1 Hemsley *et al.* – Supplementary Information

2

3 Persister transcriptome.

4 To characterise the physiological state of drug-tolerant cells, the transcriptome of ceftazidime-survivors was sequenced. Samples were obtained by isolating RNA from 5 stationary phase cells which had been treated with 100 x MIC ceftazidime for 24 hrs 6 at 37 °C. The transcriptome was compared to the transcriptome of bacteria which 7 8 had been grown in LB broth to mid-log phase ($OD_{600nm} = 0.5-0.6$; abbreviated as LBML) or to stationary phase (OD_{600nm} = 4.5-5.5; abbreviated as LBS). The 9 reproducibility between biological replicates was assessed by linear regression 10 analyses, where the coefficients of determination (R²) ranged between 0.9905 and 11 0.9997 between biological replicates (Fig. S1A-C). The average number of reads per 12 gene was 1,934 for mid-log phase samples (maximum of 292,702 reads). In 13 contrast, the average number of reads per gene was 1,069 for ceftazidime-survivors, 14 with a maximum of 191,687, which indicates that the overall gene expression level is 15 reduced in ceftazidime-survivors compared to actively growing cells. In LBS 16 samples, the average and maximum number of reads per gene were 1,927 and 17 1,093,163, respectively. The maximum number of reads dropped to 354,873, when 18 BTH 11628, encoding a small RNA with similarity to tmRNA, was excluded. 19

20

According to the transcriptome data, genes involved in nitrogen metabolism such as the nitric oxide reductase and the nitrate reductase (NarGHI) of the large chromosome, including the activators NarL and NarX (encoding a two component system), are induced in persister cells compared to both mid-log and stationary phase cells (Fig. S1H). *B. pseudomallei* (1) and *B. thailandensis* (2) are able to grow

anaerobically using nitrate as terminal electron acceptor. Induction of the nitrate 26 reductases in persister cell samples is surprising as no nitrate was present during 27 growth and challenge with antibiotics. Genes of the arginine deimination (ADI) 28 pathway are the most highly up-regulated genes in persisters. The first enzyme of 29 the ADI pathway, the arginine deiminase, converts L-arginine to L-citrulline and 30 ammonia (Fig. S1H). Citrulline is degraded further forming ATP, CO₂, and L-31 ornithine (3). Hence, the ADI pathway provides energy in the absence of oxygen 32 and also feeds into carbon and nitrogen metabolism. In addition, it has been 33 34 suggested that the pathway can also protect some bacteria from acidic conditions, by the production of ammonia (4). Indeed, Burkholderia pseudomallei mutants in the 35 ADI pathway have recently been described as exhibiting decreased survival rates at 36 low pH (5). The same authors described increased expression of the arcA and arcC 37 genes in certain colony morphotypes, but could not demonstrate a role for the ADI 38 pathway in survival in macrophages. B. thailandensis does not form colony 39 morphology variants to the same extend as *B. pseudomallei*. Nevertheless, the 40 results substantiate a fitness advantage of cells with an active ADI pathway. The 41 role of the arginine deimination pathway in virulence in animal models and chronic 42 disease remains to be elucidated. 43

44

The promoter prediction tool Bprom (freely available at the Softberry homepage) predicted binding sites of the transcriptional regulator Fnr in vicinity of some promoters within the denitrification pathway and the ADI pathway operons (see Fig. S1H). Fnr has been extensively studied in *E. coli*, where it mediates the transition from aerobic to anaerobic growth (6). *B. thailandensis* strain E264 possesses three proteins with homology to the *E. coli* Fnr protein, BTH_II0035, BTH_II0460, and 51 BTH_II1244. The first two proteins were expressed at slightly higher levels in 52 persisters compared to LBML samples only. However, Fnr activation in response to 53 hypoxia is achieved through changes in the redox state of an internal iron-sulfur 54 cluster (7), and not through transcriptional activation. The role of Fnr in the 55 anaerobic adaptation in *Burkholderia* remains to be established.

