

1 **Title:**

2 Eco-evolutionary mechanisms driving within-patient emergence of bacterial antimicrobial
3 resistance

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16 **Competing interest statement**

17 The authors declare no competing interests.

18

19 **Abstract:** The eco-evolutionary mechanisms of AMR emergence within patients and how these vary
20 across bacterial infections are poorly understood. Increasingly widespread use of pathogen genome
21 sequencing in the clinic now enables deeper understanding of these processes. Here, we review the
22 clinical evidence supporting four major mechanisms of within-patient AMR emergence in bacteria: (i)
23 spontaneous resistance mutations, (ii) *in situ* horizontal gene transfer of resistance genes, (iii) selection
24 of pre-existing resistance, and (iv) immigration of resistant lineages. Within-patient AMR emergence
25 occurs across a wide range of host niches and bacterial species, but the importance of each mechanism
26 varies between bacterial species and infection sites within the body. We identify potential drivers of
27 such differences and discuss how eco-evolutionary analysis could be embedded within clinical trials of
28 antimicrobials, which are powerful but underutilised tools for understanding why eco-evolutionary
29 mechanisms vary between pathogens, infections and individuals. Ultimately, improving understanding
30 of how host niche, bacterial species, and antibiotic mode of action combine to govern the eco-
31 evolutionary mechanism of AMR emergence in patients will enable more predictive and personalised
32 diagnosis and antimicrobial therapies.

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37 **1.0 - Introduction**

38

39 The antimicrobial resistance (AMR) crisis threatens to endanger modern medicine within the next 20-
40 30 years¹, with an estimated 4.95 million deaths associated with bacterial AMR in 2019². Whilst the
41 transmission of antibiotic resistant infections between patients, animals and their environment has
42 received much attention³⁻⁷, a key part of the problem that is less well studied is the evolutionary
43 emergence of bacterial AMR within patients during treatment^{8,9}. Low-cost pathogen genome
44 sequencing is increasingly revealing instances of within-patient emergence of AMR across a wide
45 variety of infection types, therapeutic regimens, and important bacterial pathogens (including
46 *Mycobacterium tuberculosis*¹⁰, *Pseudomonas aeruginosa*^{11,12}, *Staphylococcus aureus*^{13,14}, and
47 *Klebsiella pneumoniae*¹⁵). Moreover, the emergence of AMR within-patients has been associated with
48 treatment failure, increasing morbidity, and increased numbers of deaths¹⁶.

49

50 The emergence of AMR within patients is driven by antibiotic-mediated selection for resistant
51 genotypes, promoting the dominance of resistant lineages over their susceptible competitors⁹. Indeed,
52 at the population level increased outpatient antibiotic use correlates with increased antibiotic
53 resistance prevalence^{17,18}, in part due to increasing selection for resistance within treated patients.
54 Less clear, however, are the precise ecological and evolutionary mechanisms (termed collectively
55 eco-evolutionary mechanisms here for brevity) through which AMR emerges within patients, their
56 relative importance, and how this varies among bacterial pathogens, infection sites and with host
57 ecological factors (such as, nutrient supply, turnover, or immune responses, among other factors
58 explored further in the third section of this review). This is of clinical importance, as the speed of
59 resistance emergence will likely vary depending on which eco-evolutionary mechanism predominates
60 in a particular host niche and infection: for example, the presence of hypermutable strains, pre-
61 existing resistant lineages, or mobile elements capable of transferring resistance genes *in situ* can all
62 theoretically accelerate resistance emergence within patients¹⁹. Different eco-evolutionary modes of
63 resistance emergence can additionally have contrasting effects upon the long-term stability and
64 persistence of resistance: on average, horizontally transferred resistance genes impose lower fitness
65 costs compared to spontaneous resistance mutations²⁰, meanwhile the magnitude of fitness costs can
66 vary according to the antibiotic target and/or between bacterial pathogen species²¹. Consequently,
67 understanding which evolutionary mechanism dominates in a particular infection setting could guide
68 the optimal treatment strategy. For example, informing the optimal choice of antibiotic and/or other
69 interventions, such as decolonisation procedures to remove pre-existing resistant lineages or stricter
70 infection control measures to limit opportunities for exogenous resistant lineages to superinfect.

71

72 While there is a rich body of theory and lab experiments investigating the molecular and eco-
73 evolutionary mechanisms of AMR, such approaches inevitably do not capture many key features of
74 the complexity of within-host environments likely to mediate AMR emergence *in vivo*, including the
75 physiological heterogeneity of infection sites as well as the host's immune responses. Although
76 evidence for the clinical importance of within-patient bacterial AMR evolution is rapidly growing,
77 there is little crosstalk between these fields limiting the translation of eco-evolutionary concepts into
78 the clinic. Here, we review the fast-growing body of clinical case reports describing within-patient
79 AMR emergence and synthesise this clinical literature with eco-evolutionary theory to classify the
80 underlying causal mechanisms driving within-patient AMR emergence in the clinic. We focus on
81 bacterial infections, contrasting a wide range of pathogens and infection sites, and using studies
82 documenting emergence of resistance within patients, where individuals were sampled before and
83 after antibiotic treatment and the genetic mechanisms of resistance responsible were identified. Our
84 focus is distinct from previous AMR reviews, which nonetheless are highly complementary to ours,
85 including those focusing on a within-patient dynamics of AMR in single-species^{10,15}, population
86 genetics/genomics of AMR²², molecular mechanisms of AMR²³, or One-Health²⁴ and environmental
87 dimensions^{25,26} of AMR. We highlight gaps in current understanding and identify how current
88 practices, particularly pertaining to clinical trials for new antibiotic treatments, could be augmented to
89 deliver greater understanding of eco-evolutionary mechanisms of AMR emergence, which will
90 ultimately improve prediction and control of AMR. More accurate prediction of AMR emergence
91 may allow for more informed and personalised antimicrobial treatment plans, to slow the rate of
92 resistance emergence, increase the efficacy of existing treatments, and facilitate the development of
93 novel therapies.

94

95 **2.0 - The eco-evolutionary mechanisms of within-patient AMR emergence**

96

97 We highlight four major eco-evolutionary mechanisms by which AMR can emerge: (i) spontaneous
98 resistance mutations, (ii) *in situ* horizontal gene transfer of resistance genes, (iii) selection of pre-
99 existing resistant lineages, and (iv) immigration of resistant lineages. Broadly, these can be defined as
100 either *de novo* mechanisms driven by the formation of new resistant genotypes (i and ii), or
101 mechanisms selecting on pre-existing (or standing) genetic variation for resistance (iii and iv) (Fig. 1).
102 In this section, we summarise the clinical evidence for these mechanisms occurring within-patients
103 treated with antibiotics and discuss how these mechanisms may modulate the response of an infection
104 to antimicrobial therapy. It is important to note that these mechanisms are not necessarily mutually
105 exclusive, they may co-occur and interact, and in many clinical situations the boundaries between
106 mechanisms may be blurred and/or the necessary evidence may not exist to definitively distinguish

107 between them. Key challenges of identifying and distinguishing these mechanisms are explored
108 further in supplementary Box S1.

109

110 **2.1 - Spontaneous resistance mutations**

111

112 Spontaneous resistance mutations include point mutations, insertions/deletions, and other genetic
113 rearrangements that can increase resistance to an antimicrobial. They occur randomly within a
114 previously susceptible population and are positively selected during antimicrobial treatment (also
115 referred to as *de novo* mutation). Case reports of spontaneous mutations occurring within-patient
116 during treatment are relatively common, with large numbers of single-patient case studies (e.g., for
117 example 21 out of 27 clinical studies cited in the paragraph below are single-patient reports). This
118 literature covers a wide range of infection sites throughout the body (Fig. 2), particularly in the
119 lung²⁷⁻³⁴ and bloodstream infections³⁵⁻⁴³, but also including urinary tract infections⁴⁴⁻⁴⁶, infections of
120 burns⁴⁷, the endocardium^{48,49}, gut⁵⁰, liver cysts⁵¹, biliary tract³⁶, cerebrospinal fluid⁵² and implant
121 associated bone infections⁵³.

122

123 Spontaneous resistance mutations occurring within patients can be grouped into three functional
124 categories: i) Target modifying mutations that reduce antibiotic efficacy for example by reducing
125 drug-binding. These include mutations in *gyrAB* (DNA gyrase enzyme) or *parCDE* (DNA
126 topoisomerase IV enzyme) associated with ciprofloxacin resistance^{29,39}, *rpoB* (RNA polymerase β -
127 subunit) conferring rifampicin resistance²⁹, *rpsJ* (Ribosomal protein subunit S10) conferring
128 tetracycline resistance³³, or mutations affecting penicillin-binding-protein (PBP) genes for resistance
129 to β -lactams^{32,47,54,55}. Such mutations sometimes confer cross-resistance, typically to drugs of the
130 same class or those with related targets⁴¹. ii) Mutations modifying pre-existing resistance genes to
131 enhance or modify the spectrum or level of resistance they confer, e.g., modification of the *tetA* efflux
132 pump gene resulting in enhance tigecycline export³¹, and mutation of the PDC cephalosporinase gene
133 conferring both ceftolozane-tazobactam and ceftazidime-avibactam resistance^{36,56} (again, likely due
134 to the shared mechanism of these drugs); iii) regulatory mutations affecting resistance gene
135 expression, for example mutation of the regulator *ampD* resulting in upregulation of AmpC β -
136 lactamase⁵³, loss of repression from repressors RamR or AdeS conferring tigecycline resistance
137 through increased AcrAB^{43,46} and AdeABC⁵⁷ efflux pumps, respectively, and loss of MexT regulator
138 function derepressing *mexAB* efflux pump expression to confer ciprofloxacin and other resistances⁵⁸.
139 Since efflux systems commonly export multiple antibiotics, including those belonging distinct classes
140 such as ciprofloxacin and meropenem resistance efflux by AdeABC⁴⁷, increased expression of efflux
141 can lead to multidrug resistance.

