidrug-resistant *Mycobacterium tuberculosis*. *Trends Genet* 29:160–169. <https://doi.org/10.1016/j.tig.2012.11.005>.
15. Gagneux S, Long CD, Small PM, Van T, Schoolnik GK, Bohannon BJM. 2006. The competitive cost of antibiotic resistance in *Mycobacterium tuberculosis*. *Science* 312:1944–1946. <https://doi.org/10.1126/science.1124410>.
16. Castro RAD, Ross A, Kamwela L, Reinhard M, Loiseau C, Feldmann J, Borrell S, Trauner A, Gagneux S. 2020. The genetic background modulates the evolution of fluoroquinolone-resistance in *Mycobacterium tuberculosis*. *Mol Biol Evol* 37:195–207. <https://doi.org/10.1093/molbev/msz214>.
17. Colman RE, Schupp JM, Hicks ND, Smith DE, Buchhagen JL, Valafar F, Crudu V, Romancenco E, Noroc E, Jackson L, Catanzaro DG, Rodwell TC, Catanzaro A, Keim P, Engelthaler DM. 2015. Detection of low-level mixed-population drug resistance in *Mycobacterium tuberculosis* using high fidelity amplicon sequencing. *PLoS One* 10:e0126626. <https://doi.org/10.1371/journal.pone.0126626>.
18. Doyle RM, Burgess C, Williams R, Gorton R, Booth H, Brown J, Bryant JM, Chan J, Creer D, Holdstock J, Kunst H, Lozewicz S, Platt G, Romero EY, Speight G, Tiberi S, Abubakar I, Lipman M, McHugh TD, Breuer J. 2018. Direct whole-genome sequencing of sputum accurately identifies drug-resistant *mycobacterium tuberculosis* faster than MGIT culture sequencing. *J Clin Microbiol* 56:e00666-18. <https://doi.org/10.1128/JCM.00666-18>.
19. Nimmo C, Shaw LP, Doyle R, Williams R, Brien K, Burgess C, Breuer J, Balloux F, Pym AS. 2019. Whole genome sequencing *Mycobacterium tuberculosis* directly from sputum identifies more genetic diversity than sequencing from culture. *BMC Genomics* 20:433. <https://doi.org/10.1186/s12864-019-5841-8>.
20. Soporì ML, Kozak W. 1998. Immunomodulatory effects of cigarette smoke. *J Neuroimmunol* 83:148–156. [https://doi.org/10.1016/s0165-5728\(97\)00231-2](https://doi.org/10.1016/s0165-5728(97)00231-2).
21. Piao WH, Campagnolo D, Dayao C, Lukas RJ, Wu J, Shi FD. 2009. Nicotine and inflammatory neurological disorders. *Acta Pharmacol Sin* 30:715–722. <https://doi.org/10.1038/aps.2009.67>.
22. Pagliotto ADF, Caleffi-Ferracioli KR, Lopes MA, Baldin VP, Leite CQF, Pavan FR, Scodro Rb de L, Siqueira VLD, Cardoso RF. 2016. Anti-*Mycobacterium tuberculosis* activity of antituberculosis drugs and amoxicillin/clavulanate combination. *J Microbiol Immunol Infect* 49:980–983. <https://doi.org/10.1016/j.jmii.2015.08.025>.
23. Diacon AH, Van Der Merwe L, Barnard M, Groote-Bidlingmaier FV, Lange C, Garcia-Basteiro AL, Sevens E, Ballell L, Barros-Aguirre D. 2016. Beta-lactams against TB: teaching a new trick to an old dog. *Top Antivir Med* 24:63–64.
24. Buriánková K, Doucet-Populaire F, Dorson O, Gondran A, Ghnassia JC, Weiser J, Pernodet JL. 2004. Molecular basis of intrinsic macrolide resistance in the *Mycobacterium tuberculosis* complex. *Antimicrob Agents Chemother* 48:143–150. <https://doi.org/10.1128/aac.48.1.143-150.2004>.
25. Dooley KE, Obuku EA, Durakovic N, Belitsky V, Mitnick C, Nueremberger EL. 2013. World Health Organization group 5 drugs for the treatment of drug-resistant tuberculosis: unclear efficacy or untapped potential? *J Infect Dis* 207:1352–1358. <https://doi.org/10.1093/infdis/jis460>.
26. Cavalieri SJ, Biehle JR, Sanders WE. 1995. Synergistic activities of clarithromycin and antituberculous drugs against multidrug-resistant *Mycobacterium tuberculosis*. *Antimicrob Agents Chemother* 39:1542–1545. <https://doi.org/10.1128/aac.39.7.1542>.