56

57 The download link for the persister datasets is:

58 <u>http://osslab.ex.ac.uk/downloads/RNAseq-data_Hemsley_et_al.xls</u>

59

60 Comparison with the transcriptomes of *M. tuberculosis* and *E. coli* persister

61 <u>cells.</u>

In order to identify commonalities between persisters cells from different organisms, 62 the B. thailandensis transcriptome described here was compared to reported 63 microarray data from *M. tuberculosis* persisters (6) and *E. coli* persisters (7, 8). This 64 identified 16 genes that were commonly induced in drug-tolerant B. thailandensis 65 and in *M. tuberculosis* persister cells (Fig S2). Seven of the genes encode predicted 66 inner membrane proteins, and one conserved hypothetical protein of M. 67 tuberculosis, Rv0140, has been found in culture filtrates (9). Two transcriptional 68 regulators are among the list of commonly induced genes, and a variety of stress-69 related gene product such as heat-shock proteins and a universal stress protein. 70 The latter, however, was only found to be up-regulated in persisters compared to 71 mid-log phase cells but not to stationary-phase cells, thereby highlighting the effect 72 of the reference sample in such transcriptome analyses. Only two of the commonly 73 induced genes, the heat shock protein htpX and a member of the Hsp20 family, 74 have also been described as being expressed in E. coli persisters, thereby 75

indicating a possible role of the heat shock response in persister cell formation
across various organisms. Interestingly, over-expression of *hspX* in *M. tuberculosis*has been shown to result in reduced growth rates both *in vivo* and *in vitro* (10),
thereby adding evidence to a possible role of HspX in maintaining a slow-growing
phenotype.

81

82 Role of the heat-shock response.

We could not confirm activation of the heat-shock response in drug-tolerant cells (Fig. 83 84 S3). The heat shock sigma-factor *rpoH* was expressed at 3-times higher levels in drug-tolerant cells compared to both mid-log and stationary phase cells. The heat 85 shock proteins HtpX (BTH_I0131) and two members of the Hsp20 family 86 (BTH_I2809 and BTH_I2810), as well as the chaperone protein DnaK (BTH_I1308) 87 were also induced in persisters. We also found elevated expression levels of some 88 ATP-dependent proteases such as Lon and Clp proteases in drug-tolerant cells 89 compared to mid-log phase cells, but not compared to stationary phase cells. Only 90 one copy of the Zn-dependent oligopeptidase PrIC (BTH_I1860) was 2.4-fold 91 induced in drug-tolerant cells. Finally, BTH I2241 encoding the FtsH protease 92 activity modulator HflK, and its downstream gene were up-regulated in drug-tolerant 93 cells compared to stationary phase cells. FtsH has been shown to regulate RpoH 94 levels and is important for maintaining membrane integrity (11). In *M. tuberculosis*, 95 FtsH is involved in resistance to reactive oxygen species and regulates the cell 96 division protein FtsZ (12) and in C. crescentus, ftsH mutants exhibit increased 97 sensitivity to antibiotics and form filaments (13). 98

99

100 Contribution of β -lactamases and multidrug efflux pumps.

According to our RNAseg data, four putative β -lactamases, BTH 1908, 101 BTH_II0372-0373, and BTH_II1450, were induced in drug-tolerant cells (Fig S4A). 102 The latter represents a homologue of PenA, a secreted β -lactamase that increases 103 resistance to β-lactam antibiotics when over-expressed in *B. pseudomallei* (14). 104 However, we found that the remaining ceftazidime concentration after 24 hrs 105 incubation at 37°C in the absence and presence of bacteria was approx. 95% of the 106 initial concentration (i.e. 95x MIC) in both cases, thereby ruling out loss of activity by 107 natural or bacterial degradation (see Fig S4 B&C). The expression of several 108 multidrug-efflux pumps, including the BpeAB-OprB systems and the AmrAB-OprA 109 system, was induced in persister (Fig S4A). 110

111

112 Comparison of the Two, Three and Four State Stochastic Switching models.

This section combines arguments, data and simulations to describe the relative 113 effectiveness of the two, three and four state models in accounting for the data 114 presented in the main text. In addition to changing the atmospheric oxygen levels, 115 we also determined the number of persister cells in response to differences in the 116 volume of the culture. Because these cultures were unshaken, and considering the 117 diffusion of oxygen into water and broth was governed similarly, we were able to 118 predict the oxygen levels in the broth in response to differences in culture volume 119 (Fig S5A). Figures 1B and S5B demonstrate that the number of bacterial survivors 120 under 24 hrs of ceftazidime treatment varied over about five orders of magnitude 121 between conditions of high and low oxygen supply. However, in both experiments 122 where oxygen levels were altered, either being explicitly controlled or when it was 123 varied indirectly by changing the assay volume, the frequency of survivors revealed 124 after 24 hrs of ciprofloxacin treatment stayed consistently at around 10⁻⁴. 125

