1	High-content imaging to phenotype antimicrobial effects on individual bacteria at scale
2	
3	Sushmita Sridhar ^{1,2¶} , Sally Forrest ^{1¶} , Benjamin Warne ¹ , Mailis Maes ¹ , Stephen Baker ¹ ,
4	Gordon Dougan ^{1*} , Josefin Bartholdson Scott ^{1*}
5	
6	¹ Cambridge Institute of Therapeutic Immunology & Infectious Disease, University of
7	Cambridge Department of Medicine, Jeffrey Cheah Biomedical Centre, Puddicombe Way,
8	Cambridge Biomedical Campus, Cambridge CB2 0AW, United Kingdom
9	² Wellcome Sanger Institute, Hinxton, CB10 1SA, United Kingdom
10	
11	[¶] These authors contributed equally
12	
13	*Corresponding authors: e-mail: jb2143@cam.ac.uk & gd312@medschl.cam.ac.uk
14	
15	
16	
17	
18	
19	
20	
21	
22	
23	
24	
25	

26 Abstract

27 High-content imaging (HCI) is technique for screening multiple cells in high resolution to 28 detect subtle morphological and phenotypic variation. The method has been commonly 29 deployed on model eukaryotic cellular systems, often for screening new drugs and targets. 30 HCI is not commonly utilised for studying bacterial populations but may be powerful tool in 31 understanding and combatting antimicrobial resistance. Consequently, we developed a high-32 throughput method for phenotyping bacteria under antimicrobial exposure at the scale of 33 individual bacterial cells. Imaging conditions were optimised on an Opera Phenix confocal 34 microscope (Perkin Elmer) and novel analysis pipelines were established for both Gram-35 negative bacilli and Gram-positive cocci. The potential of this approach was illustrated using 36 isolates of Klebsiella pneumoniae, Salmonella enterica serovar Typhimurium, and 37 Staphylococcus aureus. HCI enabled the detection and assessment of subtle morphological 38 characteristics, undetectable through conventional phenotypical methods, that could 39 reproducibly distinguish between bacteria exposed to different classes of antimicrobials with 40 distinct modes of action (MOA). In addition, distinctive responses were observed between 41 susceptible and resistant isolates. By phenotyping single bacterial cells, we observed intra-42 population differences, which may be critical in identifying persistence or emerging 43 resistance during antimicrobial treatment. The work presented here outlines a comprehensive 44 method for investigating morphological changes at scale in bacterial populations under 45 specific perturbation.

46

47 Importance

High-content imaging (HCI) is a microscopy technique that permits the screening of multiple
cells simultaneously in high resolution to detect subtle morphological and phenotypic
variation. The power of this methodology is that is can generate large datasets comprised of

51	multiple parameters taken from individual cells subjected to range of different conditions. We
52	aimed to develop novel methods for using HCI to study bacterial cells exposed to a range of
53	different antibiotic classes. Using an Opera Phenix confocal microscope (Perkin Elmer) and
54	novel analysis pipelines we created a method to study the morphological characteristics of
55	Klebsiella pneumoniae, Salmonella enterica serovar Typhimurium, and Staphylococcus
56	aureus when exposed to antibacterial drugs with differing modes of action. By imaging
57	individual bacterial cells at high resolution and scale, we observed intra-population
58	differences associated with different antibiotics. The outlined methods are highly relevant for
59	how we begin to better understand and combat antimicrobial resistance.
60	
61	Keywords
62	High-content imaging, image analysis, bacteria, antimicrobial resistance, phenotyping
63	
64	
65	
66	
67	
68	
69	
70	
71	
72	
73	
74	
75	

76 Introduction

77 Antimicrobial resistance (AMR) is one of the greatest current challenges in human health, 78 with rising cases of antimicrobial resistant bacterial infections and a lack of new classes of 79 licensed antimicrobials. Advances in bacterial genomics have revolutionised our ability to 80 genotype antimicrobial resistant bacterial isolates at scale. However, it remains critical to link 81 genotype with phenotype in order to interpret the biological and clinical relevance of AMR. 82 Some phenotyping methods have been adapted to work at scale (e.g. antimicrobial 83 susceptibility testing using semi-automated platforms such as the bioMérieux VITEK 84 system), yet many others either rely on low throughput methods or aggregate data from 85 mixed populations of bacterial cells. The analysis of bulk bacterial populations rather than 86 individual cells potentially overlooks persister cells or the emergence of resistant or tolerant 87 bacteria within that population. High throughput imaging of bacterial populations at the scale 88 of individual cells has received limited attention but may be achieved by exploiting high-89 content microscopy.

90

91 High-content imaging (HCI) can be utilised as a powerful phenotypic screening approach that 92 combines automated microscopy with image analysis to quantify multiple morphological 93 features. This approach may capture subtle differences in structure and shape not discernible 94 by the human eye or conventional phenotypic methods. Such image-based profiling has great 95 potential in high-throughput drug screening, which has mainly been applied to eukaryotic 96 cells and tissue(1, 2). In the field of microbiology, HCI has predominantly been used to study 97 intracellular pathogens such as Mycobacterium tuberculosis(3-6) and Salmonella species as 98 they interact with host cells(7), but only recently to screen individual bacteria growing as a 99 population in batch culture(8, 9). Pogliano and colleagues developed a bacterial cytological 100 profiling assay to identify morphological changes in Escherichia coli and other species in

response to different classes of antimicrobials using fluorescence microscopy(10–13).
Analysis of image data enabled the assignment of distinct morphological profiles correlating
with the mechanism of action of the antimicrobial compounds tested(10). This method
opened up a novel way of screening new therapeutic compounds simultaneously for efficacy
and mode-of-action (MOA) using bacterial imaging(8–13).

107 Given the variety of AMR mechanisms harboured by bacterial species and, in many cases, by 108 isolates of the same species, it is important to optimise HCI approaches for a range of 109 bacteria. In this study, we developed and optimised a high-throughput imaging method based 110 on HCI to systematically screen individual bacteria from three different species grown under 111 antimicrobial exposure. We optimised bacterial imaging conditions using an Opera Phenix 112 confocal microscope (Perkin Elmer) and established novel analysis pipelines for image 113 segmentation and bacterial morphological analysis for both Gram-negative bacilli and Gram-114 positive cocci. The combination of HCI and image analysis enabled the detection of subtle 115 morphological characteristics that differed between different antimicrobial classes. This work 116 contributes to the expansion of microbial phenotyping from population-level to single-cell 117 analysis and provides a comprehensive method of bacterial phenotypic screening at scale. 118

119 Materials and Methods

120 Bacterial isolates

121 A number of reference bacterial isolates, representing clinically important species, were

analysed. This panel included one Gram-positive (Staphylococcus aureus) and two Gram-

- 123 negative (Salmonella enterica serovar Typhimurium and Klebsiella pneumoniae) species.
- 124 Two isolates were included per species, each with broadly different antimicrobial
- 125 susceptibility profiles (Table 1).

