

# 1 Rapid antimicrobial sensitivity testing by single cell 2 nanoscale optical interference

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## 10 Abstract

11 Growing antimicrobial resistance (AMR) is a serious global threat to human health, with  
12 estimates of AMR leading to 10 million deaths per year and costing the global economy  
13 \$100tn by 2050<sup>1,2</sup>. Current methods to detect resistance include phenotypic antibiotic  
14 sensitivity testing (AST) which measures bacterial growth and is therefore hampered by slow  
15 time to result (~12-24 hours). Therefore new rapid phenotypic methods for AST are urgently  
16 needed<sup>3</sup>. Here we describe a novel method for detecting phenotypic antibiotic resistance in  
17 ~45 minutes, capable of detecting single bacteria. The method uses a sensitive laser and  
18 detector system to measure nanoscale optical interference of single bacterial cells present in  
19 media, with simple sample preparation. This provides a read out of bacterial antibiotic  
20 resistance by detecting growth (resistant) or death (sensitive), much faster than current  
21 methods. We demonstrate the potential of this technique by determining resistance in both  
22 lab and clinical strains of *E. coli*, a key species for clinically burdensome urinary tract  
23 infections. This work provides the basis for a simple and fast diagnostic tool to detect  
24 antibiotic resistance in bacteria, reducing the health and economic burdens of AMR.

25

## 26 Main

27 Antimicrobial resistance (AMR) is steadily increasing and poses a major threat to global  
28 health. The increase in AMR has been caused by several factors including the overuse of  
29 antibiotics<sup>4</sup>. Despite the growth of AMR, methods for antibiotic susceptibility testing (AST)  
30 have remained relatively unchanged for several decades. In common AST methods bacterial  
31 growth is used as a measure of sensitivity to antibiotics, determined directly by an increase in

32 media turbidity (the number of bacteria) or indirectly by the release of fluorescent  
33 metabolites. These phenotypic methods provide *in vitro* confirmation of resistances in  
34 isolated bacterial species, which are inferred from known resistance genes in genetic  
35 methods. However phenotypic methods are inherently limited by the speed of bacterial  
36 growth (for example, the doubling time of *E. coli* is 20 minutes, whereas *M. tuberculosis* is  
37 15-12 hours), meaning these methods require long culture times (12-24 hours, or longer for  
38 some species) for an observable change to occur. These delays result in empirical prescribing  
39 of antibiotics for patients instead of targeted treatment, which has been shown to increase  
40 mortality from sepsis fivefold<sup>5</sup>, in addition to being a driver of resistance. Having access to  
41 the identity and antibiogram of the pathogen just a few hours earlier could avoid unnecessary  
42 costs associated with inappropriate prescribing, increase patient welfare, and reduce the  
43 effects of AMR<sup>6,7</sup>. Therefore to reduce the damaging effects of AMR, we require solutions in  
44 the form of novel diagnostic tools to detect resistance and improve antibiotic stewardship,  
45 surveillance and patient management<sup>8</sup>.

46

47 Recent developments in this field have exploited single cell methods for faster and more  
48 sensitive detection of antibiotic resistance. This has been achieved by miniaturising the  
49 volume observed using microfluidics<sup>9-11</sup>, measuring mass or mechanical changes<sup>11-14</sup>, or by  
50 exploiting machine learning techniques for video tracking analysis of single cells<sup>15-17</sup>. Despite  
51 advances in the detection limit, and speed of testing, these are mostly complex set-ups, which  
52 remain far from point of care.

53

54 Here we report a novel optical method for rapid detection of antibiotic resistance in bacterial  
55 solutions with single cell resolution. This method uses a laser and sensitive photodetector to  
56 measure the effect of antibiotics on bacterial growth, as briefly described here. A reflective  
57 surface (small cantilever) is immersed in filtered growth media, off which a laser is reflected  
58 onto a photodiode detector (Figure 1a). In media without bacteria we observe no movement  
59 in the laser (Figure 1b). On inoculation with bacteria, bacteria free in the growth media move  
60 through the path of the laser. This movement interferes with the laser beam, causing it to shift  
61 on the detector, observable as peaks in the signal (Figure 1c). On addition of antibiotic to the  
62 media, cell death occurs in sensitive bacteria, and fewer bacteria are detected passing through  
63 the laser. This results in a decrease in the number of peaks after ~45 minutes (Figure 1d).