Two-State Model. The two-state model (see Fig. S6A) proposes that there are two 127 phenotypically distinct states, one with (p) and one without (n) antibiotic tolerance. 128 Cells are assumed to stochastically switch between these states at rates a and b. As 129 seen in Figure 1C, kill curves for both ciprofloxacin and ceftazidime exposure 130 showed biphasic killing. These persister subpopulations however plateau at 131 132 significantly different levels. Under ciprofloxacin treatment, the slope of the second phase (corresponds to the death rate k_p) was not very steep. This, combined with the 133 134 consistency in survivor numbers to ciprofloxacin treatment after 24h under different oxygen regimes, strongly suggests that the switching rates (a and b) are not 135 changing if p is to represent the ciprofloxacin-tolerant persister state. However, the 136 very opposite is suggested under exposure to ceftazidime, where the killing efficacy 137 changes many orders of magnitude with oxygen levels (Fig. S5B). The largest 138 change that needs to be accounted for from the experiments is the difference 139 observed by changing the assay volume from either 0.25 ml or 0.5 ml to 1 ml or 2 ml. 140

141

Using the analytical solution for the two-state switching model derived by Balaban et 142 al. (15), we tested extensive regions of the parameter space summarised in Table 143 S2 to match the killing data for ceftazidime and ciprofloxacin in various regimes. 144 145 While some of these ranges appeared to overlap, a closer examination revealed that the parameter ranges must vary in order to account for the three separate killing 146 regimes as illustrated in Figure S7 for the switching rates a and b when considered 147 in tandem. In particular, comparing the ceftazidime treatment regimes revealed that 148 the switching rates between the two states must change in order to account for the 149 data. This is further illustrated in Figure S8, which plots two cases in which it is 150

assumed that $k_p = 0$ so that the other three parameters can be plotted. Very clear parameter changes are indicated by the separation of the parameter sets between high and low survivor numbers under ceftazidime treatment at different volumes. Thus the consistency of survivors from the ciprofloxacin treatment suggests there needs to be at least one additional subpopulation and hence a phenotypic state in the model. We conclude that at least a three-state stochastic switching model is required to explain the data.

158

159 Three-State Model. The most general three-state stochastic switching model is presented in Figure S4B, with presumably two persister states p_1 and p_2 in addition 160 to the fully susceptible *n* state. For convenience, we chose the p_2 persisters to be 161 those tolerant to ciprofloxacin. Kill curve data and the oxygen controlled assays 162 (whether through volume or direct oxygen control) indicate that the killing of this 163 subpopulation is seemingly independent of the oxygen content. This suggests that, 164 independent of the oxygen concentration, this subpopulation should only be reduced 165 to about 10⁻⁴ after 24 hrs of ciprofloxacin treatment. We know that the number of 166 ceftazidime persisters is dependent on the oxygen level available. Thus the rates a 167 and b are likely to vary under different oxygen conditions as the population size of 168 anaerobic cells (p_1) needs to be mediated by oxygen. Therefore, we conclude that 169 170 the switching rates in Figure S4B have the property that f=d and e=c (giving us the new model in Fig. S6C). Otherwise the consistency in the number of cells surviving 171 ciprofloxacin treatment would also be oxygen dependent. To confirm this argument, 172 parameter regimes of the ODE model presented in Figure S6C were extensively 173 searched for solutions that could fulfil both the variability in ceftazidime-killing under 174 different oxygen conditions and the consistent ciprofloxacin-killing. In each search, 175

the rates *c* and *d* were fixed to the best possible fit to the kill curve of ciprofloxacin at 176 a 1 ml assay volume (see Fig. S5B). The other rates were then seeded over varying 177 ranges and random walks were conducted of up to 1,000 steps searching for 178 solutions which would fulfil the requisite high and low survivor numbers under 179 ceftazidime-killing at various assay volumes. In no cases was a parameter set found 180 which could even approach the total low number of survivors that a 0.5 ml assay 181 volume would find (between 7.6 x 10^{-7} and 1.7 x 10^{-6} after 24 hrs ceftazidime). This is 182 not surprising as the necessarily non-zero rates of *c* and *d* cause a buffering effect. 183 184 These simulations did assume that the ciprofloxacin-tolerant persisters possess a similar tolerance to the ceftazidime drug as the anaerobic cells. The model could 185 account for the data if it is assumed that the drug tolerance of the ciprofloxacin-186 tolerant cells with respect to ceftazidime exposure is oxygen dependent. However 187 the model then loses any real explanatory or predictive power and becomes a purely 188 fitting based exercise with no indication of how the behaviour may or may not 189 change under different conditions. 190

191

Thus while the three-state model can capture some of the features of the experiments, it is unable to simultaneously capture all of the possible killing under different conditions while including all the features which have been indicated to be present from the experiments. The most evident being the independence of the number of ciprofloxacin survivors from oxygen conditions.