126

127 Antimicrobial susceptibility testing

128 Antimicrobial susceptibility testing was performed for a range of clinically relevant 129 antimicrobials with different MOAs (Table 1). Minimum Inhibitory Concentrations (MICs) 130 were determined by ETESTs (bioMérieux) according to the manufacturer's instructions. 131 Briefly, pure bacterial cultures were diluted in saline to 0.5 MacFarland standard, 100 µl of 132 solution was inoculated and spread onto Isosensitest plates (Oxoid, CM0471), and an ETEST 133 strip was placed on top. Plates were incubated for 16-18 hours at 37°C before the result was 134 read. 135 136 Preparation of plate coatings 137 Coating matrices were prepared according to manufacturer recommendations in sterile 138 conditions (Table S1). All coatings, except poly-L-lysine, were incubated in ultra-thin 96 139 well plates (Perkin Elmer CellCarrier Ultra, 6655308) overnight at 37°C. The following day, 140 wells were rinsed 1-3 times with wash buffer (Table S1). For poly-L-lysine, wells were 141 coated for 5 minutes. The solution was aspirated, and wells were left to dry overnight at

142 143

144 Bacterial imaging assay

37°C.

145 Overnight bacterial cultures were diluted in LB broth and mixed with antimicrobials to a final

146 antimicrobial concentration of 5x MIC. Where an MIC could not be measured (i.e., where

147 bacterial growth continued along the whole length of the ETEST), the upper limit of the

- 148 ETEST was arbitrarily used in place of the MIC. The bacteria were incubated with and
- 149 without antimicrobials in static incubators in ultra-thin 96 well plates for 2 hours at 37°C.
- 150 The plates were aspirated, and the remaining adherent bacteria were fixed with 4%

151	paraformaldehyde (Alfa Aesar, J61899.AK) for 10 minutes. The wells were washed once
152	with 50µl of DPBS (Thermo, 10010023) before staining. Fixed cells were stained with FM4
153	64 (2µg/ml, Thermo, T13320), SYTOX green (0.25µM, Thermo, S7020) and 4',6-
154	Diamidino-2-phenylindole dihydrochloride (DAPI, 2µg/ml, Sigma, D9542). Staining was
155	performed at ambient temperature for 20 minutes in the dark followed by a wash with $50\mu l$
156	PBS. Finally, 50μ l of PBS was added to wells and the plates were imaged within 24 hours.
157	
158	High-content imaging and image analysis
159	High-content confocal imaging was performed using an Opera Phenix (Perkin Elmer), using
160	a 63x water immersion lens. 10 fields of view (equating to 0.4 mm ²) were imaged for each
161	well, with 3 z-stacks per field at 0.5 μ m intervals to ensure comprehensive imaging of the
162	bacterial monolayer. Triplicate biological and technical replicates were performed for all
163	experiments. Image analysis was performed using Harmony (v4.9). Optical correction was
164	performed using flatfield and brightfield correction. The detailed full analysis pipelines are
165	shown in Table S1 and S2. Data were exported and plotted in GraphPad Prism and R(14).
166	

167 **Results**

168 High-content imaging and analysis of individual bacteria

169 An HCI workflow was established using two reference isolates from each of the three

170 bacterial pathogens S. Typhimurium, K. pneumoniae, and S. aureus (Figure 1A; Table 1).

171 Organisms were selected to have contrasting AMR profiles within each species. Each of the

172 isolates was exposed to different antimicrobial agents, and HCI was used to collect

173 phenotypic data for numerous individual bacteria within each assay. To this end, overnight

bacterial cultures were grown in 96-well microtiter plates in the presence or absence of each

antimicrobial for two hours to capture multiple early morphological changes. The

antimicrobials used are listed in Table 1. To capture images, the bacteria in each well were
stained *in situ* with markers for the cell membrane (FM4-64), nucleic acid (DAPI), and
membrane permeability (SYTOX green)(10) (Figure 1B). Imaging was performed on an
Opera Phenix and image analysis was conducted using Harmony software.
As both rod- and cocci-shaped bacteria were imaged at the single cell scale, it was necessary

182 to build separate, parallel pipelines for accurate analysis of microorganisms with different 183 morphology. Examples of image segmentation and analysis of Gram-negative rods and 184 Gram-positive cocci are shown in Figure 2 and detailed in Table S2 and Table S3, 185 respectively. Images of Gram-negative rods were initially filtered using FM4-64 intensity 186 patterns to enhance stained objects and subtract the background (Figure 2A). All images were 187 then calculated based on DAPI and FM4-64 intensity and resized to include both cytosolic 188 (DNA) and membrane regions; filtering was performed to remove artefacts such as 189 incomplete bacterial bodies (Figure 2A and C). Morphological features and stain intensities 190 were calculated for each defined bacterial cell, including area, roundness, width and length, 191 as well as various measures of the intensity, symmetry and distribution of each of the FM4-192 64, DAPI and SYTOX green channels within each object (Figure 2B and D). For Gram-193 negative rods, the segmented objects were further classified as either single bacterial cells, 194 dividing cells, or artefacts (Figure 2B) with a manually trained linear classifier using 195 Harmony PhenoLOGIC. Subsequent analyses were conducted on the single cells only. The 196 number of individual bacteria captured and analysed per well was dependent on isolate and 197 treatment but was a minimum of 2,000 bacteria per untreated well for all replicates.

198

199 *Optimising bacterial imaging*

200 The bacterial isolates displayed different degrees of adhesion to the base of the 96-well 201 plates, which significantly affected the image quality and downstream analysis. For example, 202 the non-motile K. pneumoniae isolates spontaneously strongly adhered to the bottom of the 203 wells, whereas the motile S. Typhimurium isolates displayed relatively poor adhesion 204 resulting in blurry superimposed images of the flattened z-stack (maximum projection) 205 (Figure S1A and B). Consequently, it was necessary to assess multiple plate coating 206 conditions for each S. Typhimurium isolate to identify the optimal conditions for binding and 207 image clarity. Ultimately, the image segmentation pipelines were used to quantify individual 208 bacteria on 11 commercially available coating matrices (thick and thin rat tail collagen, 209 Matrigel, vitronectin, fibronectin, Cell-Tak, laminin, wheat germ agglutinin (WGA), poly-L-210 lysine, gelatine, and mouse collagen) in comparison to non-coated wells (Table S1). 211 212 The optimal coating conditions were found to differ between species, and to a lesser extent 213 for each isolate (Figure 3). The K. pneumoniae isolates displayed the best adhesion and image 214 quality on non-coated, WGA, and fibronectin-coated plates (Figure 3A, Figure S2A). 215 Therefore, all subsequent experiments with these isolates were conducted on non-coated 216 wells. In contrast, image quality and the number of adherent S. Typhimurium bacteria 217 improved dramatically upon optimization of the plate coating (Figure S1C). Both S. 218 Typhimurium isolates displayed very poor adhesion to non-coated wells (Figure 3B; Figure 219 S2B), but adhered sufficiently, although to different extents, to wells coated with thick rat tail 220 collagen, Matrigel, and vitronectin (Figure 3B; Figure S2B), with collagen and vitronectin 221 being the optimal conditions for NCTC 13347 and NCTC 13348 adhesion, respectively. 222 While different coatings were chosen for the two isolates, it would have been feasible to use 223 the same coating, as the number and image quality of adhered organisms on rat tail collagen, 224 Matrigel, and vitronectin were sufficient for analysis. To overcome any residual lack of

