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Multiple evolutionary pathways lead to vancomycin resistance in *Clostridioides difficile*

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1 Abstract

2 Clostridioides difficile is an important human pathogen, for which there are very limited 3 treatment options, primarily the glycopeptide antibiotic vancomycin. In recent years 4 vancomycin resistance has emerged as a serious problem in several Gram positive pathogens, 5 but high level resistance has yet to be reported for *C. difficile*, although it is not known if this 6 is due to constraints upon resistance evolution in this species. Here we show that resistance to vancomycin can evolve rapidly under ramping selection but is accompanied by severe 7 8 fitness costs and pleiotropic trade-offs, including sporulation defects that would be expected 9 to severely impact transmission. We identified two distinct pathways to resistance, both of 10 which are predicted to result in changes to the muropeptide terminal D-Ala-D-Ala that is the 11 primary target of vancomycin. One of these pathways involves a previously uncharacterised 12 D,D-carboxypeptidase, expression of which is controlled by a dedicated two-component signal transduction system. Our findings suggest that while *C. difficile* is capable of evolving 13 14 high-level vancomycin resistance, this outcome may be limited clinically due to pleiotropic 15 effects on key pathogenicity trains. Moreover, our data provide a mutational roadmap to 16 inform genomic surveillance.

17 Introduction

Clostridioides difficile is the most common cause of antibiotic-associated diarrhoea 18 19 worldwide, resulting in significant morbidity and mortality¹ that places a huge burden on healthcare systems^{2,3}. Most cases of nosocomial *C. difficile* infection (CDI) follow recent 20 21 antibiotic treatment, which, through altering microbial diversity in the colon, reduces microbiota-mediated colonisation resistance⁴. Although restoring microbial community 22 diversity through faecal microbiota transplantation (FMT) is a potential future treatment for 23 CDI⁵, current treatment relies on additional antibiotics, most commonly metronidazole or 24 vancomycin. While these can resolve the CDI, they further exacerbate damage to the 25 26 microbiota, leading to recurrent CDI in up to 25% of cases⁶. Due to increasing incidence of 27 resistance and consequently poor patient outcomes, use of metronidazole has declined in 28 recent years, and vancomycin is now the recommended front line antibiotic in the UK⁷.

Vancomycin is a glycopeptide antibiotic that binds to the terminal D-Ala-D-Ala 29 30 residues on peptidoglycan muropeptide precursors, sterically blocking transglycosylation and 31 transpeptidation reactions⁸. Although relatively slow to emerge, resistance to vancomycin is 32 now found globally in Staphylococcus aureus (VRSA) and is widespread in several 33 *Enterococcus* spp. (VRE)^{9,10}. Vancomycin resistance in Enterococci is usually associated with one of a number of *van* gene clusters which encode the enzymes that modify peptidoglycan 34 35 to remove the vancomycin binding site. Here, the D-Ala-D-Ala is replaced with either D-Ala-D-Lac (e.g. vanA, also seen in VRSA), conferring high level resistance, or D-Ala-D-Ser (e.g. 36 *vanG*), conferring low level resistance¹¹ Vancomycin susceptibility of *C. difficile* is not routinely 37 tested in clinical laboratories making monitoring the emergence of resistance extremely 38 challenging. However, the effectiveness of vancomycin against CDI has declined over time¹² 39 40 and multiple case reports show reduced vancomycin susceptibility in individual clinical isolates¹³. Although a complete *vanG* cluster is found in diverse *C. difficile* strains¹⁴, whether 41 this confers vancomycin resistance is unclear¹⁵. However, increased expression of the vanG 42 cluster does appear to be associated with reduced susceptibility to vancomycin. Moreover, 43 44 mutations in the vanSR-encoded two component system that reduce vancomycin 45 susceptibility through derepression of the *vanG* cluster, have been reported in both clinical isolates and laboratory evolution experiments^{16,17}. Beyond such regulatory changes of the 46 47 *vanG* cluster, the molecular mechanisms underpinning the evolution of vancomycin 48 resistance in *C. difficile* remain unknown. For example, we do not know if other mutations 49 occurring within the *vanG* cluster or at other loci in the *C. difficile* genome contribute to 50 increasing resistance observed clinically. Moreover, whether the evolution of vancomycin 51 resistance is associated with pleiotropic phenotypic effects or fitness costs in *C. difficile* is 52 poorly understood.