197

Four State Model. The four-state model presented in Figure 3A proposes the incorporation of two stochastic switches into each cell. The switching rates between ciprofloxacin non-tolerance (n or n) and tolerance (p or p) is assumed to be

independent of oxygen. Thus, independent of the oxygen content of the 201 environment, the number of ciprofloxacin persisters which survive treatment is 202 consistent. However the other switch between a presumably aerobic or anaerobic 203 metabolism is assumed to depend on the oxygen level of the environment. This new 204 model can now fit all the data we have presented as indicated in Figure 3B. Indeed, 205 in that example, only the switch out rate b from the anaerobic state needed to be 206 modified to account for the different killing profiles observed under different levels of 207 208 oxygen.

209

210 Burkholderia pseudomallei anaerobic nitrate respiration (denitrification)

B. thailandensis and B. pseudomallei can grow anaerobically in the presence of 211 nitrate (1, 2), using the denitrification pathway. Denitrification requires a series of 212 reductase enzyme; nitrate reductase (NarGHI), nitrite reductase (Nir), nitric oxide 213 reductase (Nor) and nitrous oxide (Nos), to sequentially reduce nitrate to dinitrogen 214 gas (16). The first step in the denitrification (also referred to as anaerobic nitrate 215 respiration pathway) utilises a membrane-bound nitrate reductase (NarGHI), 216 required for the reduction of nitrate to nitrite and generation of a PMF (17). B. 217 pseudomallei encodes two membrane-bound nitrate reductase enzymes narGHJI 218 (BPSL2309-2312), predicted to be the main Nar required for anaerobic respiration 219 220 and *narZYWV* (BPSS1156-1159) displaying homology to the cryptic NarZYWV found in *E. coli* and *Salmonella*. Both of these gene clusters share high sequence identity 221 (90-99%) with respective homologous gene clusters in B. thailandensis BTH_I1851-222 BTH_I1854 (narGHJI) and BTH_II1249-1252 (narZYWV). The BTH_I11851-1854 223 gene cluster was shown to be upregulated in persister cells, whereas the second 224 nitrate reductase BTH_II1249-1252 (narZYWV) was downregulated. 225

In order to determine the role of anaerobic respiration in persister cell formation a 227 narG (BPSL2309) deletion mutant was constructed using the pDM4 suicide vector. 228 The deletion of *narG* was confirmed using two sets of PCR, one using primers 229 binding to a 300 bp internal region of the gene (not shown) and one using primers 230 binding 300 bp up and downstream of the target gene, generating a 600 bp product 231 in the mutant and a much larger product (approximately 3,500 bp) in the wild-type. 232 No wild-type PCR product was detected under the cycle conditions used in this study 233 234 (Fig. S9A). Deletion of *narG* ($\Delta narG$) prevented growth of *B. pseudomallei* anaerobically in the presence of nitrate, but did not affect aerobic growth in either M9 235 minimal media or L-broth (Fig. S9 B – C and E). In order to confirm the $\Delta narG$ 236 mutant was deficient in Nar activity a Griess reaction (Promega) was performed on 237 aerobically grown wild-type and mutant B. pseudomallei cultures (grown in M9 238 minimal media supplemented with 20 mM sodium succinate and 20 mM sodium 239 nitrate). The concentration of nitrite produced during aerobic growth and therefore 240 relative Nar activity was determined using the Griess reaction and a nitrite standard 241 curve. Only the wild-type accumulated significant amounts of nitrite after 8 hours of 242 growth under aerobic conditions (Fig. S9D). After 24 hours of aerobic growth in the 243 presence of nitrate wild-type B. pseudomallei accumulated around 256 µM nitrite 244

whereas the $\Delta narG$ mutant accumulated only 7 µM nitrite. This indicated that NarGHI may be active under both aerobic and anaerobic conditions. The results from these experiments indicate BPSL2309-2312 (*narGHJI*) encodes the main nitrate reductase in *B. pseudomallei* required for anaerobic respiration and nitrate reductase activity.

249

250 **Regrowth rates and phenotypes on plates**

We observed differences in the re-growth rates between ceftazidime and 251 ciprofloxacin-survivors (see Fig. S10), with the former forming colonies of uniform 252 size on LB agar plates after 24 hrs incubation at 37 °C, whereas ciprofloxacin 253 survivors requiring 48 hrs incubation before visible colonies of variable sizes are 254 formed on plate. This suggests that ceftazidime-survivors seem to be able to rapidly 255 resume growth, which might be the result of the maintained metabolic activity. The 256 differences in the re-growth rates between ciprofloxacin- and ceftazidime survivors 257 also indicate that they might have reached different dormancy levels. It has 258 previously been suggested that dormancy levels can be divided into three stages, 259 which coincide with arrests of growth, DNA replication, and protein synthesis, 260 respectively, and that these levels correlate with resistance to different classes of 261 antibiotics i.e. β -lactams, quinolones, and aminoglycosides (18). 262