142

143 Provided that they escape genetic drift⁵⁹, genotypes carrying new spontaneous resistance mutations
144 are expected to increase in frequency driven by antibiotic selection, displacing their susceptible
145 neighbours⁶⁰. Direct evidence for such selective sweeps within patients is limited due to requiring
146 intensive longitudinal sampling of multiple bacterial isolates to capture these evolutionary dynamics.
147 However, within-patient selective sweeps have been observed, principally in chronic lung infections:
148 For example, in *P. aeruginosa* lung infections, *pbpB* (penicillin-binding-protein) mutation during
149 aztreonam therapy of one patient⁵⁵, *ampD* (β -lactamase) and *oprD* (outer membrane porin) mutation
150 during meropenem and ceftazidime therapy²⁷, and *oprD*, *wbpM* (O-antigen biosynthesis enzyme) and
151 *mexAB/oprM* (efflux pump and porin) during carbapenem and colistin combined therapy⁶¹, have all
152 been observed to occur by spontaneous mutation and sweep to high frequency within-patients driven
153 by antibiotic treatment. Similarly, in a report of *M. tuberculosis* infection, resistance to multiple
154 antibiotics emerged by spontaneous mutation and swept to high frequency in a single patient across 9
155 years of therapy, with rises in frequencies of mutations to the *katG*, (catalase-peroxidase enzyme),
156 *embB* (Mycobacterial cell wall synthesis enzyme), *rpoB* (RNA polymerase β -subunit), *rpsL/rrs*
157 (Ribosomal protein S12 / 16S rRNA) and *ethA* (drug-activating monooxygenase enzyme) genes²⁹,
158 observed during antibiotic treatment with isoniazide, ethambutol, rifampicin, streptomycin, and
159 ethionamide, respectively. Interestingly, in the *M. tuberculosis* study, fluctuations in prevalence of
160 particular resistance mutations occurred during the course of therapy, with the predominant
161 streptomycin resistance mutation switching from an *rpsL* to an *rss* mutant. Fitness costs associated
162 with spontaneous resistance mutations can also drive their loss after treatment is withdrawn^{57,62}. More
163 complex selection dynamics among competing spontaneous resistance mutations have been observed
164 in lung infections caused by these organisms: For example, replacement of an original *rpsL* mutation
165 conferring streptomycin resistance by a different mutation in the same gene was observed in an *M.*
166 *tuberculosis* lung infection during treatment⁶². In a *P. aeruginosa* lung infection, two coexisting *pbpB*
167 mutations providing resistance to aztreonam fluctuated in frequency during long-term treatment with
168 combinations of 10 different antibiotics. These *pbpB* mutations provided differing levels of cross-
169 resistance against ceftazidime and piperacillin⁵⁵, drugs that were also used periodically during
170 treatment of this patient.

171

172 Supply of spontaneous resistance mutations is determined by the pathogenic bacterium's population
173 size and mutation rate, but can also be affected through stress-induced mutagenesis⁶³ or exposure to
174 chemical or physical mutagens (e.g., reactive oxygen species produced by host inflammatory immune
175 responses⁶⁴, or the antibiotic treatment itself^{65,66}) which can vary with infection site. Mutation
176 frequency may also vary during different stages of an infection, for example an elevated mutation rate

177 ten times higher during the initial infection phase compared to established infection has been reported
178 for *Helicobacter pylori* infection. This study implicated elevated production of mutagenic reactive
179 oxygen species by the host inflammatory response⁶⁷. Note, however, other *H. pylori* studies have not
180 reported such variation in mutation supply with infection stage⁶⁸. Alternatively, infecting populations
181 may evolve elevated mutation rates through gaining mutations in mismatch repair systems that
182 substantially increase the supply of all mutations including those involved in AMR, and thus
183 potentially accelerating the emergence of AMR⁶⁹. Such hypermutator lineages have frequently been
184 observed to arise in chronic lung infections, particularly in cystic fibrosis patients^{54,70}. Hypermutators
185 have been reported to contribute to within-patient AMR emergence by increasing supply of
186 spontaneous resistance mutations in lung infections caused by *P. aeruginosa*^{54,71}, *Achromobacter*
187 *spp.*^{30,72}, *Burkholderia pseudomallei*⁷³, and *Stenotrophomonas maltophilia*^{74,75}. Hypermutable *S.*
188 *aureus* genotypes with mutations affecting recombinational repair protein RecQ have also been
189 implicated in driving within-patient AMR emergence⁷⁶. However, beyond individual case reports, the
190 link between hypermutation and accelerated AMR evolution is debated, with conflicting evidence on
191 whether hypermutators really are more likely to show higher AMR than non-hypermutators on
192 average⁷⁷.

193

194 **2.2 – *In situ* horizontal gene transfer**

195

196 Within-patient AMR emergence in a focal pathogen has been observed in a range of clinical studies to
197 occur through *in situ* horizontal gene transfer (HGT)^{78–83}. This evolutionary process, whereby
198 resistance determinants are exchanged between bacterial cells, is often mediated by mobile genetic
199 elements (MGEs), such as plasmids, transposons and bacteriophages, but can also occur via uptake of
200 DNA from the extracellular environment (natural competence)⁸⁴. Despite long-standing interest in
201 HGT of AMR, reports of this process within-host are almost entirely confined to the gut^{78,80–83}, with
202 some limited reports of MGEs aiding AMR evolution through transposons increasing resistance gene
203 copy number in bloodstream infection⁸⁵. It is unclear whether this pattern represents a fundamental
204 difference between the gut and other niches (e.g., higher microbial diversity or the species causing
205 infections, explored further in the third section of this review) or a bias in the literature.

206

207 Of particular concern are cases where harmful bacterial pathogens can gain new resistance genes by
208 HGT from coexisting commensals within the patient's microflora, which has most commonly been
209 observed in the gut where high diversity microbial communities coexist at high densities^{86,87}. Plasmid
210 transmission in the gut is likely highly frequent. For example, *Clostridium difficile* gained a 46 Kbp
211 conjugative plasmid with the *ermB* clindamycin resistance gene from enteric commensal

212 *Faecalibacterium prausnitzii* during therapy with clindamycin⁸⁰. *K. pneumoniae* isolated from the gut
213 of a patient undergoing treatment for bacteraemia gained a 36 Kbp plasmid, enabling duplication of
214 several resistance genes⁸¹ during therapy with a variety of antimicrobials. A *Salmonella enterica*
215 infection treated with ceftriaxone has also been reported gaining a 309 Kbp plasmid within-patient
216 containing three β -lactamase genes (*bla*_{CTX-M15}, *bla*_{TEM-1b}, *bla*_{OXA-30}) conferring ceftriaxone resistance,
217 transferred from an unknown donor species, but likely one co-existing in the gut microflora⁸⁸.
218 Additionally, extensive within-patient transfer of a *bla*_{OXA-48} β -lactamase carrying plasmid pOXA-48
219 has been shown to occur in the gut microbiomes of intensive care patients. The same plasmid was
220 isolated from 8 different bacterial species across 105 different patients and transfer of the plasmid was
221 directly observed in at least five different patients⁸². Transfer of this plasmid has also been reported
222 within the gut of a patient from *K. pneumoniae* to *Escherichia coli* during amoxicillin treatment⁷⁸.
223 This pOXA-48 plasmid acted not only as a vehicle for AMR transfer, but also as a catalyst for the
224 evolution of altered levels of resistance: In one patient receiving meropenem treatment, pOXA-48
225 evolved a higher plasmid copy number conferring enhanced meropenem resistance, whereas in other
226 patients, the *bla*_{OXA-48} gene was lost from the plasmid, reducing the plasmid fitness cost in the absence
227 of antibiotic⁸³.
228
229 Transposable elements also play an important role mobilizing AMR genes between bacterial
230 chromosomes, plasmids and other MGEs, and have been shown to contribute to within-patient AMR
231 emergence. For example, the *bla*_{KPC-53} carrying transposon Tn4401 providing resistance to
232 imipenem/relebactam expanded in copy number during therapy with ceftazidime/avibactam by
233 transposing to multiple plasmids in the *K. pneumoniae* genome leading to enhanced resistance⁸⁵.
234 Transposable elements are also important vectors for transferring resistance mechanisms between
235 lineages, for example IS26-mediated transfer of the *bla*_{NDM-1} metallo- β -lactamase drove a multi-
236 species hospital outbreak of carbapenem resistant infections, with spread of the *bla*_{NDM-1} gene across
237 *E. coli*, *K. pneumoniae*, *Citrobacter freundii*, *Morganella morganii* and *Enterobacter cloacae* isolates,
238 the majority of which were isolated from rectal swabs⁸⁹.
239
240 The importance of bacteriophages in driving *in situ* HGT of AMR within patients is debated. Analysis
241 of phage genomes from human microbiomes found that these rarely encode AMR genes⁹⁰, a pattern
242 supported by data from pigs⁹¹. However, a study of human faecal phages found higher frequency of
243 AMR genes in phage particles following ciprofloxacin treatment⁹², suggesting phage transduction
244 may more commonly transfer AMR in patients exposed to antibiotics and in particular those known to
245 induce phage lysis, such as ciprofloxacin⁹³. For some organisms, transduction is well known to play
246 an important role in adaptation to the human host. This is the case for *S. aureus*⁹⁴, for which efficient

247 transduction of a β -lactamase-containing plasmid has been demonstrated *in vitro*⁹⁵, indicating this
248 mechanism of AMR transfer may be likely to also act within-patient.

249

250 **2.3 - Selection of pre-existing resistant genotypes**

251

252 Resistant genotypes are sometimes stably present prior to treatment at an appreciable frequency
253 detectable by standard culturing within an infection, for example due to prior history of treatment with
254 the same antibiotic. Selection of such pre-existing resistant genotypes can accelerate AMR emergence
255 by negating any waiting time for new spontaneous mutations to arise⁹⁶, and can display lower fitness
256 costs⁹⁷. Moreover, by being present at appreciable frequencies, pre-existing resistant genotypes are
257 less likely lost to drift and can more rapidly increase in frequency upon commencement of treatment
258 than any spontaneous mutant⁶⁰. Mixed-strain lung infections of *P. aeruginosa* evolve resistance faster
259 due to selection of pre-existing resistant strains, with mixed-strain infections showing a ~20% greater
260 increase in resistance after treatment than single-strain infections⁹⁸. In the same study, patients with
261 pre-existing resistant strains ranging in their pre-treatment frequency from ~5% to ~60% displayed
262 rapid increases in the frequency of these resistant strains upon treatment. Examples of pre-existing
263 resistant lineages contributing to the evolution of AMR within patients have been reported for both
264 acute and chronic infections of the lung^{27,98,99}, the stomach^{100,101}, and some evidence for this effect in
265 the bowel¹⁰². As the global prevalence of resistant infections increases¹⁰³, this mechanism may be
266 increasingly common as infection or colonisation with pre-existing resistant bacterial lineages
267 becomes more likely.

268

269 Antibiotic treatment has been shown to trigger rapid increases in pre-existing resistance allele
270 frequency within an infecting population: during acute *P. aeruginosa* infection of a single patient,
271 *nalD* (*mexAB-oprM* efflux pump repressor), *anmK* and *sltBI* (both peptidoglycan metabolism
272 enzymes) resistance alleles at 7-8% frequency pre-treatment increased to 44-49% frequency after 12
273 days of combination therapy which started with piperacillin-tazobactam, followed by ciprofloxacin
274 and cefepime²⁷. Such selection works both ways however: the same study documented a separate
275 patient for which two independent alleles of *mexR* (repressor of *mexAB-oprM* efflux pump) conferring
276 levofloxacin resistance went extinct within 5 days of treatment with piperacillin-tazobactam only,
277 indicating purging of pre-existing resistance alleles that provided no selective advantage under the
278 prevailing treatment regime.

279

280 Clinical studies further suggest that the frequency of a pre-existing resistant lineage can offer an
281 indication of the likelihood of resistance evolution upon treatment: for example, across 200 *M.*

282 *tuberculosis* lung infections, the presence of pre-existing resistant lineages at frequencies of $\geq 19\%$
283 was found to be highly predictive of subsequent fixation of resistance following antibiotic
284 treatment¹⁰⁴. Complete fixation of pre-existing resistant lineages following treatment does not always
285 occur however, and can be highly localised to specific host niches. For example, in *H. pylori* stomach
286 infections, pre-existing resistance against clarithromycin, ciprofloxacin, and metronidazole were
287 rarely fixed following antibiotic treatment, with 20-60% of isolates remaining drug sensitive¹⁰⁰.
288 Moreover, from the same study, all isolates from one patient's stomach corpus were resistant to
289 ciprofloxacin, but all isolates from the stomach antrum remained susceptible. This example highlights
290 the potential complexities of resistance mutation dynamics along with the variable impacts of
291 infection biogeography within hosts.