adhesion of *S*. Typhimurium, image analysis of these isolates was performed on individual
planes rather than a maximum projection of three z-stacks.

227

228 The two *S. aureus* isolates displayed different adhesion properties, with ATCC 29213

showing increased adhesion on vitronectin, fibronectin, Cell-Tak, WGA, and gelatin-coated

230 wells, whereas NCTC 6571 only had sufficient cell counts on thin collagen and fibronectin

coated wells (Figure 3C). Taking adhesion and image clarity (Figure S2C) into account, thin

collagen and vitronectin were used for optimal adhesion of NCTC 6571 and NCTC 29213,

233 respectively.

234

235 *Measuring distinct morphological changes in response to antimicrobial compounds*

236 To measure the phenotypic effects of antimicrobials with distinct mechanisms of action

237 (MOAs), bacteria were incubated with 11 commercially available antimicrobials for 2 hours

and imaged as described above. Antimicrobials were used at 5x the MIC determined by

239 ETEST, or 5x the highest concentration tested if an isolate had an MIC higher than the

ETEST range (Table 1). Figure 4A, 5A, and 6A show examples of the observed

241 morphological changes at 2h post-treatment for a non-motile (*K. pneumoniae* NCTC 43816)

and a motile Gram-negative rod (S. Typhimurium NCTC 13348), and a Gram-positive coccus

243 (*S. aureus* ATCC 29213), respectively.

244

There were notable differences that the pipeline was able to capture between the effects of the same antimicrobials on Gram-negative versus Gram-positive bacteria, with more visually striking morphological changes observed in the Gram-negative bacteria. The established image analysis pipelines produced mean and standard deviation measurements for >90 morphological features and stain intensities for each bacterium imaged (Table S4-6). These

250	measurements were combined for each isolate and analysed using principal component
251	analysis (PCA). Technical replicates of each class of antimicrobial separated into distinct
252	clusters based on MOA (Figure 4B, 5B and 6B, Figure S3), and biological replicates
253	produced similar distribution by PCA, demonstrating assay reproducibility (Figure S4).
254	Although separate pipelines were required for Gram-positive and Gram-negative organisms,
255	each pipeline was able to distinguish a wide variety of phenotypes generated by antimicrobial
256	treatment, segmenting the images and identifying individual bacteria despite the
257	morphological changes associated with each antimicrobial (Figure S5).
258	
259	Antimicrobials acting on similar cellular processes generally induced comparable
260	morphological changes in each species, and these were found to cluster in a PCA. Bacteria
261	treated with tigecycline and gentamicin, which block protein synthesis by binding the 30S
262	ribosomal subunit, clustered proximally for all Gram-negative isolates tested (Figure 4B, 5B).
263	In addition, these generally also clustered near rifampicin and azithromycin treated bacteria
264	(Fig 4B, 5B); these antimicrobials affect protein synthesis by inhibiting RNA polymerase or
265	translation by binding the 50S ribosomal subunit, respectively. Antimicrobials that inhibit
266	DNA synthesis (trimethoprim/sulphamethoxazole), DNA replication (ciprofloxacin) and cell
267	wall synthesis (ampicillin, cefuroxime and meropenem) tended to induce an elongated
268	phenotype and again clustered proximally (Fig 4B, 5B). Notably, meropenem clustered
269	separately to the other β -lactams for the <i>Klebsiella</i> isolates and appeared to disrupt the
270	bacterial cell wall more potently, causing the bacteria to swell and lyse instead of elongating
271	(Figure 4, S3A). K. pneumonia NCTC 13438 was resistant to ampicillin,
272	trimethoprim/sulphamethoxazole, cefuroxime and ciprofloxacin at concentrations higher than
273	the ETEST scale, and with the exception of ciprofloxacin, these clustered with the untreated
274	control (Figure S3A). A similar phenotype was observed for S. Typhimurium NCTC 13348

treated with ampicillin (Figure 5B). This highlights that HCI screens provide novel data
regarding drug susceptibility as well as MOA.

277

278 The morphological changes observed for S. aureus were relatively subtle compared to those 279 for the Gram-negative isolates (Figure 6A). Only ciprofloxacin induced a visually discernible 280 phenotypic change, which was associated with enlarging bacterial area. However, after image 281 analysis, each antimicrobial effectively separated into unique clusters by PCA, except 282 trimethoprim/sulphamethoxazole, which clustered alongside the untreated controls. (Figure 283 6B, S3C, S5). This finding suggests that the analysis could discriminate between very subtle 284 cellular variation by capturing and analysing a large number of phenotypic parameters. 285 286 Measuring the relative importance of specific morphological and fluorescence intensity 287 parameters 288 To assess the quality of the image analysis we calculated the Z prime (Z') values using 289 Harmony, comparing treated and untreated bacteria for each species and antimicrobial 290 combinations (Table S7-9), where an ideal assay should yield values between 0.5 and 1(15). 291 The Z' values were higher for antimicrobials that clustered further from the untreated control 292 in the PCA, and the Gram-negative isolates generally had more Z' values above 0.5 than the 293 Gram-positive isolates. For example, the trimethoprim/sulphamethoxazole-treated S. aureus 294 ATCC 29213 failed to separate from the untreated control by PCA, which correlated with 295 poor Z' values (Table S9). Similarly, poor Z' values were obtained for azithromycin-treated

K. pneumoniae NCTC 43816 (which is intrinsically resistant to macrolides) and gentamicin-

treated S. Typhimurium NCTC 13348 (Table S7-8).

298

To highlight the relative importance of some of the measured parameters with high Z' values,
differences in measurements across representative antimicrobials were compared.
Morphological measurements of roundness, area and length-to-width ratio (Figure 7), as well
as threshold compactness and the radial relative deviation of the DAPI and FM4-64 staining
patterns, were plotted for a selection of antimicrobials with different MOAs (Figure S6). In
addition, SYTOX green intensity was included as this should only stain bacteria if membrane

305 integrity has been compromised (Figure S6 M-O).