263

264 SUPPLEMENTARY METHODS

265

Anaerobic growth experiments. B. pseudomallei anaerobic growth studies were 266 conducted using the BD GasPak EZ incubation system and two Gas Pak EZ 267 anaerobic container system sachets with indicator. M9 minimal media supplemented 268 with nitrate was used for the anaerobic growth studies and Griess reaction. M9 269 minimal media was supplemented with or without sodium nitrate (NaNO₃) or sodium 270 nitrite (NaNO₂) (0-20 mM) and 20 mM succinate as a carbon and electron source. 271 M9 media contained 2 mM MgSO₄, 0.1 mM CaCl₂, 20 % M9 salts (5 x stock solution; 272 85.5 gl-1 Na₂HPO₄, 15 gl-1 KH₂PO₄, 2.5 gl-1 NaCl, 5 gl-1 NH₄Cl). Wild-type B. 273 pseudomallei and $\Delta narG$ overnight cultures were standardised and inoculated into 274 M9 minimal media supplemented with or without 20 mM sodium nitrate. CFU counts 275

were performed by spot plating serial dilutions onto LB agar plates incubated aerobically at 37 °C.

278

Nitrate reductase activity assay. The concentration of nitrite produced throughout aerobic growth in M9 minimal media was measured using the Griess Reagent system (Promega). 1 mL samples were taken throughout the growth cycle and frozen at -80 °C prior to performing the Griess reaction. A nitrite standard curve was generated for each experiment to ensure accurate estimations of nitrite concentration in the sample medium. Three independent biological replicates were used, each with three technical replicates.

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SUPPLEMENTARY FIGURES



Fig. S1. RNAseq data. (A-C) Reproducibility between biological replicates. Representative scatter plots of read count data of two biological replicates of LBML samples (A), LBS samples (B), and ceftazidime-survivors (C), respectively. (D-F) Comparison of different conditions. Scatter plots of average read counts of LBS vs. LBML samples (D), ceftazidime-survivors vs. LBML samples (E), and ceftazidimesurvivors vs. LBS samples (F), respectively. In A-F, coefficients of determination (R²)

300 were calculated based on linear regression analyses. (G) Numbers within the Venn diagram represent genes that are specifically expressed under one or two 301 conditions, or which are equally expressed under all three conditions (center of 302 diagram). (H) Pathway analysis of the 306 genes that were specifically expressed in 303 ceftazidime survivors revealed that all genes within the nitrate respiration 304 (denitrification) pathway and the arginine deimination pathway were significantly 305 upregulated in persisters compared to LBML and LBS samples. These pathways are 306 controlled by various external stimuli through transcriptional regulators. 307

Protein function	Bt-P/ML	Bt-P/S	TB-P	EC-P	Bt ID	TB ID	Ec ID	Localisation
Heat shock protein HtpX					BTH_10131	Rv0563	htpX	IM
Sulphate permease family of anion transporters					BTH_11051	Rv1707	<i>ychM</i>	IM
Peptidase, M50 family / conserved transmembrane protein					BTH_11903	Rv2625c	rseP*	IM
Transcriptional regulator, MerR family					BTH_12208	Rv3334	zntR	CP
Transcriptional regulator, ArsR family					BTH_II0561	Rv2034	arsR	CP
Alanine dehydrogenase / pyridine nucleotide transhydrogenase					BTH_110922	Rv2780	pntA	IM
Heat shock protein, HSP20 family					BTH_110924	Rv0251c	ibpB	CP
Universal stress protein, putative					BTH_II1268	Rv2623	uspE	CP
AMP-binding domain protein / put. acyl-coA synthase / carnitine-CoA ligase					BTH_II1686	Rv3515c	caiC	CP
Domain of unknown function / conserved hypothetical protein				na	BTH_II1714	Rv0140	na	Secreted
Methylmalonate-semialdehyde dehydrogenase / phenylacetaldehyde dehydrogenase					BTH_II1801	Rv0753c	fea B	CP
Pyochelin synthetase / conserved hypothetical protein					BTH_II1828	Rv0100	entF	CP/IM
4-Aminobutyrate aminotransferase / put. lysine-(epsilon) aminotransferase					BTH_II2119	Rv3290c	puuE	CP
Proline iminopeptidase / predicted aminoacrylate hydrolase					BTH_II2193	Rv0840c	rutD*	CP
2-Oxoisovalerate dehydrogenase subunit / pyruvate dehydrogenase E1				na	BTH_II2304	Rv2497c	na	CP
Sulphate permease family of anion transporters					BTH_II2316	Rv1707	ychM	IM
Stress response protein / HSP20 family				na	BTH II2321	Rv0251c	na	IM
							partial	