292

293 The probability of pre-existing resistant lineages being present can depend on a variety of factors.
294 Prior antibiotic therapy is a probable driver as this can result in dynamic shifts in microbiome
295 community structure and stable increases in the frequency of AMR lineages after treatment
296 ceases^{105,106}, which in turn can increase the future risk of both antibiotic resistant and sensitive
297 infection for a particular patient¹⁰⁷. Another likely cause, especially in the context of chronic
298 infection, is the duration of the patient's infection because multiple generations of replication during
299 establishment of infection can enable diversification and the accumulation of resistance mutations
300 before any antibiotic is given: for example, up to 4% of patients colonised by *H. pylori* gain resistance
301 mutations during the infection establishment phase but before antibiotic treatment, which
302 subsequently drove treatment failure even for sequential multi-drug treatment regimes¹⁰¹. The longer
303 evolutionary history of pre-existing resistant lineages within a patient may enable compensatory
304 evolution, reducing fitness costs of resistance and enhancing their long-term persistence.

305

306 **2.4 - Immigration of resistant genotypes from elsewhere in the body**

307

308 Even if antibiotic treatment is successful at clearing the focal bacterial pathogen infection from a
309 particular body site, an AMR infection can still potentially occur by immigration of resistant
310 genotypes from elsewhere in the body. This ecological process, also referred to as strain translocation
311 or strain replacement, has received less attention than the other eco-evolutionary mechanisms, but
312 appears to be important for certain kinds of infections and body sites, including the urinary tract¹⁰⁸,
313 lungs¹⁰⁹, skin and wounds^{110,111}, and bloodstream infections¹¹². Exogenous sources of resistant
314 strains, for example from the wider hospital environment, can also play a role in AMR transmission
315 distinct from within-host processes, but are not our focus here and have been reviewed elsewhere¹¹³.

316

317 The most robust evidence to date for immigration of resistant genotypes from other sites within the
318 patient's body driving resistance emergence is a recent study investigating ~140,000 urinary tract
319 infection (predominantly caused by *E. coli*) and ~7000 wound infections (predominantly caused by *S.*
320 *aureus*). Here, the predominant mode of AMR emergence was immigration of resistant lineages to the
321 site of infection from within the patient's own microflora¹⁰⁸. This included immigration of resistant
322 lineages belonging to other species (*K. pneumoniae* and *Proteus mirabilis*) when patients had been
323 treated with antibiotics against which resistance in *E. coli* is rarely seen (fosfomycin and
324 nitrofurantoin), or invasion by other *E. coli* lineages already resistant to ciprofloxacin (*gyrB* and *parC*
325 mutations) even though resistance evolution by spontaneous mutations of these genes in *E. coli* is well
326 documented. Similarly, in a study of four patients with MRSA soft tissue infections, in two patients
327 strain replacement by MRSA lineages resistant to aminoglycoside and fluoroquinolone antibiotics
328 drove emergence of resistance upon treatment with antibiotics from these drug classes¹¹⁰. Strain
329 replacement has also been reported for *K. pneumoniae* lung infections, where carbapenem resistant
330 lineages of a different sequence type were observed to replace drug sensitive lineages in 36 of 44
331 patients who received carbapenem therapy¹¹⁴ - however this study does not conclusively rule out that
332 the invading sequence types were not present prior to treatment. A vancomycin sensitive
333 *Enterococcus faecium* lineage from a bloodstream infection was replaced during vancomycin therapy
334 by resistant lineages that were present alongside the sensitive strain in the patient's gut, from where
335 the bloodstream infection had arisen¹¹². Similarly, immigration of resistant genotypes from within the
336 patient's own gut microbiota has also been observed in pneumonia¹¹⁵, sepsis, and acute respiratory
337 distress syndrome¹¹⁶. A key challenge here is that few studies definitively identify the source of the
338 immigrating resistant strain, in part because this requires in depth study of the patient's microbiome
339 and potentially the wider hospital environment, for example using metagenomics, which is costly and
340 raises ethical considerations.

341

342 **2.5 - Mixed and interacting mechanisms**

343

344 Although it may often be challenging to distinguish between different eco-evolutionary mechanisms
345 due to the practical limitations of clinical studies (explored further in supplementary Box S1), in other
346 cases, multiple mechanisms could be at play and potentially interact, complicating interpretation.
347 There are several clinical reports where immigration of strains from another body site is followed by
348 the evolution of resistance by spontaneous mutation. These include a chronic MRSA wound and
349 bloodstream infection, which was treated with multiple Gram positive-targeting antibiotics
350 (daptomycin, clindamycin and vancomycin) driving replacement of MRSA by a succession of three
351 genetically distinct lineages of *Enterobacter hormaechei*, one of which then evolved meropenem
352 resistance by spontaneous mutation¹¹¹. Similarly, in a lower respiratory tract infection, immigration

353 of a *P. aeruginosa* strain from the gut to the lung was followed by evolution of enhanced resistance
354 through spontaneous mutation at both body sites¹¹⁵. These and other studies highlight the risk that
355 systemic antibiotic treatments are likely to enrich for resistance at non-targeted body-sites,
356 potentiating subsequent immigration of resistant lineages from this reservoir¹¹².

357

358 **3.0 - Genetic, ecological, and host factors shaping within-patient AMR emergence**

359

360 Our review of the clinical literature highlights that each eco-evolutionary mechanism occurs within
361 patients. In the following section, we explore the factors driving variation in the predominance of
362 these mechanisms across infections, including taxonomic/genetic, ecological, and host factors, and
363 synthesize the evidence for their action within-patients. The probability of each eco-evolutionary
364 mechanism is likely to vary with each of these factors, bearing implications for any attempts to predict
365 which mode is likely to predominate (Box 1). In addition, since AMR emergence has been subject of
366 intensive theoretical and experimental evolution investigation¹¹⁷⁻¹²¹, we also highlight other factors
367 known to influence AMR emergence in these *in silico* and *in vitro* studies, and consider how these
368 fundamental discoveries could be translated to improve understanding of clinical AMR emergence.

369

370 **3.1- Variation between bacterial species and strains**

371

372 It is unlikely that any bacterial species evolves AMR exclusively through a single eco-evolutionary
373 mode, but some species appear to preferentially use certain modes more than others. This is evident in
374 contrasting pangenome structures among species, wherein higher proportions of accessory (genes
375 variable among strains) versus core (genes common to all strains) genome compartments indicates
376 greater propensity for HGT. For example, whereas the accessory genome makes up ~80% of the *E.*
377 *coli* pangenome, the accessory genome is ~0-1% of the *M. tuberculosis*¹²² pangenome. Accordingly,
378 *M. tuberculosis* is rarely reported engaging in HGT¹²², perhaps owing in part to limited interactions
379 with other bacteria due to its intracellular infection lifestyle. Thus, HGT is a relatively unimportant
380 mode of AMR emergence for this organism, and *M. tuberculosis* is primarily reported to evolve AMR
381 through spontaneous mutation^{99,123,124}. Similarly, an epidemiological study of *Streptococcus*
382 *pneumoniae* carriage found that the rate of HGT was unimportant for determining the frequency of
383 AMR, which was better explained by selection for pre-existing resistance¹²⁵. Conversely, HGT of
384 resistance likely plays an important role in pathogens with larger accessory genomes, such as *S.*
385 *aureus*, *E. coli*, and *K. pneumoniae*, where MGEs make up substantial fractions of genomes and
386 commonly encode resistance genes^{94,126,127}. The drivers of taxonomic differences in rates of HGT are
387 not well-understood but are likely to have complex ecological and genomic causes, for example
388 intracellular pathogens (like many symbionts) are cut-off from the supply of mobile genetic elements,

389 whereas genome defence systems, including restriction modification or CRISPR-Cas, can act as
390 barriers to gene exchange.

391

392 Mutation supply rates can vary extensively between species¹²⁸, among strains within species¹²⁸, and
393 even for specific resistance-conferring genes¹²⁹, within a genome. Such differences are due to a
394 variety of molecular and genetic factors, which can additionally vary with prevailing physiological
395 and environmental conditions¹³⁰. Additionally, the frequency of hypermutator lineages varies
396 extensively, ranging from 0.5-2% in *E. coli* and *Haemophilus influenzae*, up to >50% in *Neisseria*
397 *meningitidis* and *P. aeruginosa* infections¹³¹. Higher mutation supply can potentiate spontaneous
398 resistance mutations and/or promote the accumulation of pre-existing standing genetic variation,
399 potentially enhancing the contribution of either of these modes of AMR emergence. Note, however,
400 that hypermutation can be detrimental after the initial gain of resistance due to subsequent
401 accumulation of deleterious mutations reducing fitness, particularly in highly competitive
402 polymicrobial communities¹³². Accordingly, there are extensive case reports of *P. aeruginosa*
403 evolving AMR by spontaneous mutation within lung infections^{28,36,54,133-135}, as well as bloodstream
404 infection⁷¹ and infections of the gut¹¹⁵. Nonetheless, both *M. tuberculosis*^{62,104} and *P. aeruginosa*^{27,98}
405 have also been reported to evolve AMR via selection upon pre-existing resistant lineages, and for *P.*
406 *aeruginosa* immigration of resistant genotypes from other body sites also plays a less well-studied
407 role¹¹⁵. Differences in evolvability between lineages can arise for a range of other reasons besides
408 mutation supply. Gain of resistance determinants by HGT can potentiate subsequent evolution of
409 resistance by spontaneous mutation¹³⁶, leading to differences in AMR evolvability attributable to the
410 presence/absence of key genes, such as efflux systems¹³⁷, per lineage. Similarly, allelic variation in
411 genes not directly related to resistance, such as metabolic genes that alter antibiotic tolerance¹³⁸ can
412 potentiate subsequent gain of resistance¹³⁹, as well as influence the ability of MDR strains displace
413 commensal bacteria in competitive host niches¹⁴⁰. Indeed, bacterial persistence, the tolerance of
414 antibiotics through cellular dormancy, has been shown in lab studies to enable subsequent evolution
415 of resistance¹⁴¹, although such dynamics have not yet been demonstrated in patients.

416

417 More comparative studies focusing on a wider range of pathogens are required to quantitatively
418 determine what drives variation in the relative contribution of the different eco-evolutionary modes of
419 AMR emergence between taxa.

420

421 **3.2 - Variation between patient body sites**

422

423 The human body constitutes a diverse range of niches, differing in moisture, pH, salinity, nutrient
424 availability, and commensal microflora abundance and diversity¹⁴². These factors are all likely to

425 influence the predominant eco-evolutionary modes of AMR emergence. Ecological factors, such as
426 transmission bottlenecks, the pathogen population size present at a body site, and the rate of
427 population turnover are likely to influence key evolutionary parameters and vary between infection
428 sites. For example, smaller transmission bottlenecks can favour more frequent but costlier resistance
429 mutations, the supply of spontaneous resistance mutations will scale with population size^{60,143},
430 whereas higher rates of population turnover may strengthen selection against more costly resistance
431 mechanisms¹⁴⁴. Crucially, the properties of these parameters are rarely well-understood within
432 patients, except perhaps at the coarse-grained level of differences of magnitude between infection
433 sites.