306

307 When plotting these parameters individually, clear differences were observed between the 308 different antimicrobials for the K. pneumoniae isolates and, to a lesser extent, the S. 309 Typhimurium isolates. For example, increased area, decreased roundness, increased length-to 310 -width ratio and FM4-64 and DAPI radial relative deviation correlated with the observed 311 elongation phenotype observed for cefuroxime and ciprofloxacin (Figure 7 A, B, D, E, G, H 312 and Figure S6 A, B, D, E). In contrast, increased FM4-64 and DAPI threshold compactness 313 as well as bacterial roundness was observed for rifampicin and tigecycline (Figure 7 D, E, 314 and Figure S6 G, H, J, K).

315

Generally, SYTOX green intensity was higher for antimicrobials disrupting the bacterial
membrane for both Gram-negative (meropenem) and Gram-positive isolates (oxacillin).
However, the effect of individual parameters on *S. aureus* were subtler than for the Gramnegative isolates, with only gentamicin treatment showing slightly decreased roundness and
increased length-to width ratio (Figure 7 F and I), demonstrating the need to observe multiple
combined phenotypic parameters.

322

323 Phenotypes within a bacterial population

324 Using violin plots, it was possible to visualize the population density and distribution of 325 bacteria for any given parameter. This analysis demonstrated the inherent morphological 326 heterogeneity in bacterial populations of the same isolate under the same growth conditions 327 (Figure 7 and S6). For example, certain antimicrobial treatments yielded a high degree of 328 heterogeneity in the length-to-width ratio, notably cefuroxime and ciprofloxacin treated K. 329 pneumoniae NTCT 43816 (Figure 8). In contrast, rifampicin treatment appeared to yield 330 decreased variability within a population as compared to untreated controls (Figure 8). By 331 phenotyping single cells, it is possible to observe within-population differences, which is 332 critical for identifying persistence or emerging resistance during antimicrobial treatment.

333

334 Discussion

In this study we optimised an experimental pipeline for high-throughput confocal imaging of motile and non-motile bacteria in liquid culture. We used this method for systematic screening of Gram-positive and Gram-negative bacteria under antimicrobial pressure, with robust and standardised image analysis pipelines to efficiently and reproducibly measure distinct morphological changes correlating with antimicrobial MOA. This analysis was built around profiling the subtle morphological phenotypes of individual bacteria in a culture, providing information on the whole population and variation within that population.

342

343 There is a number of advantages to using HCI for bacterial research. It provides flexibility in 344 experimental design, with the ability to customise and compare growth conditions and 345 individual isolates from different species in high throughput. Traditional phenotyping 346 methods rely on the collective properties of large numbers of bacteria, HCI enables 347 measurements at the scale of individual bacterial cells. Advances in image analysis permit 348 reliable segmentation of bacterial images and rapid, detailed profiling of individual bacterial

cells with the ability to demonstrate the heterogeneity of bacterial phenotypes in any givenenvironment.

351

352 Our work identified some challenges in using HCI for bacterial research, in particular 353 variation in adhesion to microtiter plates. Poor adhesion influences both image quality and 354 the number of bacteria successfully imaged for downstream analysis. This challenge was 355 overcome by testing a range of coating matrices, which demonstrated substantial inter- and 356 some intra-species variation in their ability to adhere to each substrate. For example, there 357 were notable differences in adhesion between the non-motile K. pneumoniae and motile S. 358 Typhimurium. K. pneumoniae possess an array of adhesins that allow them to adhere and 359 persist in different environments, which have contributed to their emergence as an important 360 nosocomial pathogen(16, 17). In contrast, S. Typhimurium relies on motility and more 361 specific cellular interactions and invasion for causing infection(7). These factors highlight the 362 need to optimise imaging conditions for each bacterial isolate. However, for most isolates, 363 more than one coating condition was sufficient for downstream analysis, making it possible 364 to screen multiple isolates in parallel using the same plate coating for the higher throughput 365 assays.

366

One of the most challenging aspects of image analysis was the segmentation and identification of individual bacteria. This is in part because most existing image analysis software is designed primarily to analyse images of eukaryotic cells. However, analysis pipelines to effectively segment both rod- and cocci-shaped bacteria were created using existing image analysis tools in the Harmony software. Though the analysis pipelines in this study were created using Harmony, which is a proprietary software from Perkin Elmer, there

are open access image analysis software options available - for example CellProfiler(18) and
Cellpose(19)– which have similar analysis capabilities.

375

376 It was necessary to produce separate pipelines for cocci and rod-shaped bacteria for the initial 377 segmentation. Other studies have also utilised different pipelines for phenotypically variant 378 species; for example, the analysis used by Zoffmann and colleagues for E. coli was not 379 suitable for Acinetobacter baumannii, as these species differ in size and shape(8). 380 Importantly, the pipelines created in our study could be used to reproducibly segment bacteria 381 in all growth conditions used, even as morphologies changed due to antimicrobial exposure. 382 Distinct morphological changes were observed in response to different classes of 383 antimicrobials, with different effects observed in Gram-negative versus Gram-positive 384 species. However, bacteria from the same species generally displayed similar morphological 385 distributions by principal component analysis when treated with 5x the MIC, correlating with 386 antimicrobial mechanism. In addition, different clustering was observed between susceptible 387 and resistant isolates, allowing for simultaneous evaluation of potency as well as MOA. 388 389 The phenotypic changes identified in this study in the presence of antimicrobials are 390 comparable to previous imaging studies in Enterobacteriaceae, including bacterial 391 enlargement with carbapenems and cephalosporins(20, 21), compaction of the nucleoid with 392 antimicrobials targeting the bacterial ribosome(22) and filamentous elongation in the 393 presence of fluoroquinolones(23). In agreement with other studies, we identified similar 394 morphological changes in isolates of K. pneumoniae and S. Typhimurium to those previously 395 reported for E. coli in response to a range of antimicrobial classes(10), but here we employed 396 a simplified method by removing centrifugation steps and by imaging directly in wells rather 397 than on agarose pads. This facilitates higher throughput and scalability.