Fig. S2. Common themes in persister cell transcriptomes. Transcriptome 311 datasets of *B. thailandensis* persisters (Bt; this study), *M. tuberculosis* persisters (TB, 312 Keren et al. 2011), and E. coli persisters (Ec; Keren et al. 2006) were analysed for 313 314 similarities. Expression ratios refer to Bt persisters vs. mid-log cultures (Bt-P/ML) and stationary phase cultures (Bt-P/S), respectively, TB persisters revealed with D-315 cycloserine vs. mid-log phase cells before antibiotic challenge (TB-P), and E. coli 316 persisters revealed with ampicillin vs. mid-log phase cells before antibiotic challenge 317 (Ec-P). The genome of *B. thailandensis* was searched for orthologues of reported TB 318 persister genes using BLASTP and the RNAseq dataset was analysed for 319 expression ratios of these 17 homologues. Note that one TB gene, Rv1707, has two 320 orthologues in B. thailandensis. Next, the genome of E. coli strain MG1655 was 321 searched for orthologues of the common persister genes using BLASTP and a 322 published microarray dataset (Keren et al. 2006) was analysed for expression ratios 323 in E. coli. A second E. coli dataset (Shah et al. 2006), which also includes a 324 comparison of persisters vs. stationary phase cells, was also analysed; none of the 325 orthologues exhibited induced expression levels. Details on the function and 326 subcellular localisation of the encoded protein were obtained from public databases. 327

- 328 Colour coding for expression ratios: orange = ratio < 0.5; grey = ratio 0.5 to 2; green
- 329 = ratio >2.



Fig S3. Expression of the heat shock regulon in *B. thailandensis* persister cells. Heat map of expression ratios of genes described to be part of the heat shock response in stationary phase samples vs. mid-log samples (S/ML), persisters vs. mid-log samples (P/ML) and persisters vs. stationary phase samples (P/S). Colour coding: orange = ratio < 0.5; grey = ratio 0.5 to 2; green = ratio >2.

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Fig. S5. (A) One dimensional solution of O2 diffusion into water. It was assumed 354 that the temperature of the water was 37 °C (diffusion coefficient of oxygen in water 355 at this temperature is used) and that the initial oxygen content of the water was 356 depleted to zero. The lowest line is the diffusion level after 1 hour and they progress 357 upwards at one hour intervals for each line up to 24 hrs. The vertical lines represent 358 the depth of medium in a 1.9 cm² (surface area of the base) cylindrical well plate at 359 assay volumes of 0.25 ml, 0.5 ml, 1 ml and 2 ml. (B) The generation of an oxygen-360 limited bottom layer during static incubation was assessed by using different 361 volumes of a mixture of bacteria and antibiotics and enumerating survival 362 frequencies after 24 hrs incubation. The volumes correspond to a depth of 1.3 mm 363 (0.25 ml), 2.6 mm (0.5 ml), 5.25 mm (1 ml), and 10.5 mm (2 ml) of the assay mixture 364 inside the wells. In all experiments, bacteria grown to stationary phase were used 365 and challenged with antibiotics under static incubation at 37 °C. Error bars represent 366 the standard deviation over the mean from at least three independent experiments. 367



370

Fig. S6. Two- and three-state switching models. In all models presented, the 371 different phenotypic population states are represented by circles and the switching 372 rates between them by arrows. Under antibiotic treatment, a population state i 373 declines (or grows if the drug is ineffective) at a rate k_i . (A) The original two-state 374 stochastic switching model with a population of normal (*n*) and persister (*p*) cells and 375 switching rates between them (a and b) (8). (B) The general three-state switching 376 model with normal cells (n) and two distinct persister subtypes (p1 and p2). (C) The 377 modified three-state model with p1 cells being tolerant to ceftazidime and p2 cells 378 being tolerant to ciprofloxacin. Blue arrows indicate that the switching rates between 379 the normal susceptible cells *n* and *p1* (rates *a* and *b*) depend on the oxygen 380 concentration. Red arrows indicate that the size of the ciprofloxacin-tolerant 381 population *p2* is independent of the oxygen level. This is incorporated into the model 382 by setting the switching rates f = d and e = c. 383

