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## 2 3 Supplementary Information for 4 5 Morphological profiling of tubercle bacilli identifies drug pathways of action 6 **Authors:** Trever C. Smith II<sup>1,2†</sup>, Krista M. Pullen<sup>1,3†</sup>, Michaela C. Olson<sup>1</sup>, Morgan E. McNellis<sup>1</sup>, Ian 8 Richardson<sup>1,4</sup>, Sophia Hu<sup>5</sup>, Jonah Larkins-Ford<sup>1,6,7</sup>, Xin Wang<sup>8</sup>, Joel S. Freundlich<sup>8,9</sup>, D. Michael 9 Ando<sup>10</sup>, and Bree B. Aldridge<sup>1,2,6,7,11\*</sup> 10 11 Correspondence to: 12 Bree B. Aldridge 13 Email: bree.aldridge@tufts.edu 14 15

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 **Figure S1. Innate cell-to-cell heterogeneity in Mtb morphologies.** Dot and box plots of Mtb morphological features in untreated (yellow) and drug-treated bacilli. Many bacilli are not stained with SYTO 24, therefore only non-zero values are shown for nucleoid features. Cellular heterogeneity is apparent among the most basic cytological features, including cell length, intensity of nucleoid staining, and nucleoid width, even in untreated Mtb (CV=0.31, 0.64, 0.32,

- respectively). All plots were generated using the statistical data visualization library Seaborn for
- 34  $Python^{TM}$ .



 **Figure S2. Inherent heterogeneity of Mtb and subtle morphological features requires multistep, multivariable analysis to define cytological profiles.** (A) PCA (top) and UMAP (bottom) analyses of eight drug treatments at high dose after incorporating metrics of heterogeneity [quartile 1, quartile 3, and interquartile range (IQR)], feature selection, and batch normalization (TVN; bottom). n=3032-9651 for each treatment group over biological triplicate. TVN aligns the covariance matrices resulting from PCA and whitening of untreated control data from each batch and applies this transformation batch-by-batch to enable comparison across plates and replicates. (B) cKNN connectivity map for high dose drug treatment of Mtb. Drug nodes are colored by broad target category and edge thickness corresponds to how frequently profiles are nearest neighbors in the classification trials (connections are shown for links >12%).



 **Figure S3. UMAP analysis of 34 drug treated Mtb profiles.** Mtb drug profiles from 34 different antibacterials at 3X IC90 (Table S1) after expansion of quantified features (Q1, IQR, Q3) and post TVN. Data points represent individual biological replicates of a drug treatment and are color-coded based on the broad cellular target as determined by literature review (Table S1).



 **Figure S4. Correction of batch-to-batch heterogeneity.** (PCA) of untreated data labeled by batch. (A) PCA prior to TVN normalization. (B) PCA after TVN normalization. PCA of all 34 drug treatments labeled by target pathway using all 94 features. (C) PCA prior to TVN normalization. (D) PCA after TVN normalization. No feature selection was applied.



0% 20% 80% 100% 40% 60% connection strength

 **Figure S5. Categorization of drug profiles using MorphEUS.** (A) Broad categorization for low dose (left) and high dose (right) profiles are represented as heatmaps. The connection strength is determined by the strength of the consensus k-nearest neighbor analysis.100% strength indicates that the nearest neighbor identified in every iteration was from the same target category while 0% indicates that the drug of interest was never a neighbor to a drug in that pathway. Drugs are listed in the rows and collection of drugs within the broad target pathways are listed in the columns. (B) Heatmaps of drug nearest neighbor pairings for low dose (left) and high dose (right) treatment. The heatmaps are described in Fig. 3. The high dose profiles were 91% accurate for assignment of broad drug category (compared to 26% for randomized categorization) with 68% accurate cross

- validation. The low dose profiles were 97% accurate for assignment of broad drug category with 62% accurate cross validation. Pairwise drug connection strength values can be found in Dataset S1 for low (Sheet 2) and high (Sheet 3) dose profiles.
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 **Figure S6. Transcriptional response of Mtb to antibacterials in genes involved in cell wall damage and biosynthesis. (**A) Expression profiles of the *iniBAC* operon in response to 16 antibacterials. Known inducers of the *iniBAC* operon isoniazid (INH), ethambutol (ETH), and pretomanid are plotted as controls (B) Expression profiles of genes involved in peptidoglycan

- biosynthesis in response to treatment with antibacterials. A box is drawn to highlight the profiles of cycloserine and moxifloxacin, which cluster together. Hierarchical clustering was performed using
- Pearson distance.