434

435 Whereas some host niches contain abundant, diverse microbial communities (the gut, oral cavity,
436 upper respiratory tract, skin and vagina), others may be transiently colonised or contain far lower
437 microbial abundances (lower respiratory tract, urinary tract) or are typically sterile (blood,
438 cerebrospinal fluid, brain, bone, liver, kidneys, spleen)¹⁴⁵. Accordingly, the presence and abundance
439 of AMR genes will vary between body sites¹⁴⁶, and this may promote or constrain particular eco-
440 evolutionary modes of AMR emergence. For example, the gastrointestinal tract supports high
441 bacterial abundance and diversity, and consequently high AMR gene diversity^{87,147}, and as such is
442 associated with high levels of HGT of AMR, particularly mediated by conjugative plasmids^{86,148}. By
443 contrast, HGT of AMR is rarely reported in body sites with low bacterial abundance and/or diversity,
444 including the bloodstream, urinary tract, lower respiratory tract, and normally sterile sites including
445 internal vital organs. However, it must be noted that due to low accessibility and fewer studies of such
446 sites, this apparent pattern could in part be caused by under-sampling. Diverse microbial communities
447 have been shown to limit the invasion of costly resistance or hypermutators across a variety of
448 studies^{132,149,150}, whereas species interactions within diverse communities can promote increased
449 resistance, tolerance or protection of susceptible strains against antibiotics through a variety of
450 ecological mechanisms¹⁵¹.

451

452 Body sites also vary extensively in their “openness” to immigration of bacteria from other body sites.
453 Immigration of bacteria from the gastrointestinal tract to the urinary tract is particularly common
454 owing to the physical proximity of these niches. Accordingly, AMR evolution in *E. coli* urinary tract
455 infections is dominated by immigration of resistant lineages from the gut¹⁰⁸, a common reservoir for
456 drug resistant *E. coli*¹⁵², with only sporadic reports of *E. coli* evolving AMR by spontaneous mutation
457 in the urinary tract^{44,45,153}. Other clinical examples of immigration of resistant lineages include
458 ventilator associated pneumonia and wound infections, both of which are also relatively “open” body
459 sites where immigration of bacteria from other body sites is probable in hospital settings. It should

460 also be noted that the “openness” of a body site can change, particularly during critical illness,
461 wherein gut permeability can increase, leading to greater spreading of commensal bacteria and
462 opportunistic pathogens to other body sites¹⁵⁴. Conversely, for more sheltered body sites that offer
463 significant isolation from microbial communities inhabiting other body sites, reports of within-patient
464 AMR emergence most commonly involve spontaneous mutation. For example, an infection of liver
465 cysts by *E. coli*⁵¹, and an implant-associated *Cedecea davisae* bone infection⁵³ both evolved β -lactam
466 resistance through spontaneous mutations in *ampC* and *ampD* (both β -lactamases) respectively.
467 Additionally, a cerebrospinal fluid infection by *Staphylococcus capitis* evolved rifampicin resistance
468 through *rpoB* (RNA polymerase β -subunit) mutation⁵², and infections of the endocardium by
469 *Staphylococcus* spp. evolved daptomycin⁴⁹, rifampicin⁴⁸, and vancomycin⁴² resistance through *mprF*/
470 *yycH* (phospholipid synthesis enzyme / regulator of cell envelope turnover) and *rpoB* mutations
471 respectively.

472

473 Mode of delivery and the pharmacokinetic properties of antimicrobials also vary between body sites,
474 influencing which eco-evolutionary mode predominates and the wider impact of treatment on AMR
475 within the patient microbiome more generally¹⁵⁰. In accessible body sites, where drug can be
476 delivered topically, orally, or by inhalation, it is likely that a high dose can readily be achieved, which
477 could limit AMR emergence by spontaneous mutation¹⁵⁵, favouring other modes such as pre-existing
478 variation, or immigration where resistant lineages are present at appreciable frequencies prior to
479 treatment. By contrast, spontaneous mutation may be more favoured in body sites where it is
480 challenging to achieve an effective dose¹⁵⁶ due to low drug penetrance. The systemic distribution of
481 an antibiotic will also influence resistance selection in other body sites, and thus the risk of
482 immigration of resistance to the infection site, or the distribution of resistance reservoirs within the
483 host, relevant to future infections. For example, meropenem treatment for a urinary tract infection
484 selected for meropenem resistant *P. aeruginosa* present in the gut, which subsequently immigrated to
485 the lungs of an ICU patient¹¹⁵.

486

487 **3.3 - Variation with infection duration**

488

489 Both acute and chronic infections can provide ample opportunity for spontaneous resistance mutations
490 to arise, provided sufficiently large numbers of generations occur during population expansion.
491 Chronic infections provide even greater potential for the accumulation of pre-existing resistant
492 genotypes, because these infecting populations genetically diversify *in situ* and such high genetic
493 diversity is typically stably maintained^{157–159}. Moreover, hypermutator phenotypes are a common
494 feature of many chronic infections^{160–162}, further potentiating genetic diversification and the

495 emergence of AMR. Chronic infections are likely to have already experienced rounds of treatment,
496 leading to accumulation of resistant lineages within the patient that further complicate subsequent
497 treatments¹⁶³.

498 Accordingly, chronic infections are expected to have a higher likelihood of resistance emergence¹⁶⁴.
499 For example, during persistent bacteraemia caused by *S. aureus*, significant standing diversity is
500 generated during establishment of infection, with ~50% of patients acquiring resistance mutations
501 prior to treatment³⁸, that were subsequently selected during the chronic stage of the infection. As a
502 result, resistance against a wide range of antibiotics is observed in these chronic infections, including
503 vancomycin, daptomycin, linezolid and trimethoprim/sulfamethoxazole and ciprofloxacin
504 resistance^{37,39,165}. Chronic infections may also offer greater opportunity for rare events to occur, such
505 as HGT and immigration of resistant lineages. For instance, immigration of resistant strains is
506 observed in ~14% of patients with persistent *S. aureus* bacteraemia⁴⁰.

507

508 **3.4 – Variation with bacterial lifestyle**

509

510 Both chronic¹⁶⁶, and acute¹⁶⁷ infections of some body sites, such as the lungs, commonly adopt a
511 biofilm lifestyle, which could potentially influence the eco-evolutionary mode of resistance
512 emergence. The biofilm matrix and multicellular structure can offer physical protection from
513 antibiotics, which could reduce the effective dose and potentiate the emergence of resistance by
514 spontaneous mutation if, for instance, this requires the accumulation of multiple small effect
515 mutations¹⁵⁵. Moreover, biofilms can promote genetic diversification and stabilise coexistence of
516 multiple genotypes through spatial structure¹⁶⁸, potentially enabling accumulation of pre-existing
517 standing genetic variation for resistance. Biofilm structures can additionally generate oxygen
518 gradients¹⁶⁹, and be comprised of multiple microbial species, which can both further affect
519 susceptibility to antimicrobials^{170,171}. How such factors impact the eco-evolutionary mechanisms of
520 AMR emergence is not well characterised within-patients, however hypoxia can induce elevated
521 mutation rates and alter mutational spectra *in vitro* for *E. coli*¹⁷² and alter the costs of resistance genes
522 in *S. aureus*¹⁷³. The effect of biofilms on HGT appears to be complex, in some cases enhancing rates
523 of HGT¹⁷⁴, but their spatial structure can also inhibit the spread of mobile genetic elements reducing
524 HGT¹⁷⁵.

525

526 **3.5 – Variation with host immunity**

527

528 Host immune responses have been shown to alter eco-evolutionary modes of AMR evolution. For
529 example, effective host immunity suppresses diversification of *Acinetobacter baumannii*, which is
530 reversed in immunosuppressed patients¹⁷⁶. The immune system can also be vital for clearance of
531 resistant infections once they are established. For example, meropenem resistant strains of *P.*
532 *aeruginosa* that arose by spontaneous mutation were driven extinct by the host immune response in an
533 acute lung infection⁶¹. Mathematical modelling also predicts that strong immune responses that
534 persist even after bacterial population decline due to antibiotic treatment reduces the likelihood of
535 evolution of resistance¹⁷⁷. Additionally, pathogen immigration to an infection site is typically
536 suppressed by the immune system, an effect that is reduced as immune function declines¹⁷⁸,
537 potentially promoting immigration of resistant strains. Bacterial growth at an infection site can also
538 alter immune function, for instance both *P. aeruginosa* and *S. aureus* generate microaerophilic or
539 anaerobic infection microenvironments^{179,180}, affecting their sensitivity to antimicrobials^{181,182} and
540 impairing immune cell activity¹⁸³. Available clinical data suggest that very ill patients with impaired
541 immunity, such as those in intensive care units, are especially prone to secondary infection by
542 resistant strains immigrating from other body sites. For example, in a study of 310
543 immunocompromised patients with bloodstream infections, 31% suffered from reinfection, which
544 included 3 confirmed cases of reinfection with MDR bacterial strains¹⁸⁴. Additionally, in a CF patient
545 receiving a lung transplant, the allograft lungs were re-colonised by resistant *P. aeruginosa* from the
546 patient's sinuses, during prophylactic antibiotic treatment with azithromycin, imipenem and
547 tobramycin¹⁸⁵. Despite its clear potential to impact within-patient AMR emergence, the role of the
548 immune system remains relatively understudied. Further studies testing how host immunity shapes the
549 mode of eco-evolutionary AMR emergence are urgently required.

550

551 **4.0 – Future directions and Concluding remarks**

552

553 Our review demonstrates that all four eco-evolutionary mechanisms of AMR emergence occur in the
554 clinic, varying in their relative importance between infection sites and pathogen species. There are,
555 however, important limitations of the current literature (Fig. 3A; explored in-depth in Box S1): First,
556 it is likely that predominance of single-patient case reports biases the literature towards serious illness
557 or unusual treatment outcomes. Moreover, this absence of replication at the patient-level limits our
558 understanding of the unbiased real-world incidence of each mechanism, particularly those that occur
559 more rarely. Secondly, bacterial population sampling per patient, especially prior to treatment, is very
560 rarely adequate to confidently rule out alternate mechanisms, and it is therefore possible that reports
561 of spontaneous mutations may instead be due to pre-existing resistance present at too low frequency
562 to be sampled. Thirdly, clinical studies often lack control groups, and finally, lack access to patient
563 metadata, particularly previous antimicrobial therapy, limiting our ability to understand what drives

564 variation in mechanisms of AMR emergence between patients. Together these limitations make
565 identifying and distinguishing between eco-evolutionary mechanisms challenging. Many of these
566 limitations could be overcome by integrating eco-evolutionary studies within future clinical trials of
567 antimicrobial treatments (Box S1). Although such clinical trials often measure changes in pathogen
568 load, it is much less common that resistance emergence or the associated genetic mechanisms are
569 measured (Box 2). Whilst adding such analyses universally is unlikely to be cost-effective, routine
570 biobanking of patient samples derived from antimicrobial clinical trials, as outlined in Fig. 3B,
571 alongside consent for their re-use for eco-evolutionary analysis may provide an economical solution
572 that would transform our understanding of AMR emergence.