384



Fig. S7. Parameter sweeps of the two-state model plotting only a and b. In all 389 plots, the green squares correspond to parameter sets that yield a low survivor 390 number at 24 hrs of between 7.6 \times 10⁻⁷ and 1.7 \times 10⁻⁶, which corresponds to the 391 survivor frequencies found after ceftazidime treatment in the 0.5 ml assay volume. 392 The blue circles correspond to parameter sets that yield a high survivor number at 24 393 hrs and 48hrs of between 0.0238 and 0.0302 and 0.0106 and 0.037 respectively. 394 These correspond to the survivor frequencies found after ceftazidime treatment in 395 the 1 ml assay volume. (A) Initial population of p = 0 and n = 1. (B) Initial population 396 of p = 0.01 and n = 0.99. (C) Initial population of p = 0.05 and n = 0.95. (D) Initial 397 population of $p = 10^{-4}$ and $n = 1 \cdot 10^{-4}$. Note that in all plots the two parameter sets are 398 distinct and separate. 399





Fig. S8. Extended parameter sweeps of the two-state model. In both plots, each 405 blue dot represents a distinct set of a, b and kn parameters of the two-state model 406 (see Fig. S4A), which predict a survival frequency between 2.4x10⁻² and 3.0x10⁻² 407 after 24 hrs. This corresponds to the survival frequencies observed after 24 hrs 408 ceftazidime treatment in 1 ml volumes (see Fig. S2B). Every green square is a 409 distinct set of a, b and kn parameters of the two-state model resulting in a survival 410 frequency between 7.6x10-7 and 1.7x10-6 after 24 hrs. This corresponds to the 411 survival frequencies observed after 24 hrs ceftazidime treatment in 0.5 ml volumes 412

(see Fig. S1B). **(A)** Parameter sweeps over ranges of *a*, *b* and k_n (assume $k_p = 0$), assuming an initial population of $p = 10^{-6}$ and $n = 1 \cdot 10^{-6}$. **(B)** Parameter sweeps over ranges of *a*, *b* and k_n (assume $k_p = 0$), assuming an initial population of p = 0 and n =1. **(C)** Solutions of the two-state model which conform to high and low survivor counts observed under different assay volumes *in vitro*. For these solutions, it is assumed that $k_p = 0$ and that at t = 0, n = 1 and p = 0. The parameter sweeps were run over the parameter ranges depicted on the axes.

420



Fig. S9. Deletion of narG in B. pseudomallei prevents anaerobic growth on 423 nitrate and causes a significant reduction in nitrate reductase activity under 424 425 **aerobic conditions.** An in-frame nitrate reductase deletion mutant ($\Delta narG$) was created using the suicide vector pDM4 carrying the BPSL2309 knockout cassette 426 (pD2309). (A) PCR confirmation of the $\Delta narG$ deletion mutant using primers binding 427 300 bp up and downstream of the target gene (BPSL2309 - narG), generating a 600 428 bp product in the mutant and a much larger band in the wild-type (over 3,500 bp), not 429 detected under the PCR conditions used. Lane 1 - 1 kb DNA ladder; lane 2 -430

negative control (H₂O); lane 3 – pD2309; lane 4 – wild-type *B. pseudomallei* lysate; 431 lane 5 – Δ *narG* colony lysate. Lack of a WT band and presence of 600 bp band for 432 Δ narG confirms the deletion of BPSL2309. (B) Wild-type B. pseudomallei (black 433 bars) and $\Delta narG$ (grey bars) were grown anaerobically in the presence or absence of 434 20 mM sodium nitrate. No anaerobic growth was seen for $\Delta narG$. Data is the 435 average of three independent biological replicates. Asterisks (**) denote statistically 436 significant difference between WT and $\Delta narG$ (P < 0.01). (C) Aerobic growth in M9 437 minimal media supplemented with 20 mM sodium succinate and 20 mM sodium 438 nitrate for wild-type (filled squares) and $\Delta narG$ (filled circles). Samples were taken 439 throughout aerobic growth to determine the level of nitrate reductase activity in both 440 B. pseudomallei cultures. (D) Relative nitrate reductase (NAR) activity was 441 measured by determining the concentration of nitrite in the media using the Griess 442 reagent system. Concentration of nitrite produced during aerobic growth was 443 measured for the wild-type (black bars) or $\Delta narG$ (grey bars). The $\Delta narG$ deletion 444 mutant could not reduce nitrate to nitrite, confirming a lack of NAR activity in this 445 mutant. (E) Aerobic growth of wild-type *B. pseudomallei* and $\Delta narG$ in L-broth. 446 Results presented are the mean of three biological replicates, each with two 447 technical replicates used when performing the Griess reaction. Error bars ± 448 standard deviation (SD). 449



Fig. S10. Phenotypes of different types of persister cells. *B. thailandensis* strain E264 from stationary growth phase was challenged with 100 x MIC ceftazidime or 10 x MIC ciprofloxacin for 24 hours. Images represent LB agar plates with colonies derived from serial dilutions of washed surviving cells. The ceftazidime-derived persister plate shown in left panel was incubated for 24 hours at 37 °C, whereas the ciprofloxacin-derived plate shown in the right panel had been incubated for 48 hours at 37 °C.