64

65 To determine the origin of the peaks in the signal, we reduced the bacterial concentration  
66 level to  $\sim 10^5$  CFU (colony forming units, a standard measure of bacterial concentration). At  
67 this concentration individual peaks within the signal can be observed (Figure 2a). When a  
68 single bacterium is tracked optically crossing the path of the laser (Figure 2b, blue circle), a  
69 corresponding peak in the signal can be observed in the data (Figure 2c). These peaks are of  
70 varying width and amplitude, due to differing angle and distance at which the bacteria pass  
71 through the laser. As more bacteria are added to the system (i.e. increasing CFU), the number  
72 of peaks in the signal also increases (Figure 2d), indicating that it is the bacteria giving rise to  
73 the signal.

74

75 We have shown that we can link the number of peaks observed to the number of viable  
76 bacteria in solution, which we can exploit to determine antibiotic resistance. If we determine  
77 the number of peaks (or bacterial crossings) at distinct time points during an experiment (for  
78 example ‘media only’ (blue box), ‘inoculated media’ (green box), ‘inoculated media  
79 containing antibiotic’ (red box)) (SI Figure 1), we can see a distinct pattern where bacterial  
80 crossings increase on addition of bacteria to the system (Figure 3a, at blue dotted line), and  
81 decrease around 45 minutes (about two replication cycles for *E. coli*) after the addition of  
82 antibiotic (yellow dotted line) in the case of sensitive strain. This pattern is not observed in a  
83 control with solution added containing no antibiotic (SI Figure 2). To note is that the two  
84 peaks observed in the signal which correspond to the addition of bacteria and antibiotic  
85 (Figure 3a, blue and yellow dotted lines, respectively) occur due to mixing of the system.  
86 These peaks settle to a baseline and are observed in control experiments (SI Figure 2, points  
87 ‘3’ and ‘4’).

88

89 Using this method we can differentiate sensitive and resistant strains of *E. coli*. As described  
90 above, we observe a reduction in signal after addition of antibiotic for sensitive strains  
91 (Figure 3a, green); for resistant strains, there is an increase in signal (Figure 3a, red). Though  
92 the trend remains the same, the magnitude of the signal change can vary (SI Figure 3a) based  
93 on multiple factors which effect growth rates, including inoculant concentration, strain, and  
94 temperature, for example. We therefore normalise the data to the baseline before the addition  
95 of antibiotic when comparing between experiments ( $S_{\text{baseline}}$ ) (SI Figure 3b).

96

97 To obtain a systematic readout of antibiotic sensitivity across experiments, including multiple  
98 strains and antibiotics, we obtain a normalised measure of bacterial growth as follows. We

99 define antibiotic sensitivity as  $r_{\text{sensitivity}}$ : the ratio of  $S_{\text{baseline}}$  and 45 minutes post-antibiotic  
100 treatment ( $S_{\text{antibiotic}}$ ), shaded blue in Figure 3a.  $r_{\text{sensitivity}}$  provides a binary readout of  
101 sensitivity,  $r_{\text{sensitivity}} \leq 1$  indicates cell death or inhibition of bacterial growth, and sensitivity to  
102 the antibiotic in solution;  $r_{\text{sensitivity}} > 1$  indicates bacterial growth, and therefore resistance to  
103 the antibiotic used. This method allows for both bactericidal and bacteriostatic antibiotics to  
104 be used, as  $r_{\text{sensitivity}} < 1$  indicates a decrease in cell number, or cell death (bactericidal);  
105  $r_{\text{sensitivity}} = 1$  would indicate inhibition of growth, but little cell death (bacteriostatic). For  
106 Figure 3a with ampicillin,  $r_{\text{sensitivity}} = 0.5$  for the green strain (sensitive) and  $r_{\text{sensitivity}} = 1.1$  for  
107 the red strain (resistant). For kanamycin,  $r_{\text{sensitivity}} = 0.92$  for a sensitive strain and  $r_{\text{sensitivity}} =$   
108 2.0 for a resistant strain (green and red, respectively SI Figure 4).