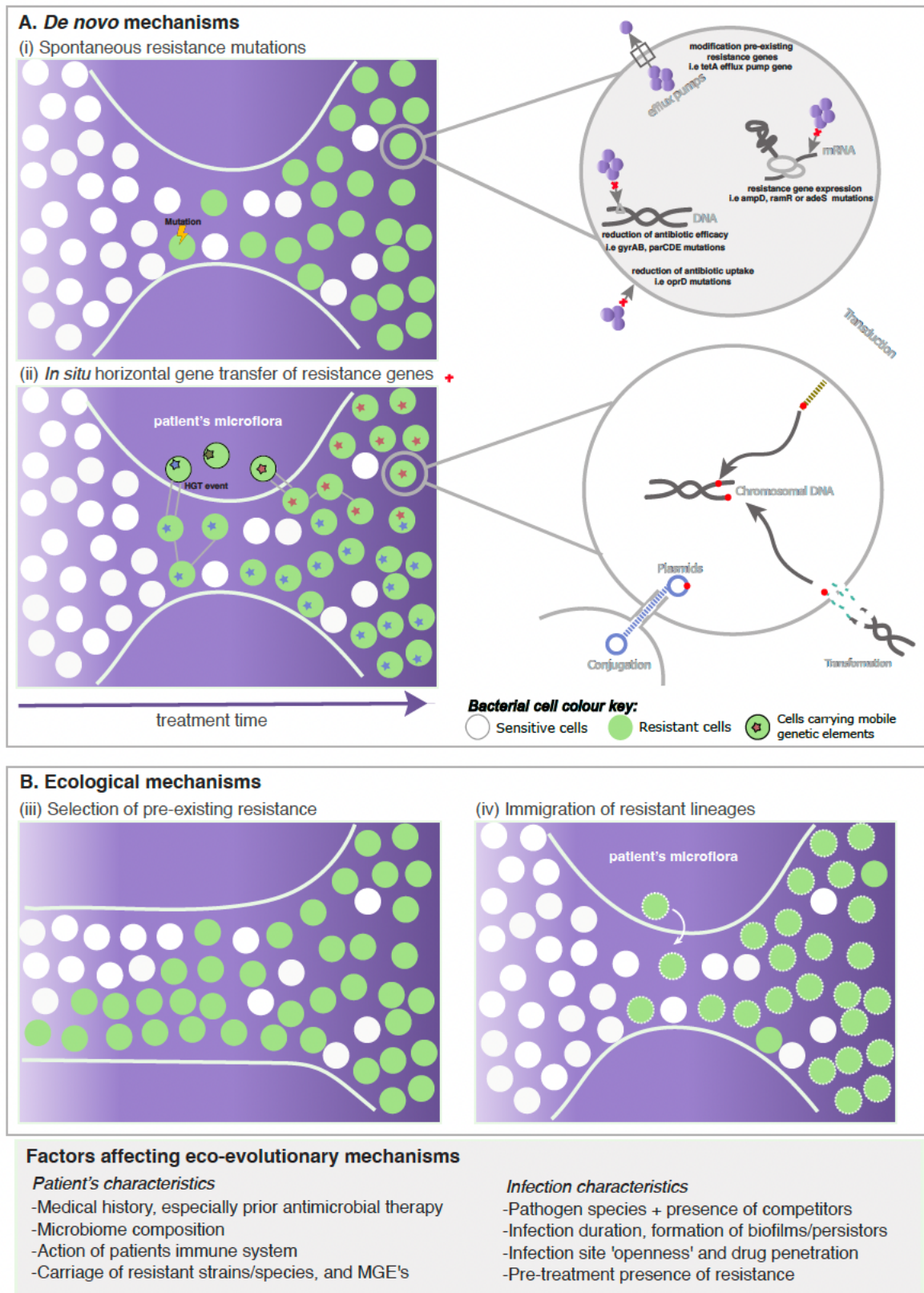
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574 Within-patient emergence of AMR is a significant clinical issue, and better understanding of the eco-
575 evolutionary mechanisms through which emergence can occur will aid future therapeutic design and
576 development. Our synthesis of the clinical literature reveals four eco-evolutionary mechanisms, each
577 of which operates within human infections, but which vary in their importance among pathogens and
578 between body sites and infection types. Despite this, significant gaps remain in our understanding,
579 particularly around the frequency and importance of each eco-evolutionary mechanism in particular
580 clinical settings. Better understanding of this would help to explain why resistance evolves in some
581 scenarios and patients but not others. Eco-evolutionary theory provides a framework to understand
582 and predict this variation in treatment outcome. Predicting within-patient AMR emergence on the
583 basis of a deep mechanistic understanding of the eco-evolutionary processes has enormous potential
584 to guide improved treatment decisions that limit AMR, extending the longevity of existing and new
585 antimicrobial drugs. Such a personalised approach to treating individual infections, guided by eco-
586 evolutionary principles could in future pave the way for the development of more robust and durable
587 antimicrobial therapeutics, ultimately benefiting global health and improving patient outcomes.
588 Translation of this body of eco-evolutionary theory requires that we embed this perspective within
589 clinical studies and clinical trials. Achieving this will require the involvement of regulators and
590 pharmaceutical companies to prioritise the study of AMR in clinical trials through improved trial
591 designs and biobanking of samples for re-use in eco-evolutionary studies.

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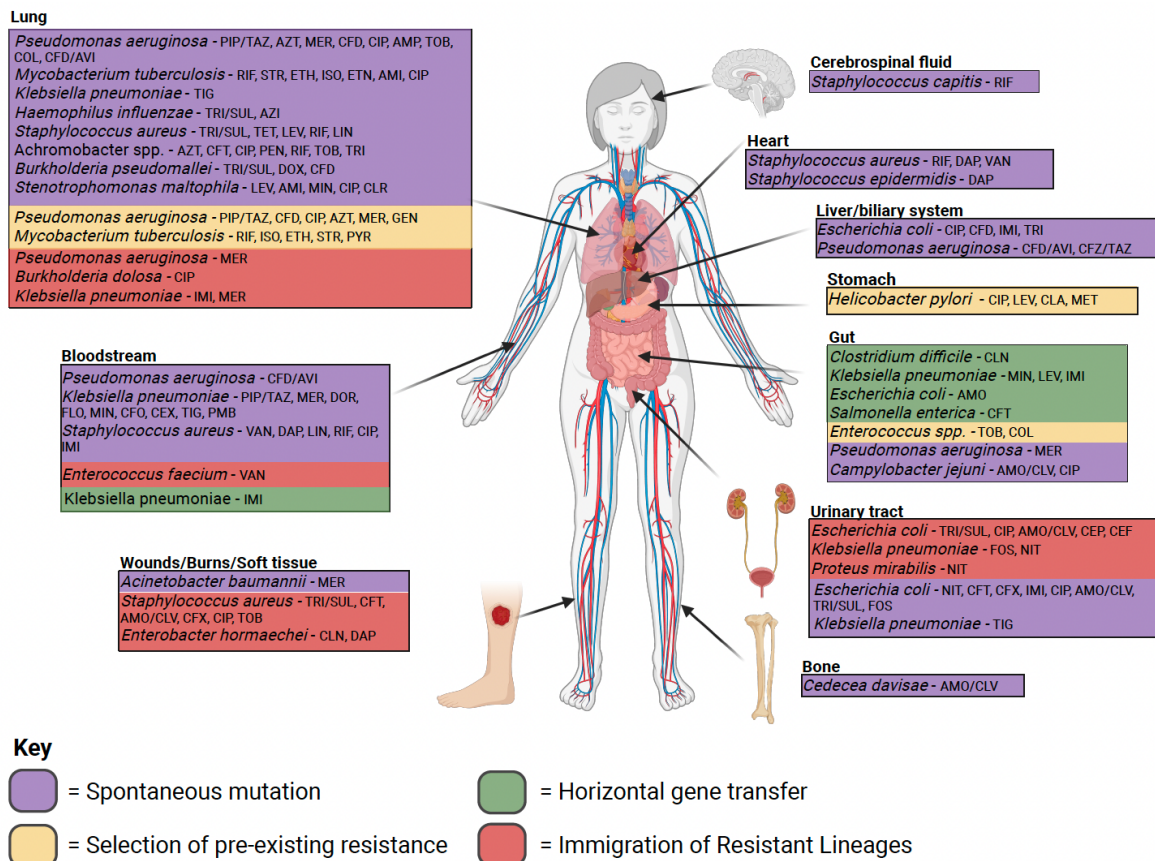
594 **END OF MAIN TEXT**



595

596 **Figure 1: Four eco-evolutionary mechanisms of within-patient emergence of antimicrobial**
 597 **resistance.** A) Resistance emergence through *de novo* evolutionary mechanisms: spontaneous
 598 mutation and horizontal gene transfer. Upon commencing antimicrobial treatment, the infecting
 599 bacterial population will decline (indicated in the left hand panels by the number of sensitive cells

600 (white) between the white borders, with treatment time running left to right) due to the negative effect
601 of therapy and/or the immune system on bacterial growth. Spontaneous mutations arise continuously
602 at random within a bacterial population, and if a mutation occurs that reduces susceptibility to the
603 antibiotic treatment, these nascent resistant cells (green) will then increase in frequency along with an
604 expansion of the infection population due to escape from the effects of therapy. These cells will have
605 gained a resistance determinant - common mechanistic bases of these are indicated within the
606 highlighted cells on the right hand side. Spontaneous mutations will typically act through i) reducing
607 antibiotic efficacy, for example by reducing drug-binding or drug uptake ii) regulatory mutations
608 affecting resistance gene expression or the activity of resistance determinants for example efflux
609 pumps, and iii) modifying pre-existing resistance genes to enhance or modify the spectrum or level of
610 resistance they confer, for example modification to drug-inactivating enzymes or export pumps.
611 Spontaneous mutations often incur fitness costs negatively impacting growth of resistant cells in
612 laboratory conditions, which may affect their survival within patients. Horizontal transfer of
613 resistance genes within-patients occurs through three key mechanisms - conjugal transfer of plasmids,
614 bacteriophage transduction, or uptake of DNA from the cell's environment (natural competence). B)
615 Ecological mechanisms of resistance emergence. Selection of pre-existing resistance will occur
616 immediately upon start of treatment, and may reduce the impact of treatment on bacterial population
617 size. Resistant cells (green) will increase in frequency as sensitive cells (white) decline in frequency
618 as treatment progresses, and the infecting population will continue to expand, escaping the inhibitory
619 effect of the drug. Pre-existing resistant cells may be stably present at an appreciable frequency within
620 an infecting population due to prior treatment with the same antibiotic, and due to their longer
621 evolutionary history may already have undergone compensatory evolution to reduce fitness costs of
622 resistance. In the case of immigration of resistant lineages, a resistant strain or species will be
623 transferred to the infection site during therapy, which may occur from the host microflora or from
624 elsewhere. This may occur at any time, however as the infection is cleared by the antimicrobial
625 treatment, this reduces competition for an invading bacterial strain or species, which may aid its
626 establishment at the infection site. The resistant lineage is then selected for, and the infecting
627 population of this lineage may expand. Factors that can affect the probability and action of these eco-
628 evolutionary mechanisms of within-patient AMR emergence are listed beneath the panels, broken
629 down into patient and pathogen. A patient's medical history (in particular prior treatment with
630 antimicrobials), and the nature of their infection will significantly impact likelihood of these eco-
631 evolutionary modes of AMR emergence.
632