398

399 The fluorescent staining protocol previously optimised by Nonejuie *et al.*(10) worked well 400 across all the isolates tested in this study. FM4-64 stains the cell membrane, and the staining 401 patterns should relate to membrane integrity. DAPI and SYTOX green both stain nucleic 402 acids, but only DAPI is permeable through an intact cell membrane, making SYTOX green 403 intensity an additional measurement of membrane integrity after antimicrobial exposure(24, 404 25). In addition, nucleic acid stains can distinguish between subtle alterations in nucleic acid 405 distribution patterns. Plotting individual phenotypic parameters was sufficient when an 406 antimicrobial induced a strong visual phenotypic effect, for example, length-to-width ratio 407 could be used for ciprofloxacin or cefuroxime treated K. pneumoniae. However, in most 408 cases, and in particular for the smaller cocci-shaped S. aureus isolates where the phenotypic 409 effects were subtler, a combination of morphological as well as stain intensities, distribution 410 and symmetry measurements were required to efficiently evaluate the data. This highlights 411 that the software can detect important variations that are not obvious in conventional 412 phenotypic methods.

413

414 Our methods contribute to moving microbial phenotyping from a population-based analysis 415 to the scale of individual bacterial cells and provides a comprehensive method of bacterial 416 phenotypic screening at scale. This approach has a wide range of applications, but the ability 417 to provide analysis of diverse collections of isolates simultaneously in a range of growth 418 conditions gives it important potential in the fight against AMR. In addition to existing roles 419 in compound screening for antimicrobial efficacy and simultaneous MOA prediction(26), the 420 technology could be used for more detailed mechanistic follow up studies using mutant 421 libraries to assess genes that are protective against individual drugs(27). Large numbers of 422 compounds and bacterial isolates, representing species with diverse genetic backgrounds, can

423 be screened at scale. We have previously shown the utility of bacterial HCI for therapeutic 424 antibody screening(9), and there is potential to assess synergy between antimicrobials and 425 monoclonal antibodies against multi-drug resistant bacteria that would be challenging using 426 other platforms. Importantly, by analysing individual bacteria within a culture, it is possible 427 to detect differential effects and persister cells during drug treatment and be able to truly 428 evaluate the efficacy of a compound. 429 430 Acknowledgments 431 We thank James Hutt and Achim Kirsch for their help with the analysis pipelines. This work 432 was supported by a Innovate UK Commercial in Confidence grant to purchase the Opera 433 Phenix. SS and SB are funded by the Wellcome Trust (206194 and 215515/Z/19/Z). SF, BW, 434 MM, SB, GD and SJB are supported by funding from the National Institute for Health 435 Research [Cambridge Biomedical Research Centre at the Cambridge University Hospitals 436 NHS Foundation Trust] and National Institute for Health Research AMR Research Capital 437 Funding Scheme [NIHR200640]. The views expressed are those of the authors and not

438 necessarily those of the NHS, the NIHR or the Department of Health and Social Care.

439

440 Data Availability

441 All data underlying the results are available in the supplementary files associated with the442 article.

443

444 **References**

van Vliet E, Daneshian M, Beilmann M, Davies A, Fava E, Fleck R, Julé Y, Kansy M,
 Kustermann S, Macko P, Mundy WR, Roth A, Shah I, Uteng M, van de Water B,
 Hartung T, Leist M. 2014. Current approaches and future role of high content imaging

448		in safety sciences and drug discovery. ALTEX 31:479-493.
449	2.	Bray M-A, Singh S, Han H, Davis CT, Borgeson B, Hartland C, Kost-Alimova M,
450		Gustafsdottir SM, Gibson CC, Carpenter AE. 2016. Cell Painting, a high-content
451		image-based assay for morphological profiling using multiplexed fluorescent dyes. Nat
452		Protoc 11:1757–1774.
453	3.	Christophe T, Ewann F, Jeon HK, Cechetto J, Brodin P. 2010. High-content imaging
454		of Mycobacterium tuberculosis-infected macrophages: an in vitro model for
455		tuberculosis drug discovery. Future Med Chem 2:1283–1293.
456	4.	Barczak AK, Avraham R, Singh S, Luo SS, Zhang WR, Bray MA, Hinman AE,
457		Thompson M, Nietupski RM, Golas A, Montgomery P, Fitzgerald M, Smith RS,
458		White DW, Tischler AD, Carpenter AE, Hung DT. 2017. Systematic, multiparametric
459		analysis of Mycobacterium tuberculosis intracellular infection offers insight into
460		coordinated virulence. PLoS Pathog 13:1-27.
461	5.	Manning AJ, Ovechkina Y, McGillivray A, Flint L, Roberts DM, Parish T. 2017. A
462		high content microscopy assay to determine drug activity against intracellular
463		Mycobacterium tuberculosis. Methods 127:3-11.
464	6.	Greenwood DJ, Dos Santos MS, Huang S, Russell MRG, Collinson LM, MacRae JI,
465		West A, Jiang H, Gutierrez MG. 2019. Subcellular antibiotic visualization reveals a
466		dynamic drug reservoir in infected macrophages. Science (80-) 364:1279–1282.
467	7.	Antoniou AN, Powis SJ, Kriston-Vizi J. 2019. High-content screening image dataset
468		and quantitative image analysis of Salmonella infected human cells. BMC Res Notes
469		12:1–4.
470	8.	Zoffmann S, Vercruysse M, Benmansour F, Maunz A, Wolf L, Blum Marti R, Heckel
471		T, Ding H, Truong HH, Prummer M, Schmucki R, Mason CS, Bradley K, Jacob AI,
472		Lerner C, Araujo del Rosario A, Burcin M, Amrein KE, Prunotto M. 2019. Machine

473		learning-powered antibiotics phenotypic drug discovery. Sci Rep 9:1-14.
474	9.	Maes M, Dyson ZA, Smith SE, Goulding DA, Ludden C, Baker S, Kellam P, Reece
475		ST, Dougan G, Scott JB. 2020. A novel therapeutic antibody screening method using
476		bacterial high-content imaging reveals functional antibody binding phenotypes of
477		Escherichia coli ST131. bioRxiv 2020.05.22.110148.
478	10.	Nonejuie P, Burkart M, Pogliano K, Pogliano J. 2013. Bacterial cytological profiling
479		rapidly identifies the cellular pathways targeted by antibacterial molecules. Proc Natl
480		Acad Sci 110:16169–16174.
481	11.	Quach DT, Sakoulas G, Nizet V, Pogliano J, Pogliano K. 2016. Bacterial Cytological
482		Profiling (BCP) as a Rapid and Accurate Antimicrobial Susceptibility Testing Method
483		for Staphylococcus aureus. EBioMedicine 4:95–103.
484	12.	Lamsa A, Lopez-Garrido J, Quach D, Riley EP, Pogliano J, Pogliano K. 2016. Rapid
485		Inhibition Profiling in Bacillus subtilis to Identify the Mechanism of Action of New
486		Antimicrobials. ACS Chem Biol 11:2222–2231.
487	13.	Htoo HH, Brumage L, Chaikeeratisak V, Tsunemoto H, Sugie J, Tribuddharat C,
488		Pogliano J, Nonejuie P. 2019. Bacterial Cytological Profiling as a Tool To Study
489		Mechanisms of Action of Antibiotics That Are Active against Acinetobacter
490		baumannii. Antimicrob Agents Chemother 63:1–11.
491	14.	Team RC. 2014. R: A language and environment for statistical computing. R Core
492		Team (2014) R A Lang Environ Stat Comput R Found Stat Comput Vienna,
493		Austria(available from http//wwwR-project.org/).
494	15.	Zhang J-H, Chung TDY, Oldenburg KR. 1999. A Simple Statistical Parameter for Use
495		in Evaluation and Validation of High Throughput Screening Assays. J Biomol Screen
496		4:67–73.
497	16.	Di Martino P, Cafferini N, Joly B, Darfeuille-Michaud A. 2003. Klebsiella