27. Hicks ND, Yang J, Zhang X, Zhao B, Grad YH, Liu L, Ou X, Chang Z, Xia H, Zhou Y, Wang S, Dong J, Sun L, Zhu Y, Zhao Y, Jin Q, Fortune SM. 2018. Clinically prevalent mutations in *Mycobacterium tuberculosis* alter propionate metabolism and mediate multidrug tolerance. *Nat Microbiol* 3:1032–1042. <https://doi.org/10.1038/s41564-018-0218-3>.
28. Hicks ND, Carey AF, Yang J, Zhao Y, Fortune SM. 2019. Bacterial genome-wide association identifies novel factors that contribute to ethionamide and prothionamide susceptibility in *mycobacterium tuberculosis*. *mBio* 10:e00616-19. <https://doi.org/10.1128/mBio.00616-19>.
29. Ng KCS, Supply P, Cobelens FGJ, Gaudin C, Gonzalez-Martin J, de Jong BC, Rigouts L. 2019. How well do routine molecular diagnostics detect rifampin heteroresistance in *Mycobacterium tuberculosis*? *J Clin Microbiol* 57:e00717-19. <https://doi.org/10.1128/JCM.00717-19>.
30. Ford CB, Lin PL, Chase MR, Shah RR, Iartchouk O, Galagan J, Mohaideen N, loerger TR, Sacchettini JC, Lipsitch M, Flynn JL, Fortune SM. 2011. Use of whole genome sequencing to estimate the mutation rate of *Mycobacterium tuberculosis* during latent infection. *Nat Genet* 43:482–488. <https://doi.org/10.1038/ng.811>.
31. Lieberman TD, Wilson D, Misra R, Xiong LL, Moodley P, Cohen T, Kishony R. 2016. Genomic diversity in autopsy samples reveals within-host dissemination of HIV-associated *Mycobacterium tuberculosis*. *Nat Med* 22:1470–1474. <https://doi.org/10.1038/nm.4205>.
32. Andres S, Gröschel MI, Hillemann D, Merker M, Niemann S, Kranzer K. a. 2018. A diagnostic algorithm to investigate pyrazinamide and ethambutol resistance in rifampin-resistant *Mycobacterium tuberculosis* isolates in a low-incidence setting. *Antimicrob Agents Chemother* 63:e01798-18. <https://doi.org/10.1128/AAC.01798-18>.
33. WHO. 2015. Implementing tuberculosis diagnostics: a policy framework. WHO, Geneva, Switzerland.
34. Van Sooling D, Hermans PWM, De Haas PEW, Soll DR, Van Embden JDA. 1991. Occurrence and stability of insertion sequences in *Mycobacterium tuberculosis* complex strains: evaluation of an insertion sequence-dependent DNA polymorphism as a tool in the epidemiology of tuberculosis. *J Clin Microbiol* 29:2578–2586. <https://doi.org/10.1128/JCM.29.11.2578-2586.1991>.
35. Kohl TA, Utpatel C, Schleusener V, De Filippo MR, Beckert P, Cirillo DM, Niemann S. 2018. MTBseq: a comprehensive pipeline for whole genome sequence analysis of *Mycobacterium tuberculosis* complex isolates. *PeerJ* 11:e5895. <https://doi.org/10.7717/peerj.5895>.
36. Dreyer V, Utpatel C, Kohl TA, Barilar I, Gröschel MI, Feuerriegel S, Niemann S. 2020. Detection of low-frequency resistance-mediating SNPs in next-generation sequencing data of *Mycobacterium tuberculosis* complex strains with binoSNP. *Sci Rep* 10:7874. <https://doi.org/10.1038/s41598-020-64708-8>.
37. Merker M, Kohl TA, Barilar I, Andres S, Fowler PW, Chrissyanthou E, Ångeby K, Jureen P, Moradigaravand D, Parkhill J, Peacock SJ, Schön T, Maurer FP, Walker T, Köser C, Niemann S. 2020. Phylogenetically informative mutations in genes implicated in antibiotic resistance in *Mycobacterium tuberculosis* complex. *Genome Med* 12:27. <https://doi.org/10.1186/s13073-020-00726-5>.
38. Kohl TA, Diel R, Harmsen D, Rothgänger J, Meywald Walter K, Merker M, Weniger T, Niemann S. 2014. Whole-genome-based *Mycobacterium tuberculosis* surveillance: a standardized, portable, and expandable approach. *J Clin Microbiol* 52:2479–2486. <https://doi.org/10.1128/JCM.00567-14>.