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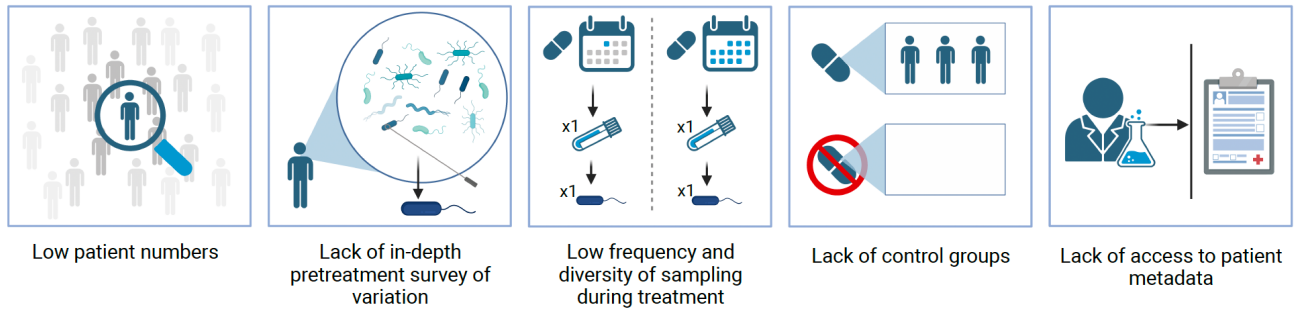
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635 **Figure 2: Reports of within-patient emergence of antimicrobial resistance across eco-**
 636 **evolutionary modes, body sites, organisms, and antibiotics.** Studies of within-patient AMR
 637 emergence included in this review cover a wide range of body sites, antimicrobials and bacterial
 638 pathogens. Reports of spontaneous resistance evolution are the largest single group, and the lungs are
 639 a site where resistance emergence is most frequently documented. Reports for some host niches are
 640 dominated by particular eco-evolutionary mechanisms, for example HGT in the gut, and spontaneous
 641 mutation in more isolated infection sites including bone, cerebrospinal fluid, heart and liver
 642 infections. It should be noted however that the distributions of eco-evolutionary modes, sites and
 643 pathogens may reflect biases due to ease of study, methodologies used, or other factors.

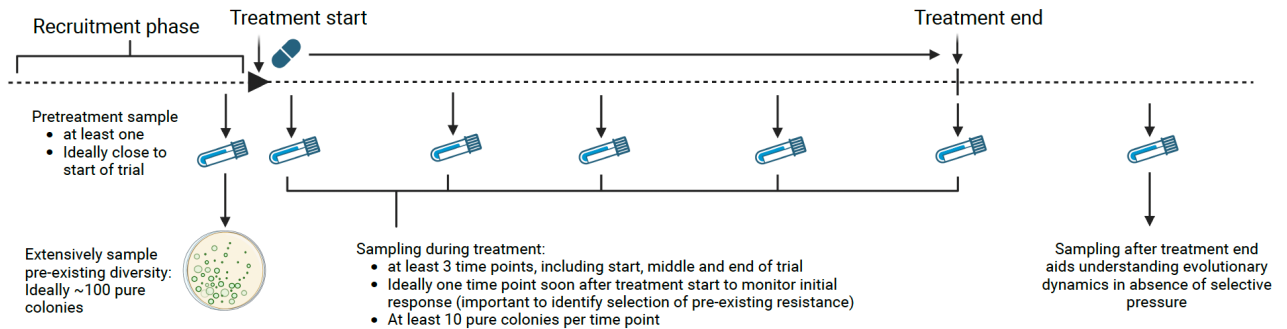
644 Antibiotics to which resistance evolved within-patient are detailed by 3 letter codes as follows: AMP =
 645 Ampicillin, AMI = Amikacin, AMO or AMO/CLV = Amoxicillin or Amoxicillin-clavulanate, AZI = Azithromycin, AZT =
 646 Aztreonam, CEF - Cefuroxime axetil, CEP = Cephalexin, CEX = Cefotaxime, CFO = Cefazolin, CFD or CFD/AVI-
 647 Ceftazidime or Ceftazidime-avibactam, CFZ/TAZ = Ceftolozane-tazobactam, CFX = Cefixime, CFT = Ceftriaxone, CIP =
 648 Ciprofloxacin, CLA = Clarithromycin, CLR = Chloramphenicol, CLN = Clindamycin, COL = Colistin, DAP = Daptomycin,
 649 DOX = Doxycycline, DOR = Doripenem, ETH = Ethambutol, ETN = Ethionamide, FLO = Flomoxef, FOS = Fosfomicin,
 650 GEN = Gentamicin, IMI = Imipenem, ISO = Isoniazid, LEV = Levofloxacin, LIN = Linezolid, MET = Metronidazole, MIN
 651 = Minocycline, NIT = Nitrofurantoin, PEN = Penicillin, PIP/TAZ = Piperacillin-tazobactam, PMB = Polymixin B, PYR =
 652 Pyrazinamide, RIF = Rifampicin, STR = Streptomycin, TET = Tetracycline, TIG = Tigecycline, TOB = Tobramycin,
 653 TRI/SUL = Trimethoprim-sulfamethoxazole, VAN = Vancomycin.

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A Common limitations of investigations of within-patient AMR evolution



B Addressing limitations through study of AMR evolution during clinical trials of antimicrobials



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Figure 3: Addressing limitations of within-patient AMR emergence studies. A) Five key factors limit the power of studies investigating within-patient AMR emergence. Low patient numbers are a common feature of the literature on within-patient AMR, with single-patient case studies predominating. These are often unusual cases or those where treatment has failed, and can subsequently lack generalisability to the wider patient population. There is also a lack of sampling frequency and depth both prior to and during treatment. This risks missing key population and evolutionary dynamics during the course of infection, and can make it difficult to identify ecological eco-evolutionary mechanisms - particularly pre-existing resistance. Studies of within-patient resistance emergence typically also lack control groups - with good reason, as refusing safe and effective treatment would be unethical. However a lack of controls can render the disentanglement of adaptation to the host and adaptation to the antimicrobial more challenging. Finally, access to patient metadata which can provide important context can often be limited, for example information on prior antibiotic treatment history. B) How sampling regimens can be structured during a clinical trial of an antimicrobial to maximise power for investigating within-patient AMR emergence. Where possible, samples from the infection site taken prior to and during treatment should be bio-banked to facilitate later investigations. These investigations should include extensive surveying of bacterial diversity

675 from pre-treatment samples, which is essential for later identification of cases involving pre-existing
676 resistance as a mechanism. For samples taken during treatment, a sufficient number of bacterial
677 isolates should be investigated at each time point for evolutionary dynamics to be identified, as
678 identifying the timing of any *de novo* resistance or immigration of resistant lineages may provide key
679 insight to driving factors. Collection of samples after treatment can also provide useful insight,
680 particularly for chronic infections which may not have been cleared, but also for understanding the
681 aftereffects of antimicrobial therapy on host microflora at the infection site. We additionally provide
682 and in-depth discussion of these limitations and how they could be addressed in supplementary box
683 S1.

684

685 **Box 1: Towards predicting within-patient AMR emergence**

686

687 **Probabilities of AMR evolving within-patient**

688 The four eco-evolutionary mechanisms reviewed here - spontaneous mutation, selection of pre-
689 existing resistance, immigration of resistant lineages, and horizontal gene transfer - fundamentally
690 differ from one another through the process by which resistance is initially generated. Once a resistant
691 bacterial lineage is generated within an infection, it will be selected for by the relevant antimicrobial
692 therapy applied and sweep to high frequency or fixation¹⁹. The eco-evolutionary mode that generated
693 resistance will affect the probability that such resistance becomes established within-patient. This is
694 primarily through both the frequency and the prevalence at which resistant lineages originate within
695 the infection. Spontaneous mutation and horizontal gene transfer can both in theory begin with the
696 generation of a single resistant cell. This generates a degree of randomness as to whether the nascent
697 resistant cell survives or is killed before it can propagate itself, which will be affected by the
698 frequency of spontaneous mutant generation or horizontal transfer within the infection¹⁸⁶. In contrast
699 is resistance evolution from pre-existing resistant lineages, or immigration of such lineages from
700 elsewhere. These mechanisms are far more likely to feature a larger population of resistant cells
701 which may accelerate resistance evolution⁹⁸, particularly as systemic antimicrobial therapy may
702 elevate the prevalence of resistant lineages in host microflora and reservoirs such as the gut¹⁸⁷,
703 making transfer of resistant lineages from host niches to the infection site more probable.

704

705 **Predicting within-host AMR evolution**

706 Predicting within-host emergence of AMR will require a variety of data on the patient and infection in
707 question. To predict the likelihood of pre-existing resistance or immigration of resistant lineages
708 occurring, an excellent starting point is information regarding the treatment history and past infections
709 of the same type that the patient may have experienced. Alongside more traditional microbiological
710 techniques and antibiotic susceptibility testing of isolates, this can be used to estimate an individuals

711 risk of resistance through these two mechanisms. Indeed, Stracy *et al.*, trained machine-learning
712 models on such data, and found such a model could reduce the predicted risk of resistance emergence
713 in UTIs (which occurs primarily through strain invasion) by 70%¹⁰⁸. More challenging, are predicting
714 likelihoods of spontaneous mutation and HGT in generating AMR within-patients. Mutational
715 frequency, epistasis, and fitness cost trade-offs can each vary considerably between specific resistance
716 genes^{21,129} and resistance-transferring mobile genetic elements^{188,189}. Despite recent advances in
717 predicting HGT of resistance at a population level¹⁹⁰, the ability to predict risk of HGT mediated
718 resistance evolution within a specific patient remains out of reach, as does prediction of a particular
719 spontaneous mutation occurring. Efforts to do so will require complex knowledge of the bacterial
720 fitness landscape within patients in the context of resistance¹⁹¹, alongside mutational probabilities,
721 which can vary across bacterial chromosomes¹⁹² and with a variety of environmental and population
722 factors¹⁹³.

723

724 **Box 2: Clinical trials of antimicrobials typically lack reporting on in-depth investigation of** 725 **AMR development**

726

727 To understand previous investigations in determining microbial response and AMR trials, we
728 reviewed antibiotic clinical trials registered in the WHO International Clinical Trials Registry
729 Platform (ICTRP) related to a range of body sites (The list of clinical trials and searching and binning
730 methods are shown in Supplementary table 1). We found that many trials of antimicrobials did not
731 report microbial responses (consisting of bacterial load and/or resistance related measures) in the
732 primary or secondary outcomes registered on the platform. Respiratory tract and bloodstream
733 infection trials did not report microbial responses for 85% and 82% of reports respectively, despite
734 these two infection sites having the highest mortality statistics associated with and attributed to AMR
735 in 2019². Only 8% of respiratory infection trials and 6% of bloodstream infection trials of antibiotics
736 reported AMR as an outcome. In contrast, trials of treatments for urinary tract infections and gut
737 infections reported microbial response as an outcome in 64% and 47% of cases, and documented
738 AMR outcome in 15% and 20% of cases respectively.