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Strain	Description	Reference
Burkholderia thai	landensis	
E264	Environmental isolate, sequenced strain	(19, 20)
DW503	E264 efflux pump mutant ($\Delta amrR$ -oprA, Km ^S Gm ^S	(21)
	Sm ^s)	
Phuket 4W-1	Water isolate from Thailand	(22)
CDC3015869	Clinical isolate from Texas	(23)
CDC2721121	Clinical isolate from Louisiana	(23)
Burkholderia pse	udomallei	
K96243	Clinical isolate from Thailand, sequenced strain	(24)
K96243 ∆ <i>nar</i> G	Deletion of BPSL2309 in strain K96243	This study
Escherichia coli		
MG1655	K-12 derivative ($F^{-}\lambda^{-}$ <i>ilvG</i> ⁻ <i>rfb</i> -50 <i>rph</i> -1)	(25)

Table S1. Bacterial strains used in this study.

Table S2. Parameter sweeps of the two-state switching model. The ranges shown are those that either conform to the high survivor numbers at 24 hrs and 48 hrs for the ceftazidime killing under 1 ml assay volume (High cefta) or the 24 hrs low survivor numbers at 0.5 ml assay volume (Low cefta) or the ciprofloxacin kill curve (Ciprofloxacin). Parameter ranges tested: $a \in (0, 10), b \in (0, 10), k_n \in (-5, 0), k_p \in (-0.05, 0)$.

Initial	Tested Range	Parameter ranges	found to be within error ranges of high or			
Conditions	Solution	low survivor numbe	pers under ceftazidime treatment or to have			
(n , p)	Ranges	agreed with ciprofloxacin kill curve (min, max)				
		a b	k _n k _p			
(1,0)	High	0.002 , 0 , 0.05	-0.5 ,016 -0.05 , 0			
	ceftazidime	0.04				
	Low	10 ⁻⁶ , 10 ⁻⁷ , 10	-5 , -0.6 -0.05 , 0			
	ceftazidime	10				
	Ciprofloxacin	0.02 , 0,0.12	-2.2 , -1.7 -0.1 , 0			
		0.03				
(0.99 , 0.01)	High	0,0.04 0,0.05	-0.5 , -0.17 -0.05 , 0			
	ceftazidime					

	Low	10 ⁻⁷ ,	0.04 , 10	-5 , -0.6	-0.05 , 0
	ceftazidime	10			
	Ciprofloxacin	0.01 ,	0.02 , 0.1	-2.2 , -2	-0.1 , -0.02
		0.03			
(0.9999 , 10 ⁻⁴)	High	0.01 ,	0,0.04	-5 , -0.25	-0.05 , 0
	ceftazidime	0.4			
	Low	10 ⁻⁷ ,	0.02 , 10	-5 , -0.6	-0.05 , 0
	coftazidimo	10			
	Centaziuline	10			
	Ciprofloxacin	None	None	None	None
(0.95 , 0.05)	Ciprofloxacin High	None 0 , 0.02	None 0 , 0.05	None -0.5 , -0.21	None -0.05 , 0
(0.95 , 0.05)	Ciprofloxacin High ceftazidime	None 0 , 0.02	None 0 , 0.05	None -0.5 , -0.21	None -0.05 , 0
(0.95 , 0.05)	Ciprofloxacin High ceftazidime Low	None 0 , 0.02 10 ⁻⁷ ,	None 0 , 0.05 0.4, 10	None -0.5 , -0.21 -5 , -0.6	None -0.05 , 0 -0.05 , 0
(0.95 , 0.05)	Ciprofloxacin High ceftazidime Low ceftazidime	None 0,0.02 10 ⁻⁷ , 10	None 0 , 0.05 0.4, 10	None -0.5 , -0.21 -5 , -0.6	None -0.05 , 0 -0.05 , 0
(0.95 , 0.05)	Ciprofloxacin High ceftazidime Low ceftazidime Ciprofloxacin	None 0,0.02 10 ⁻⁷ , 10 None	None 0 , 0.05 0.4, 10 None	None -0.5 , -0.21 -5 , -0.6 None	None -0.05 , 0 -0.05 , 0 None

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