498		pneumoniae type 3 pili facilitate adherence and biofilm formation on abiotic surfaces.
499		Res Microbiol 154:9–16.
500	17.	Hassan MZ, Sturm-Ramirez K, Rahman MZ, Hossain K, Aleem MA, Bhuiyan MU,
501		Islam MM, Rahman M, Gurley ES. 2019. Contamination of hospital surfaces with
502		respiratory pathogens in Bangladesh. PLoS One 14:e0224065.
503	18.	Carpenter AE, Jones TR, Lamprecht MR, Clarke C, Kang IH, Friman O, Guertin DA,
504		Chang JH, Lindquist RA, Moffat J, Golland P, Sabatini DM. 2006. CellProfiler: image
505		analysis software for identifying and quantifying cell phenotypes. Genome Biol
506		7:R100.
507	19.	Stringer C, Wang T, Michaelos M, Pachitariu M. 2020. Cellpose: a generalist
508		algorithm for cellular segmentation. Nat Methods https://doi.org/10.1038/s41592-020-
509		01018-x.
510	20.	Van Laar TA, Chen T, You T, Leung KP. 2015. Sublethal concentrations of
511		carbapenems alter cell morphology and genomic expression of Klebsiella pneumoniae
512		biofilms. Antimicrob Agents Chemother 59:1707–1717.
513	21.	Horii T, Kobayashi M, Sato K, Ichiyama S, Ohta M. 1998. An in-vitro study of
514		carbapenem-induced morphological changes and endotoxin release in clinical isolates
515		of gram-negative bacilli. J Antimicrob Chemother 41:435–442.
516	22.	Zimmerman SB. 2006. Shape and compaction of Escherichia coli nucleoids. J Struct
517		Biol 156:255–261.
518	23.	Zahller J, Stewart PS. 2002. Transmission electron microscopic study of antibiotic
519		action on Klebsiella pneumoniae biofilm. Antimicrob Agents Chemother 46:2679-
520		2683.
521	24.	Roth BL, Poot M, Yue ST, Millard PJ. 1997. Bacterial viability and antibiotic
522		susceptibility testing with SYTOX green nucleic acid stain. Appl Environ Microbiol

523 63:2421–2431.

524	25.	McKenzie K, Maclean M, Grant MH, Ramakrishnan P, MacGregor SJ, Anderson JG.
525		2016. The effects of 405 nm light on bacterial membrane integrity determined by salt
526		and bile tolerance assays, leakage of UV-absorbing material and SYTOX green
527		labelling. Microbiol (United Kingdom) 162:1680–1688.
528	26.	Ang MLT, Pethe K. 2016. Contribution of high-content imaging technologies to the
529		development of anti-infective drugs. Cytom Part A 89:755-760.
530	27.	Zahir T, Camacho R, Vitale R, Ruckebusch C, Hofkens J, Fauvart M, Michiels J.
531		2019. High-throughput time-resolved morphology screening in bacteria reveals
532		phenotypic responses to antibiotics. Commun Biol 2:269.

533

534 Figures and Tables

535

536 **Table 1: Minimum Inhibitory Concentrations.**

Species	ID	Ampicillin	Azithromycin	Trimethoprim- sulfamethoxazole	Ciprofloxacin	Gentamicin	Rifampicin	Meropenem	Tigecycline	Cefuroxime	Oxacillin	Vancomycin
K. pneumoniae	NCTC 13438	>256*	32	>32*	>32*	6	16	0.125	0.094	>256*	ND	ND
K. pneumoniae	ATCC 43816	64	2	0.5	0.16	1.5	>32*	0.125	0.5	1.5	ND	ND
S. Typhimurium	NCTC 13347	0.5	2	0.25	0.012	2	16	0.125	0.19	3	ND	ND
S. Typhimurium	NCTC 13348	>256*	3	0.25	0.012	3	24	0.064	0.19	3	ND	ND
S. aureus	NCTC 6571	ND	ND	0.125	0.094	2	ND	ND	ND	ND	0.19	1
S. aureus	ATCC 29213	ND	ND	0.19	0.25	2	ND	ND	ND	ND	0.38	1.5

537

538 MICs were determined by ETEST and are presented in µg/ml. * indicates the MIC is above the highest antimicrobial concentration on the

539 ETEST; ND indicates that MIC was not determined.

540

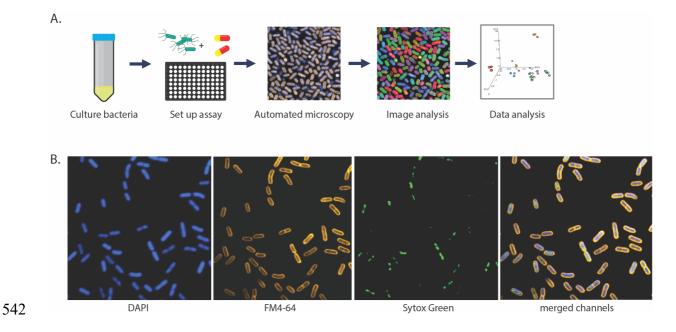
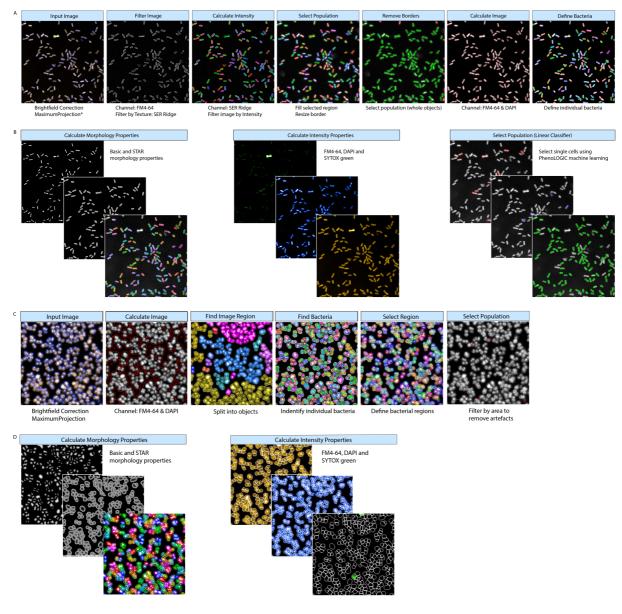