109

110 Having shown that we can use  $r_{\text{sensitivity}}$  as a measure of bacterial sensitivity, we now apply  
111 this method to a range of concentrations of ampicillin to determine the minimum inhibitory  
112 concentration (MIC) for the *E.coli* strain BL21 (Figure 3b). The MIC value is defined as the  
113 lowest concentration of an antibiotic that will inhibit the visible growth of a bacterial strain<sup>18</sup>,  
114 and is used to inform clinical breakpoints and provide patient-dose information for  
115 prescribing treatment. At low ampicillin concentrations (0-12.5  $\mu\text{g/mL}$ )  $r_{\text{sensitivity}} > 1$ , however  
116 at increased ampicillin concentrations (50-125  $\mu\text{g/mL}$ )  $r_{\text{sensitivity}} < 1$ . This indicates an MIC of  
117 12.5-50  $\mu\text{g/mL}$  ampicillin for this strain. This result is within the range determined by broth  
118 microdilution, the gold standard method (8-16  $\mu\text{g/mL}$ ). Despite difficulties in variability of  
119 measuring MICs<sup>19,20</sup>, these values are used by clinicians when making decisions about patient  
120 care (antibiotic selection and dosing), and hence are an important result for any new  
121 diagnostic tool to accurately measure.

122

123 Uropathogenic *E. coli* (UPEC) is the leading cause of urinary tract infections (UTIs)<sup>21</sup>, and is  
124 clinically burdensome across the globe. AMR has increased in UTIs and hence represents an  
125 excellent clinical target for a new diagnostic tool. Here we demonstrate potential for the  
126 optical interference method by testing on an *E. coli* clinical isolate. As shown in Figure 3c,  
127 treatment of the clinical isolate with 125  $\mu\text{g/mL}$  ampicillin and trimethoprim resulted in no  
128 decrease in signal, and gave  $r_{\text{sensitivity}} > 1$  within 45 minutes (Figure 3d). This was confirmed  
129 by broth microdilution (resistance  $>256 \mu\text{g/mL}$  ampicillin and trimethoprim). These detected  
130 resistances were in agreement with the resistance spectrum obtained from the hospital (Great  
131 Ormond Street Hospital, London) measured by the gold standard method in the clinical  
132 laboratory (SI Table 1). This study demonstrates the ability of this method to successfully

133 carry out an AST for a strain of bacteria isolated from a patient within 45 minutes of the  
134 addition of antibiotic.

135

136 To conclude, in the face of AMR novel rapid methods to detect resistance in bacteria are  
137 needed to prevent its further spread and development. We have shown that our novel optical  
138 interference method can rapidly differentiate between resistant and sensitive phenotypes in  
139 lab and clinical strains of *E. coli* and determine MIC values to the same range as current gold  
140 standard methods. We obtain a read out of bacterial sensitivity within ~45 minutes of the  
141 addition of antibiotic. This method lends itself to miniaturisation and automation, requiring a  
142 stable reflective surface which could be embedded within a 96-well plate for automated  
143 reading, with a laser and photodetector readout. This method can be exploited as a new rapid  
144 phenotypic method for AST, to provide these time-critical results to inform patient care and  
145 antibiotic stewardship.

## 146 **Methods**

147

148 Experimental method:

149 A stiff AC160 TS cantilever ( $k = 26$  N/m; Olympus, Japan) was loaded onto an AFM head  
150 (JPK Nanowizard 3 ULTRA Speed; JPK Instruments, Germany) and immersed in filtered  
151 Luria Broth (LB; Sigma-Aldrich, USA) in a 35 mm diameter glass bottom petri dish (WillCo  
152 Wells, Netherlands). The cantilever spring constant was calibrated using the thermal noise  
153 method in the JPK software to convert vertical deflection from volts to nm. The cantilever  
154 was allowed to equilibrate for 15 minutes, during which time vertical deflection of the laser  
155 was measured. The LB media was then inoculated with bacteria to a constant concentration  
156 ( $\sim 10^5$  CFU) and recording was started again for another 40 minutes to obtain pre-antibiotic  
157 baseline. Antibiotic solution was then added to directly to the LB + bacteria solution to a  
158 desired final concentration, and deflection recording was then measured.