53 To understand the evolutionary dynamics and molecular mechanisms of vancomycin resistance in C. difficile we experimentally evolved ten replicate populations at increasing 54 concentrations of vancomycin. Within just 250 generations we observed the evolution of 16 55 56 to 32-fold increased vancomycin minimum inhibitory concentration (MIC). To identify the causal genetic variants, we genome sequenced both endpoint resistant clones and whole 57 58 populations at multiple time-points during the evolution experiment, and reintroduced key 59 mutations, observed in the evolved resistant lineages, into the wild-type ancestral genetic 60 background. Evolution of increased vancomycin resistance was associated with mutations in two distinct pathways, occurring either in *vanT* within the *vanG* cluster or in a gene encoding 61 62 a regulator of a previously uncharacterised D,D-carboxypeptidase. Mutations in either 63 pathway are predicted to modify the vancomycin target in the cell wall peptidoglycan and 64 both were associated with defects in growth and sporulation of varying severity. Together our 65 results propose a new mechanistic model for vancomycin resistance emergence in C. difficile, potentially expanding the genetic determinants of resistance that should be monitored in 66 67 clinical genomic epidemiology. Moreover, our data suggest that the initial emergence of vancomycin resistance in *C. difficile* in the clinic may be severely constrained by pleiotropic 68 fitness trade-offs with key transmission and virulence traits. 69

70 Results

Vancomycin resistance evolves rapidly in *C. difficile* during *in vitro* experimental evolution. 71 72 We first generated genetically barcoded ancestral strains in an avirulent background with 73 either wild-type or elevated mutation rates. Specifically, R20291, a clinically relevant ribotype 74 027 C. difficile strain, was rendered avirulent through complete deletion of 18 kb spanning 75 the entire PaLoc that includes the genes encoding both major toxins and associated 76 regulatory proteins, creating strain R20291 Δ PaLoc. A subsequent deletion, removing the 77 mutSL genes encoding a DNA-damage repair system, generated a hypermutable variant 78 R20291 Δ PaLoc Δ mutSL, with an approximately 20-fold higher mutation rate than the wild-79 type. Five distinct derivatives of each ancestral strain were then generated through 80 introduction of a 9-nucleotide barcode sequence downstream of the pyrE gene, resulting in 81 10 individually barcoded replicate lines used in the evolution experiment (R20291\Delta PaLoc 82 pyrE::barcode 1-5; R20291 Δ PaLoc Δ mutSL pyrE::barcode 7-11). Each of the 10 barcoded 83 strains was used to inoculate a 6-well plate containing media supplemented with vancomycin 84 at 0.25x, 0.5x, 1x, 2x, 4x, and 8x the initial MIC of 1 µg/ml. Populations were passaged every 48 h, whereby cells from the well with the highest antibiotic concentration permitting growth 85 86 were propagated in a 1:400 dilution to a fresh 6-well plate. This process was repeated for a total of 30 serial transfers per replicate line, with adjustment of the vancomycin gradient over 87 88 time as the growth-permitting vancomycin concentration in each evolving line increased. Ten corresponding control populations were propagated under equivalent conditions in the 89 90 absence of vancomycin. Populations underwent approximately 8.64 generations per transfer, 91 yielding approximately 259 generations throughout the course of the experiment.