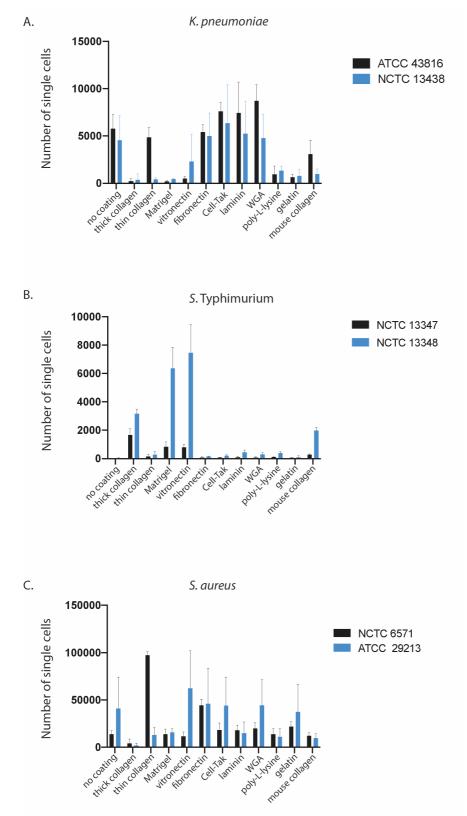
Figure 1: Bacterial high-content imaging. (A) Schematic of the bacterial high-content imaging workflow. Overnight bacterial cultures are added to ultra-thin bottom plates, and incubated, with or without antimicrobial compounds. Adherent bacteria are fixed and stained before being imaged on an Opera Phenix high-content confocal microscope using a 63x water immersion objective. Images were analysed using Harmony software and data was exported and plotted in R. (B) Representative image of K. pneumoniae NCTC 43816 stained with FM4-64 (cell membrane), DAPI (nucleic acid, membrane permeable) and SYTOX green (nucleic acid, membrane impermeable).



559

560 Figure 2: Harmony bacterial image analysis workflow for Gram-negative rods (A-B) and Gram-positive cocci (C-D). (A) Using basic brightfield correction and maximum 561 562 projection, images were segmented by filtering the images using texture properties based on 563 the FM4-64 channel to remove any background. The image region was filled and resized and 564 border objects were excluded to include only whole objects. The image region was further 565 calculated using FM4-64 and DAPI fluorescence and individual bacteria were defined. 566 (*single planes were analysed for S. Typhimurium) (B) Bacterial morphology and stain intensity properties were calculated using DAPI, SYTOX Green and FM4-64 fluorescence. 567 568 Finally, a linear classifier was used to train the software to define single bacterial cells and

- 569 exclude any artefacts. (C) Using basic brightfield correction and maximum projection, the
- 570 bacterial region was defined using a calculated image based on DAPI and FM4-64 channels.
- 571 Individual bacteria were identified within the image region, and the bacterial regions were
- 572 defined and resized into individual bacterial cells. Any artefacts were removed using size
- 573 filters. (D) Bacterial morphology and stain intensity properties were calculated using DAPI,
- 574 SYTOX Green and FM4-64 fluorescence.
- 575

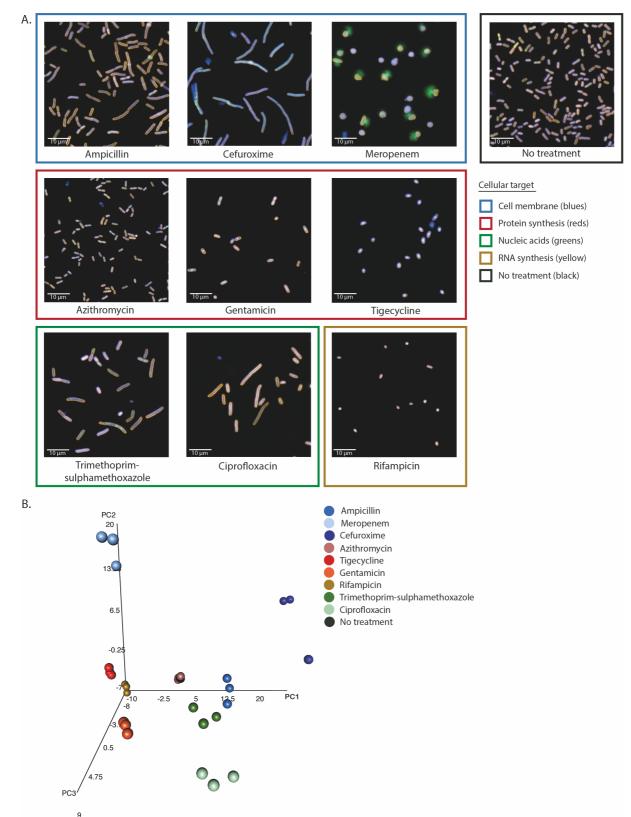


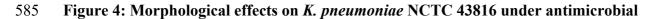


577 Figure 3: Optimising plate coating for bacterial adhesion. Isolates were grown in ultra578 thin 96 well plates on different surface matrices and the Harmony analysis pipelines were
579 used to count the number of adherent bacteria after fixing, washing and staining. Graphs are

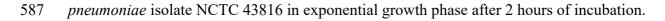
- 580 comparing the adhesion of two representative isolates of *K. pneumoniae* (A), *S.* Typhimurium
- 581 (B) and S. aureus (C) on each substrate. Error bars represent standard deviation of three
- 582 biological replicates.

583

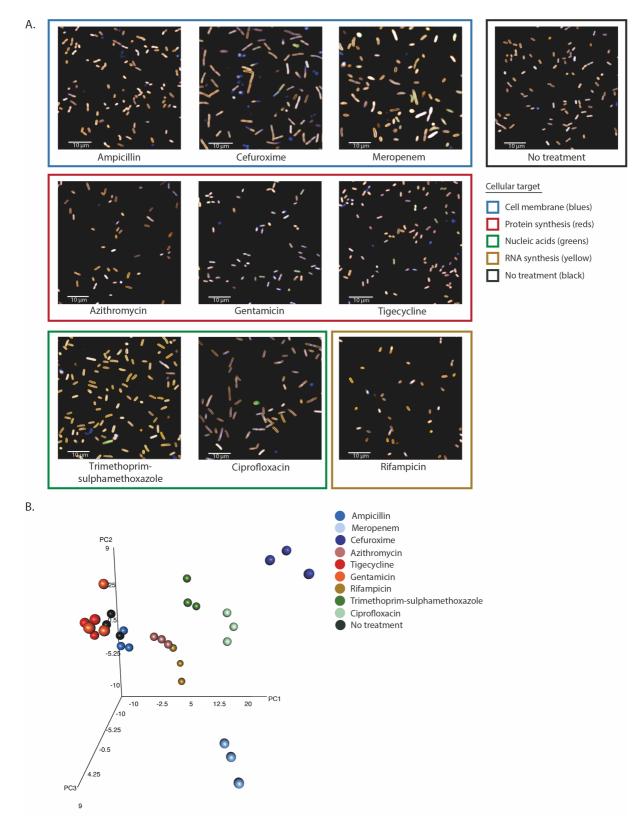




pressure. (A) Representative images of the effect of different antimicrobials on the *K*.