159

160 During experiments only the real-time scan function was used to monitor vertical deflection  
161 of the laser. Experiments were conducted at 28°C in an acoustic isolation hood. Prior to the  
162 start of the experiments, the AFM laser was left on for ~2 hours to ensure the laser had  
163 warmed up fully and to reduce laser power fluctuations which would affect the drift of the  
164 signal.

165

166 Reagents:

167 Luria broth (LB) and antibiotics (ampicillin, kanamycin, trimethoprim) were all supplied by  
168 Sigma-Aldrich (USA).

169

170 Bacterial Strains:

171 *E. coli* BL21(DE3)pLysS competent cells (Promega, UK) were selected for their suitability  
172 for transformation with a plasmid containing ampicillin resistance (pRSET/EmGFP plasmid;  
173 Invitrogen, UK).

174

175 A clinical isolate of *E. coli* was obtained from the microbiology repository of Great Ormond  
176 Street Hospital (London, UK).

177

178 Bacterial preparation:

179 An LB media (Sigma-Aldrich) plate was streaked with BL21 *E. coli* (Promega) or clinical  
180 isolate *E. coli* (obtained from Great Ormond Street Hospital) from frozen stocks in a sterile  
181 hood. These were grown up overnight at 37°C. A single colony was used to inoculate 4 mL  
182 LB media, which was incubated at 37°C for 2 hours (225 r.p.m. shaking), to obtain mid-log  
183 phase growth. The OD<sub>600</sub> of the culture was measured using a Nanodrop One-C (Thermo  
184 Scientific), and a final OD<sub>600</sub> for bacterial inoculation for experimental measurement was  
185 adjusted to keep as constant as possible.

186

187 Bacterial transformation with ampicillin resistance:

188 An aliquot of competent bacterial stock was thawed on ice for 20-30 minutes. 1-5 µL (10pg-  
189 100ng) pRSET-EmGFP plasmid (Invitrogen, CA, USA) was mixed with 25 µL thawed  
190 bacterial solution and incubated for 5-10 minutes on ice, followed by heat shock treatment at  
191 42°C for 40 seconds and returned to ice for two further minutes. 500 µL warmed SOC media  
192 was added, and this was incubated at 37°C at 225 r.p.m. for one hour. 50 µL was plated onto  
193 an agar plate which contained 50 µg/mL nafcillin/ampicillin mixture. This plate was  
194 incubated overnight at 37°C and colonies used were made into frozen stocks for experimental  
195 use.

196

197 Data analysis:

198 Vertical deflection data (nm) was recorded on JPK Nanowizard 3 software at 20 kHz  
199 sampling frequency. This raw data (SI Figure 5a) was then processed in 800 second “chunks”

200 using analysis code written in Matlab. This code applies a Savitzky-Golay finite impulse  
201 response (FIR) smoothing filter of polynomial order 2 to the data, with a filtering frequency  
202 of 101 Hz (SI Figure 5b). A Savitzky-Golay smoothing filter was chosen as this function can  
203 filter noisy data effectively without removing high frequency data.

204

205 To identify the number of bacterial crossings, both local maxima and minima were identified,  
206 as bacteria moving through the laser was observed to cause both peaks and dips in the signal  
207 (SI Figure 5c, peaks labelled with blue triangles). A “Peak Finder” function was used to  
208 identify local minima/maxima in the signal, where a “peak” was defined as having a  
209 threshold drop of at least 0.5 nm on each side. This was to ensure that only the larger peaks  
210 were counted, which correspond to bacteria moving across the laser. Smaller “noise” seen in  
211 the signal was not attributed to actual bacterial crossings, but could be due to partial  
212 crossings, or a change of orientation of bacteria within the laser during a crossing. This  
213 threshold peak prominence value of 0.5 nm was applied empirically across all files when  
214 carrying out the analysis to remove any bias of identifying peaks in the signal.

215

216 Across the experiment, the number of peaks was calculated for a subsampled time frame to  
217 increase the resolution of the data from 800 seconds to 267 seconds, and plotted across the  
218 experimental conditions of LB media, addition of bacteria, addition of antibiotic (SI Figure  
219 5d).