 **Figure S7. Transcriptional response of Mtb to antibacterials in genes involved in DNA damage.** Expression profiles of genes involved in the Mtb SOS response upon treatment with antibacterials. Genes involved in the Mtb SOS response were defined as *recA-lexA* regulated genes in Mtb (1-3). Hierarchical clustering was performed using Pearson distance.

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 **Figure S8. Morphological features and response to drug treatment are dependent on growth conditions.** (A) Mtb were treated with each antibiotic (or untreated control) following adaptation to different growth media: rich (blue), butyrate as the sole carbon source (orange), cholesterol as the 97 sole carbon source (cyan), and low pH in rich medium (green) as described in Methods. Red lines mark the medians, boxes mark the 25-75th percentiles, and the whiskers extend the full range, excluding outliers (n=1029-6733). (B) PCA of morphological profiles in Mtb treated with antibiotics. Mtb cells were grown in media containing carbohydrates (left) or butyrate (right) as the sole carbon source. Bedaquiline (dark blue squares) clusters close to the cell wall acting antibiotics (pink and purple shapes) in media containing carbohydrates (left) but not in butyrate (right).



0% 20% 40% 60% 80% 100% connection strength

 **Figure S9. Time and dose-dependency of profiles.** (A) Drug-drug matrix of the absolute difference in connection strengths as determined by MorphEUS analysis between low and high dose profiles with JSF-2019 applied. (B) Broad categorization of bedaquiline, moxifloxacin, 107 rifampicin, and isoniazid at 6 hours mapped onto the joint dose profiles at ~1 doubling time (Fig. 3). The 17-hour profiles are shown as a point of comparison.





**Fig. S10. Transcriptional analysis of 165 genes co-regulated by JSF-2019 and ofloxacin.** 

Hierarchical clustering by Pearson distance of the gene expression profiles for 165 genes that were

significantly co-regulated by JSF-2019 and ofloxacin by >1.5-fold.







(A, also shown in Fig. 3B) and 140 (B) classification trials. The magnitude of the difference in

connection strengths are shown in (C).

 **Table S1**. **Compounds used in this study.** Primary drug targets, off-target effects, and broad categorizations based on published studies. We determined IC90 values as the minimal concentration of drug needed to inhibit at least 90% of growth. The broad categorization of each drug was assigned based on literature review.







125 **\***Drugs that did not reach IC90

 **Table S2. Features included in analysis pipeline.** For each feature, with exception to FEATURE\_1\_count and FEATURE\_2\_count (noted in green), the median, quartile 1 (25%), quartile 3 (75%), and interquartile range were included as metrics, resulting in 94 total features. Descriptions come from the ImageJ documentation (47).





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 **Movie S1. Time-lapse imaging of** *M. smegmatis* **before, during, and after treatment with ethambutol.** RpoB-EGFP reporter *M. smegmatis* was imaged in nutrient-rich growth conditions in a constant-flow microfluidic device for 10 hours with no antibiotics, followed by a 6-hour drug 144 treatment with 9.375µg/ml of ethambutol, and followed by a 10-hour no drug recovery period.

 **Movie S2. Time-lapse imaging of** *M. smegmatis* **before, during, and after treatment with rifampicin.** RpoB-EGFP reporter *M. smegmatis* was imaged in nutrient-rich growth conditions in a constant-flow microfluidic device for 10 hours with no antibiotics, following by a 6-hour drug treatment with 75µg/ml of rifampicin, and followed by a 10-hour no drug recovery period.

 **Movie S3. Time-lapse imaging of** *M. smegmatis* **before, during, and after treatment with moxifloxacin.** RpoB-EGFP reporter *M. smegmatis* was imaged in nutrient-rich growth conditions in a constant-flow microfluidic device for 10 hours with no antibiotics, following by a 6-hour drug 152 treatment with  $0.781\mu g/ml$  of moxifloxacin, and followed by a 10-hour no drug recovery period.

 **Dataset S1 : Matrix of drug-drug connection strengths.** MorphEUS output for the joint dose (sheet 1), low dose (sheet 2), and high dose drug profiles (sheet 3). Numerical values here correspond with the heatmap representations in Figure 3B and Figure S5B, respectively.

 **Dataset S2: List of co-regulated genes between JSF-2019 and ofloxacin.** Table of genes used for hierarchical clustering in Figure S10. Functional annotation of each gene product was designated using tuberculist (http://tuberculist.epfl.ch).

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