- 588 Antimicrobials are grouped by similar cellular targets. Bacteria were stained with FM4-64,
- 589 DAPI and SYTOX green. Images were acquired on an Opera Phenix using a 63x water
- 590 immersion lens. (B) Three-dimensional principal component analysis of the mean and
- 591 standard deviation values of 62 morphological properties measured for single bacterial cells
- 592 in each well. Technical triplicate repeats are shown.



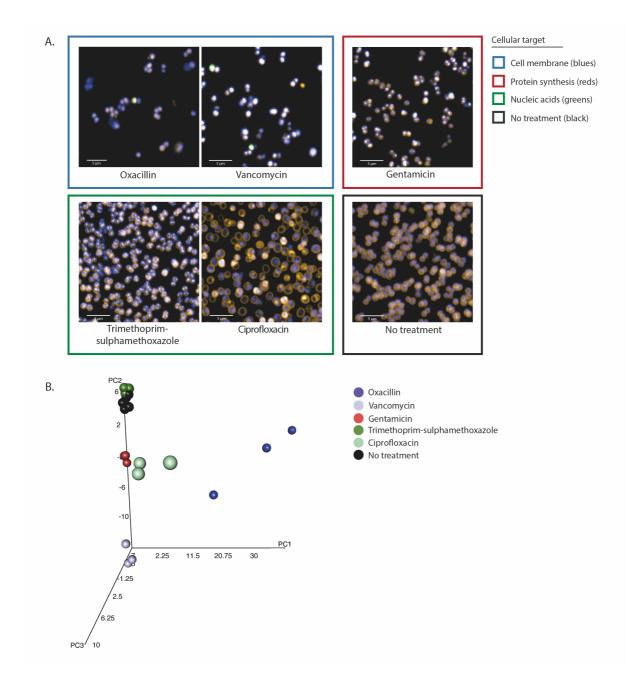
593



595 pressure. (A) Representative images of the effect of different antimicrobials on the S.

596 Typhimurium isolate NCTC 13348 in exponential growth phase after 2 hours of incubation.

- 597 Antimicrobials are grouped by similar cellular targets. Bacteria were stained with FM4-64,
- 598 DAPI and SYTOX green. Images were acquired on an Opera Phenix using a 63x water
- 599 immersion lens. (B) Three-dimensional principal component analysis of the mean and
- 600 standard deviation values of 62 morphological properties measured for single bacterial cells
- 601 in each well. Technical triplicate repeats are shown.
- 602
- 603
- 604



605

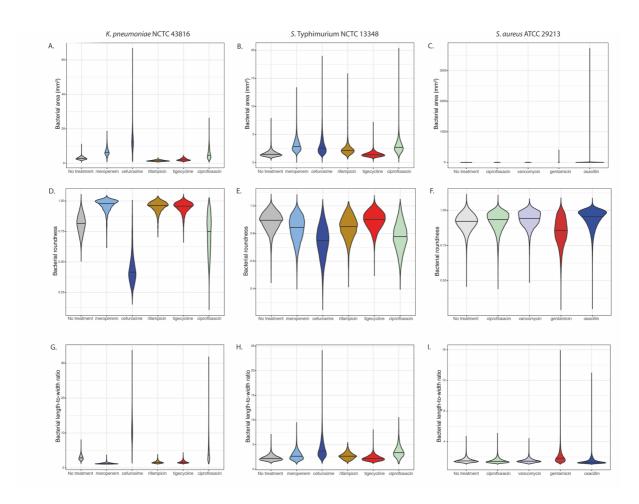
606 Figure 6: Morphological effects on S. aureus ATCC 29213 under antimicrobial

607 pressure. (A) Representative images of the effect of different antimicrobials on the S. aureus

608 isolate ATCC 29213 in exponential growth phase after 2 hours of incubation. Antimicrobials

- are grouped by similar cellular targets. Bacteria were stained with FM4-64, DAPI and
- 610 SYTOX green. Images were acquired on an Opera Phenix using a 63x water immersion lens.
- 611 (B) Three-dimensional principal component analysis of the mean and standard deviation

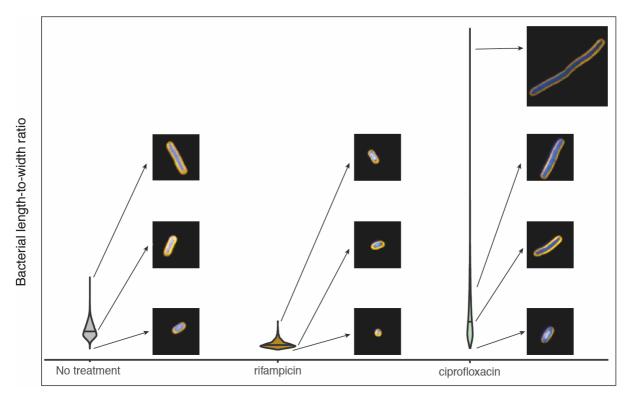
- 612 values of 62 morphological properties measured for single bacterial cells in each well.
- 613 Technical triplicate repeats are shown.
- 614
- 615



616

Figure 7: Comparison of individual basic morphological measurements. Violin plots of
bacterial area (A-C), bacterial roundness (D-F) and bacterial length-to-width ratio (G-I)
comparing *K. pneumoniae* NCTC 43816 and *S.* Typhimurium NCTC 13348 treated with
meropenem, cefuroxime, rifampicin, tigecycline and ciprofloxacin, and *S. aureus*ATCC29213 treated with ciprofloxacin, vancomycin, gentamicin and oxacillin, with
untreated controls.

623



625 Figure 8: Example of population-level length heterogeneity of *K. pneumoniae* NCTC

626 **43816.** Violin plot of bacterial length-to-width ratio comparing untreated, rifampicin treated,

- 627 and ciprofloxacin treated K. pneumoniae NCTC 43816 with inset images demonstrating the
- 628 different phenotypes observed in the same growth conditions within a single well.
- 629

624