220

221 To calculate the antibiotic sensitivity ( $r_{\text{sensitivity}}$ ) the ratio of the signal pre-antibiotic addition,  
222  $S_{\text{baseline}}$ , and 45 minutes post-antibiotic addition,  $S_{\text{antibiotic}}$  (SI Figure 5d).  $r_{\text{sensitivity}}$  provides a  
223 binary readout of sensitivity,  $r_{\text{sensitivity}} \leq 1$  indicates cell death or inhibition of bacterial  
224 growth, and sensitivity to the antibiotic in solution;  $r_{\text{sensitivity}} > 1$  indicates bacterial growth,  
225 and therefore resistance to the antibiotic used.

226

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236

237 **Author Contributions**

238 I.B., A.L.B.P. and R.M.K. designed the study. I.B. performed the optical interference  
239 experiments. I.B. and A.L.B.P. analysed the data. I.B. and A.L.B.P. wrote the paper. All  
240 authors discussed the results and commented on the manuscript.

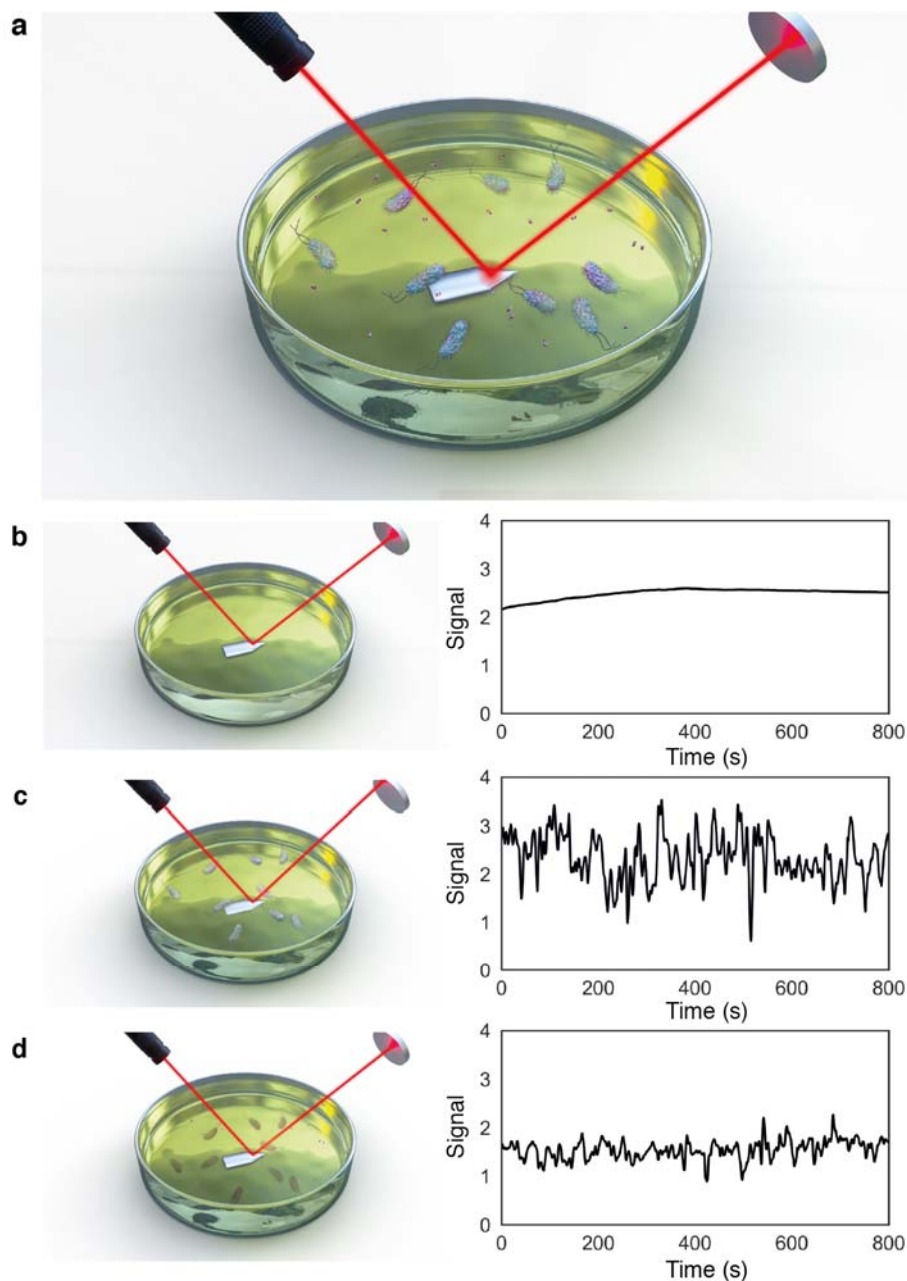
241

242 **Competing Interests**

243 The authors declare no competing financial interests.

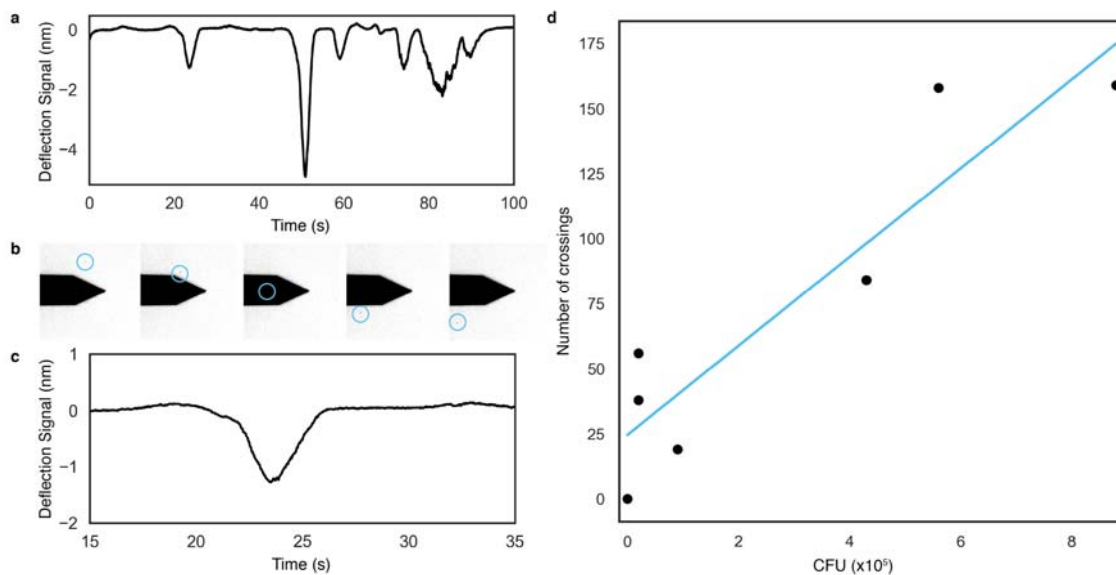


244 **Figures**



245

246 **Figure 1. Principles of nanoscale optical interference method.** a, Illustration of bacterial  
247 cells inoculated in growth media with antibiotic molecules, with laser reflecting off cantilever  
248 surface onto photodiode detector. Bacteria in solution move into the laser beam, which  
249 interfere and cause the laser to move on the detector. This results in peaks in the measured  
250 signal. Photodiode signal measured in media solution without bacterial inoculant (b), with  
251 bacteria in solution (c) and 45 mins after addition of antibiotic (d). Signal decreases after  
252 addition of antibiotic for sensitive strains.



253

254 **Figure 2. Signal caused by single bacteria decreases after 45 minutes from antibiotic**  
255 **addition. a,** At low bacterial inoculant concentration, individual peaks can be identified  
256 within the signal. Combined optical tracking and signal measurement shows movement of  
257 single bacterium (blue circle) through laser path (**b,** optical images) as a single peak in the  
258 signal (**c**). **d,** Effect of bacterial concentration (CFU,  $\times 10^5$ ) on number of bacterial  
259 crossings.