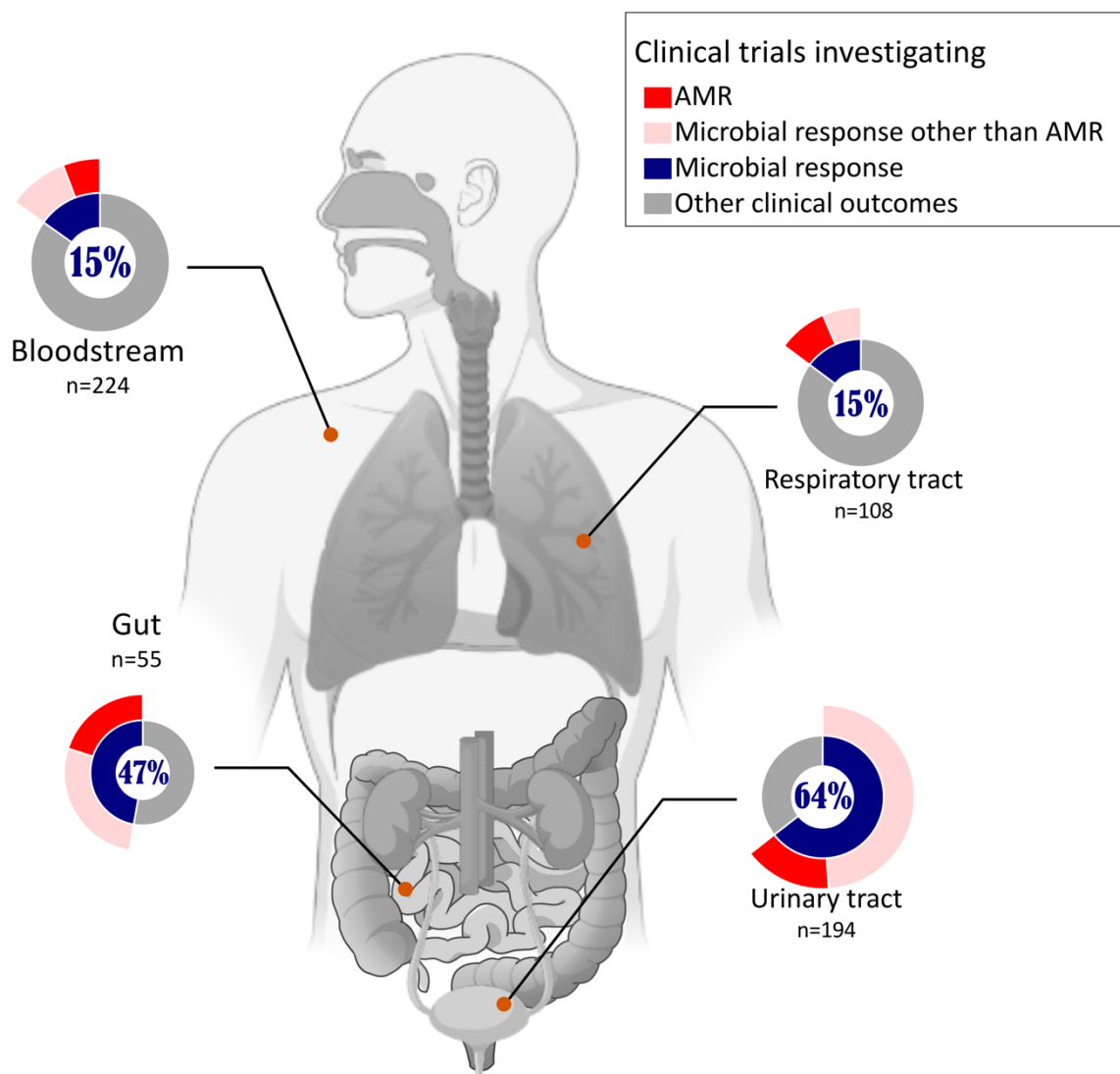
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740 **Clinical trials of antimicrobials are underutilised resources for the study of AMR evolution**

741

742 Despite the numerous advantages that the clinical trial setting offers over clinical case-study reports
743 for investigating within-patient AMR emergence, our analysis of clinical trial registered outcome
744 measures indicated that resistance-related outcomes are rarely included. In this review, we suggest
745 that such trials of antimicrobials should accommodate investigations of within-patient AMR
746 emergence. As of December 2022, there are 62 novel antimicrobial therapies in the clinical trial

747 pipeline¹⁹⁴, for which no information is known about risks and likelihood of within-patient resistance
 748 evolution. Incorporating or spinning-off evolutionary investigations from clinical trials would allow
 749 information to be gathered on the risks of within-patient AMR emergence early for these novel
 750 antimicrobials and could support clinicians to develop initial guidelines for using these drugs in a
 751 manner that reduces risk of resistance. This would allow a pre-emptive strategy aimed at preventing
 752 AMR emergence in the clinic for new antimicrobials coming into use, alongside a more personalised
 753 approach to antimicrobial chemotherapy for existing therapeutics.
 754
 755



756
 757 **Exploration of microbial and AMR investigation reporting during clinical trials of**
 758 **antimicrobials:** Proportions of surveyed registered clinical trials of antimicrobials (n=581) across
 759 four key infection sites, that list microbiological response as an reported primary or secondary

760 outcome (inner circles of plots, blue = yes, grey = no), and proportion of those that list AMR as a
761 reported outcome (outer circles of plots, dark red = yes, light red = no).

762

763

764

765

766 **Glossary:**

767

768 **Antibiotic/antimicrobial** – Agents used to kill or inhibit the growth of microorganisms such as
769 bacteria, fungi, viruses and parasites. Antibiotics refer to agents utilised against bacteria and can be
770 classified as bactericidal (killing bacteria) or bacteriostatic (inhibiting bacterial growth).

771

772 **Antimicrobial resistance (AMR)** - The ability of microorganisms, such as bacteria, viruses, fungi, or
773 parasites, to withstand the effects of drugs that would normally inhibit or kill them. This phenomenon
774 occurs when these organisms adapt and develop resistance mechanisms against antimicrobial agents,
775 rendering previously effective treatments ineffective. Clinically, AMR is defined as when the
776 minimum inhibitor concentration (MIC) of the antimicrobial required to halt growth of a bacterium
777 exceeds the clinical breakpoint – the highest concentration of that antimicrobial that can be given to a
778 patient.

779

780 **Antimicrobial tolerance** – The ability of a population of micro-organisms to survive a transient
781 exposure to a microbicidal agent. Differs from resistance in that the agent remains effective against
782 the microbe as measured by MIC but requires a more prolonged treatment in order to successfully
783 eliminate the infection.

784

785 **Bacterial persistence** – When a subpopulation of bacteria has a much higher tolerance to an
786 antibiotic than the majority, that population described as persistent. When the pressure of the
787 antibiotic is removed, this persistent community can re-emerge, creating a situation of recurrent
788 infection despite antibiotic treatment.

789

790 **Cross-resistance** - Antimicrobial resistance that evolves through adaptation to another antimicrobial.
791 This can occur when evolution of resistance to one antimicrobial confers resistance to another,
792 typically due to a resistance mechanism that equally affects the action of all drugs within a class or
793 has a non-specific mechanism such as multi-drug efflux pumps.

794

795 **Collateral sensitivity** - A situation where gain of resistance to one antimicrobial, results in increased
796 sensitivity to another. This is a type of fitness trade-off.

797

798 **Eco-evolutionary dynamics** - The reciprocal interactions and feedback between ecological processes
799 and evolutionary changes in populations over short time scales. Ecological shifts promote adaptation
800 of populations to their changing environments, and the resulting evolutionary changes can in turn
801 shape ecological interactions. In the context of infections – the within-patient niche ecology will
802 shape the evolution of antibiotic resistance, which can then affect the ecology of the infection itself
803 through failure to clear the infection, disease progression and loss of sensitive strains and microflora.

804

805 **Fitness** – A measure of reproductive success – the ability of an individual or population with the same
806 genotype to survive, reproduce, and contribute that genotype to the next generation.

807

808 **Fitness trade-off** – A compromise in which the fitness advantage conferred by one trait comes at the
809 expense of reducing the fitness effect of another trait. For example, resistance to an antibiotic may
810 reduce growth rate when the antibiotic is absent.

811

812 **Fixation** – The state in which a genetic variant becomes the only variant present for that specific
813 locus. All individuals within the population share that same allele.

814

815 **Genetic drift** - the change in frequency of an allele due to random chance. The impact of such
816 random effects is stronger at smaller population sizes. Newly generated random mutations present at
817 very low frequency must escape random loss due to genetic drift even if they provide a benefit before
818 becoming established at a higher frequency in the population.

819

820 **Horizontal gene transfer (HGT)** – The process by which microorganisms may exchange genetic
821 material that bypasses vertical transmission from parent to offspring.

822

823 **Host microenvironment** - The specific localised conditions and factors within a host organism such
824 as its physical properties, local immune response, and neighbouring microbial communities, that
825 influence the interactions between the host and its resident microorganisms at that site.

826

827 **Host microflora** – The community of microbes that reside in and on a host organism.

828

829 **Hypermutators** - A microbial strain with an unusually high mutation rate, often caused by
830 deficiencies in DNA repair mechanisms. Under selection pressure from an antimicrobial agent, this
831 rapid accumulation of spontaneous genetic mutations may accelerate the process of selection and thus
832 evolution of AMR by increasing the likelihood of a mutation conferring resistance to occur.

833

834 **Immigration of resistant lineages** – Also referred to as strain invasion or replacement. The
835 successful establishment of a novel microbial strain or species within a specific host environment.
836 Certain strains possess traits that allow them to better adapt to the environment than pre-existing
837 strains, and thus they are selected for and become the dominant strain. These novel invasive strains
838 may derive from the host microflora or from the environment.

839

840 **Population bottleneck** – A sharp reduction in the size of a population usually due to a detrimental
841 change in the environment. This significantly reduces genetic diversity and exaggerates the effect of
842 genetic drift. In the context of bacterial infections, bottlenecks will be induced by the action of the
843 immune system and antimicrobial therapy.

844

845 **Pre-existing resistance** – The presence of inherent or acquired antibiotic resistance genes within the
846 infection population and/or microbiome prior to treatment with an antimicrobial.

847

848 **Protected niche** – A host microenvironment that offers microorganisms protection from external
849 stresses, such as the host's immune system, antimicrobial therapy, competition from the host's
850 microflora, or other adverse environmental conditions. Niches can promote the establishment of
851 persistent microbial communities and may offer a safe haven for resistance evolution to occur within.

852

853 **Selection** – The process by which genetic variations within a population become more prevalent due
854 to conferring traits that influence the fitness of the organisms in their environment. In the context of
855 antimicrobial therapy, selection refers to the survival and growth of resistant strains and the loss of
856 sensitive ones during treatment.

857

858 **Selective Sweep** – An evolutionary event where a highly advantageous mutation rapidly increases in
859 frequency due to strong positive selective pressure. As it does, the genetic diversity in the region of
860 the mutation decreases, creating a detectable signature of reduced allele frequency.

861

862 **Spontaneous mutation** – Also known as *de novo* mutation. Heritable alterations in the genome of a
863 microbe that arise spontaneously during replication or repair, and were not previously present in the
864 population or acquired from external sources of genetic material.