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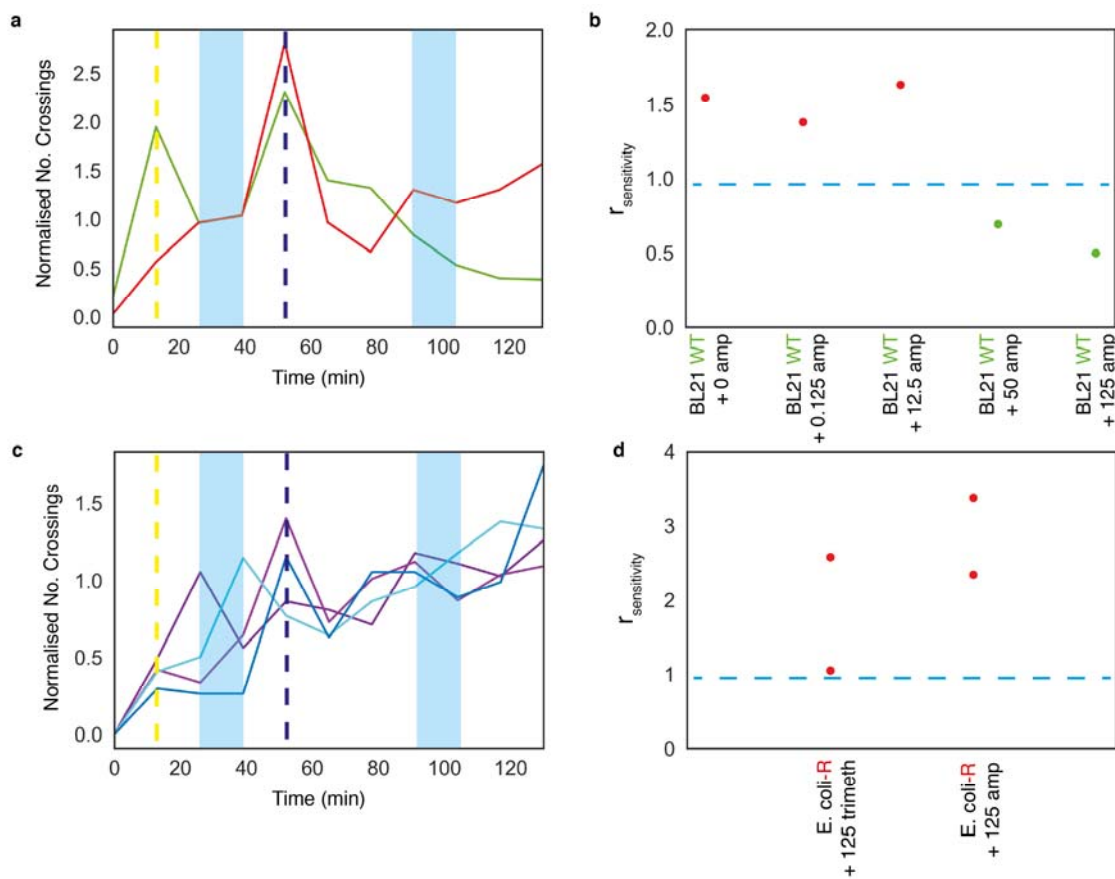
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272 **Figure 3. Systematic analysis of susceptibility in clinical and laboratory strains of *E.***

273 ***coli.* a,** Susceptibility of BL21-WT (S, green) and BL21-ampR *E. coli* (R, red) to 125  $\mu\text{g}/\text{mL}$

274 ampicillin. Addition of bacteria (yellow dotted line) and antibiotic solution (dark blue dotted

275 line) to the system cause large fluctuations in the signal as the liquid is mixed, which

276 dissipate within  $\sim 800$  seconds. Number of bacterial crossings in a given time period, here 800

277 seconds, is plotted. The number of bacterial crossings shows a decrease 45 minutes after

278 antibiotic addition. **b,** Determination of resistance profile, with sensitivity readout ( $r_{\text{sensitivity}}$ ).

279  $r_{\text{sensitivity}}$  was calculated using the ratio of crossings post-antibiotic and pre-antibiotic at set

280 time points marked in blue in **a**. Strains were determined to be sensitive (S) if  $r_{\text{sensitivity}} < 1$

281 (green); or resistant (R) if  $r_{\text{sensitivity}} \geq 1$  (red), cut off ( $r_{\text{sensitivity}} = 1$ ) shown as blue dashed line.

282 Shown for five concentrations of ampicillin and BL21 *E. coli* **c,** Susceptibility of a clinical

283 isolate of *E. coli*, determined to be resistant to both ampicillin (purple lines) and trimethoprim

284 (blue lines). **d,** Determination of resistance profile.  $r_{\text{sensitivity}}$  for repeats of clinical isolate with

285 125  $\mu\text{g}/\text{mL}$  trimethoprim and ampicillin. Antibiotic concentrations are given in  $\mu\text{g}/\text{mL}$ .

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