865

866 **Supplementary Box S1 - Critical gaps in knowledge**

867

868 **Limited numbers of patients**

869 The majority of the current literature are case reports of individual patients. These vary widely in their
870 methodology and research focus, and are rarely designed for studying evolutionary dynamics.

871 Additionally, the available literature has a strong bias towards particular eco-evolutionary modes of
872 AMR emergence, accessible host niches, and a handful of well-studied pathogens. Spontaneous
873 resistance evolution is frequently studied as it can feasibly be evidenced from relatively low numbers
874 of bacterial isolates, compared with the other eco-evolutionary modes, which require more extensive
875 sampling of pathogen populations and/or the wider microbiota. Infections of the lungs and guts are
876 commonly studied, likely due to the ease of the non-invasive sampling methods. Also, well-
877 represented are chronic infections, where repeat sampling of individual patients is more probable.
878 This bias, in turn, leads to an over-representation in the literature by pathogens common in these kinds
879 of infections, including *P. aeruginosa*, *K. pneumoniae*, and *Staphylococcus* spp. There have only been
880 very few larger-scale studies of within-patient AMR emergence that are designed to distinguish the
881 eco-evolutionary mode of AMR emergence and involve more than ~30 patients. These target a limited
882 range of infection contexts, including urinary tract and wound infections¹⁰⁸, lung infections^{27,98,104}
883 and bloodstream infections⁴⁰).

884

885 **Limited pathogen sampling**

886 Low numbers of sampled bacterial colonies limits the resolution of many studies, and as such their
887 power to distinguish between eco-evolutionary modes of resistance emergence. Many case studies
888 report either a single or a small number of colonies, with the majority of studies focusing their work
889 on <10 bacterial colonies isolated from patient samples^{31,32,35,41,42,45,46,51–53,62,81,153}. This can limit the
890 potential to capture pre-existing variation in resistance, and is likely to particularly limit the ability of
891 studies to observe more complex or mixed modes of AMR emergence. For example, in a
892 mathematical model, >38 samples were found to be required to accurately identify clonal lineages in
893 a tumour with high mutational diversity¹⁹⁵. More intensive sampling is unlikely to be feasible in all
894 studies due to clinical considerations, but where possible should aim to gain a random sample of
895 colonies such that this provides a sufficiently representative sample of genetic variation present in the
896 infection.

897

898 Many studies lack samples collected prior to commencing antimicrobial treatment<sup>43,47,48,51,53–
899 55,81,115,190</sup>, but such pre-exposure samples are essential for determining the key features of the eco-
900 evolutionary dynamics of AMR. For example, assessing whether resistant lineages were present at
901 appreciable frequencies prior to treatment is required to distinguish spontaneous resistance mutation
902 from other eco-evolutionary modes. This is evident from a study of acute *P. aeruginosa* lung
903 infections where pre-existing resistance was present at 7-8% of the population²⁷, but would have
904 likely been misclassified as spontaneous mutation without pre-treatment sampling. Pre-treatment
905 sampling is unlikely to be possible in all cases, and indeed may be impossible for critical cases where
906 immediate antibiotic treatment is required, but could feasibly be added where infections are not life-

907 threatening and/or before a new antibiotic is used against a chronic infection. Adding pre-treatment
908 sampling would greatly improve understanding of eco-evolutionary mechanisms and thus facilitate
909 AMR prediction and improve treatments.

910

911 The nature of clinical samples themselves can also pose a challenge. A clinical sample is effectively a
912 snapshot of the infecting population and therefore may not perfectly represent the population as a
913 whole. In many scenarios, this bias is unavoidable. However, partial sampling of the population
914 combined with low numbers of colonies may commonly exclude pre-existing resistant lineages,
915 potentially leading us to overestimate the role of spontaneous mutation. If indeed spontaneous
916 resistance mutation is less common than expected, this may limit the potential of adjuvant therapies
917 targeting spontaneous mutagenesis to prevent AMR evolution¹⁹⁶. Although there are trade-offs
918 associated with sampling intensity and clinical considerations imposing limits on what is reasonable,
919 frequent in-depth sampling of populations before, during and after treatment, would provide enhanced
920 understanding of eco-evolutionary mechanisms at work within patients.

921

922 **Lack of control groups and patient metadata**

923 Existing studies on within-patient AMR emergence are predominantly lacking a control group. There
924 are obvious good reasons for this, primarily it being unethical to refuse treatment. However, it can
925 then be challenging to establish causality of the antibiotic treatment in driving the observed
926 evolutionary change without comparison to control groups. For example, adaptations to the host
927 environment can be conflated as adaptations to treatment: host antimicrobial peptides can select for
928 enhanced antibiotic resistance¹⁹⁷ and adaptation to macrophages *in vitro* has been shown to increase
929 *E. coli* resistance to aminoglycosides and fitness in murine infection models¹⁹⁸. Nonetheless, often
930 clinical trials compare newer treatments to existing standards of care in a “non-inferiority” design,
931 where comparisons between two different treatment groups are possible. Here, even though there are
932 likely to be differences in bacterial targets and pharmacological properties between treatment arms of
933 the clinical trial, this design is powerful in enabling causal treatment-specific evolutionary responses
934 to be identified¹⁹⁹, partly overcoming the lack of untreated controls.

935

936 Lacking patient metadata may also complicate eco-evolutionary interpretation of clinical studies and
937 limit our understanding of the drivers of variation in outcomes among patients. A key factor is likely
938 to be antibiotic treatment history, however accessing such records can be challenging. Patients
939 previously exposed to higher levels of antibiotics are at greater risk of resistance emergence, in part
940 because their microflora is more likely to contain antibiotic resistant lineages²⁰⁰, potentiating selection
941 of pre-existing resistance or immigration of resistant strains¹⁰⁸. Additionally, individuals treated with
942 antibiotics could be receiving non-antibiotic treatments for a variety of other underlying health

943 conditions, which can impact AMR. For example, anti-inflammatory drugs (ibuprofen, naproxen,
944 diclofenac) and β -blockers increase conjugative transfer rates of plasmids²⁰¹.

945

946 Lacking patient metadata is especially problematic for individual case reports, since these represent a
947 biased sample of infections, typically including cases where treatment(s) failed, cases of unusual
948 infections, or cases of unusual clinical outcomes²⁰². While these unusual cases can provide important
949 clinical experience for future treatment approaches, lacking patient metadata limits our ability to
950 understand what features of these infections predicated them towards AMR emergence versus the
951 majority of successful treatments. A wide variety of other host factors could influence AMR
952 emergence, including the immune response, microflora, infection microenvironment, and treatment
953 history, along with patient demographics and lifestyle factors. Without such metadata it will be
954 challenging to link host and/or clinical characteristics to risks of particular eco-evolutionary modes to
955 understand how better to target antibiotic treatments and other interventions to limit AMR emergence.

956

957 **Making better use of clinical trials to understand eco-evolutionary mechanisms of AMR**

958 Phase-III randomised controlled trials (RCT) for antimicrobials could be a powerful framework to
959 study the eco-evolutionary modes of AMR emergence, solving several of the issues raised in this
960 section. Typically, such RCTs include sufficiently large numbers of patients for statistical power and
961 these patients are carefully selected to control for other sources of variation. Moreover, extensive
962 patient metadata are usually collected. Multiple treatment groups are included enabling their
963 comparison, and in some cases RCTs even include placebo groups that do not receive antimicrobial
964 treatment. Indeed, RCTs have been used previously to infer the impact of treatment on AMR: Meta-
965 analyses of clinical trial data for long-term azithromycin²⁰³ and inhaled antibiotics in chronic lung
966 infection²⁰⁴ treatment show increased risk of resistance emergence in the treatment arms.

967 Furthermore, longer durations and/or multiple-courses of antibiotics are associated with higher rates
968 of resistance emergence^{205,206}. In non-inferiority trials, increased resistance in both test treatment and
969 next-best available therapy groups have been identified, primarily through phenotypic antimicrobial
970 susceptibility testing²⁰⁷⁻²¹¹, with some trials also including genetic causes of resistance^{212,213}.

971 However, although microbiological outcomes, such as pathogen abundance are often included in RCT
972 designs, quantification of antibiotic resistance levels or genetic characterisation of resistance
973 mutations is much less common (main text Box 2). Whilst adding such components to RCTs routinely
974 may not be possible owing to limitations of costs and facilities, biobanking RCT samples for re-use
975 can provide a practical, low-cost route to enhancing the utility of RCTs. For example, re-use of
976 samples from a non-inferiority trial of ceftolozane/tazobactam versus meropenem for treating *P.*
977 *aeruginosa* respiratory infections revealed a greater contribution of spontaneous mutation to AMR in

978 the meropenem treated patients (22% of patients compared to 0% in ceftolozane/tazobactam arm), but
979 similar rates of strain immigration in both groups (~4% of patients)²¹³.

980

981 How then could RCT protocols be augmented to enable identification of the eco-evolutionary
982 mechanisms of AMR? Distinguishing spontaneous mutation from pre-existing resistance requires
983 samples to be taken before antibiotic treatment starts and for these pathogen populations to be
984 sampled at sufficient depth to capture low-frequency pre-existing resistant variants. Tracking the
985 dynamics of selection then requires dense longitudinal sampling of population diversity from multiple
986 subsequent timepoints, ideally covering both during treatment and after cessation of treatment. For all
987 these samples, testing multiple randomly selected bacterial colonies is crucial to gain an unbiased
988 estimate of the genetic and phenotypic diversity present within the population. Sampling higher
989 numbers of colonies will give a more accurate picture of the population, but after a certain point there
990 are diminishing returns to sampling more colonies and use of power analyses can help to gauge the
991 appropriate depth of sampling for the predicted level of genetic diversity (e.g., this is likely to be
992 higher for chronic than acute infections, necessitating deeper sampling of the former). Attributing the
993 source of immigrating resistant lineages or donors of resistance encoding MGEs requires analysis of
994 the wider patient microbiota using metagenomics or selective culturing. Finally, understanding why
995 patient infections vary in the eco-evolutionary mode of AMR emergence requires integration of these
996 microbiological data with patient metadata, ideally including measures of host immune responses and
997 their infection microenvironments. Although it is unlikely that all of these features can be included in
998 every RCT, awareness of the samples required, collection of them during the trial (main text Fig. 3B)
999 and biobanking for re-use would ensure that we maximise the utility of RCTs and gain the greatest
1000 insight possible from these valuable in-human eco-evolutionary experiments.

1001

1002 It is important to acknowledge the likely barriers to improving RCT design. Phase-III RCTs are
1003 already extremely expensive and mostly funded by pharmaceutical companies themselves, such that
1004 additional work and costs are unlikely to be included unless mandated by regulators. Commitment
1005 from trial funders and regulators to prioritise AMR studies as a core component of RCTs will
1006 therefore be necessary. Adding costs, however, carries inherent risks, potentially discouraging drug
1007 candidates from advancing through the pipeline. A potential alternative would be to ensure that RCT
1008 designs collect and biobank the necessary samples along with gaining appropriate ethical approval for
1009 follow-up eco-evolutionary analysis, even if not studied as part of the initial project. This would
1010 facilitate sample and data suitability and accessibility, creating a valuable resource for microbiologists
1011 and eco-evolutionary biologists to study the eco-evolutionary mechanisms of AMR through re-use of
1012 such biobanked samples whilst adding only limited costs to the RCT.

1013

1014

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