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93 Resistance evolved rapidly in all 10 replicate populations propagated with vancomycin 94 selection (Fig. 1a). Nine of the ten replicate lines evolved to grow in 2 µg/mL vancomycin (an 95 apparent MIC of 4 μ g/mL, the EUCAST breakpoint) by the end of the second passage (P2) and all ten grew in the presence of 8 µg/mL (Bc2, 3, 4) or 16 (Bc1, 5, 7, 8, 9, 10, 11) µg/mL 96 97 vancomycin by P30 (Fig. 1a). This was significantly accelerated in the hyper-mutable replicate 98 lines (Fig. 1b). At least six individual clones were isolated from each evolved population at the end-point and their vancomycin MIC was determined. Of the 82 clones tested, 38 (46%) 99 100 had an MIC consistent with the vancomycin concentration permitting growth of the 101 population from which they were isolated, while the remainder had an MIC slightly lower

than expected [for 42 clones their MIC was 2-fold lower, whereas for 2 clones their MIC was
4-fold lower than the vancomycin concentration permitting growth of the population from
which they were isolated (Supplementary Data 1)], demonstrating that significant variation
existed within evolved populations by P30.

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107 Genetic bases of evolved vancomycin resistance

108 To understand the genetic bases of increased vancomycin resistance in individual end-point 109 clones, one clone per replicate population that had an MIC representative of their population was selected for whole genome sequencing. The parental strains for each barcoded lineage, 110 111 and one random endpoint clone from each replicate control population, were also 112 sequenced. For analyses we focused on mutations that were observed in the vancomycin 113 treated lines but never in ancestral or in control evolved clones because these are the most 114 likely to have evolved in response to vancomycin selection. Sequence variants unique to the vancomycin-evolved clones were identified using Varscan¹⁸, validated with Breseq and IGV¹⁹ 115 116 after mapping to the reference R20291 genome (Fig. 1b, Supplementary Data 2). We 117 observed between 1 and 3 unique mutations per genome in wild-type clones and between 118 14 and 26 in hypermutator clones from the vancomycin treatment. Within gene SNPs 119 accounted for 43% of the variants, of which 67.4% were nonsynonymous and 32.6% were 120 synonymous. Frameshifts accounted for a further 31% of identified unique mutations.

121

122 Parallel evolution, where mutations affecting the same locus arise in multiple independently 123 evolving replicate populations, is strong evidence for the action of selection and suggests a 124 potential role for these mutations in adaptation. In the 5 wild-type vancomycin selected lines 125 we observed parallel evolution occurring at three genomic loci: *vanT* in 3 clones, *CD3437* in 2 clones and *comR* in 2 clones. Whereas mutations in *vanT* and *CD3437* were mutually exclusive 126 127 in wild-type evolved clones, mutations in *comR* always co-occurred with mutations in *vanT*. VanT is a putative Serine racemase (mutations in Bc3-5) encoded within a VanG-type cluster¹⁴ 128 that was previously implicated in decreased vancomycin susceptibility in C. difficile¹⁶ 129 (Supplementary Data 2). CD3437 encodes a predicted two-component system histidine 130 kinase (mutations in Bc1 and 2), with its cognate response regulator encoded by CD3438. 131 132 These genes had not previously been implicated in vancomycin resistance, however the 133 nearby CD3439 encodes a putative D,D-carboxypeptidase that likely plays a role in 134 modification of peptidoglycan through removal of the stem peptide terminal D-Ala. Based on these predicted functions we propose renaming these genes *dacS* (*CD3437*, histidine kinase), 135 136 dacR (CD3438, response regulator) and dacJ (CD3439, D,D-carboxypeptidase). Consistent 137 with mutations at these loci playing key roles in vancomycin resistance, all five hypermutator replicate lines had mutations in the vanG operon (1 had a nonsynonymous mutation in vanT 138 139 and 1 in vanS) or dacS, dacR (encoding the cognate response regulator) and dacJ. 140 Interestingly, one strain (Bc8) had mutations in both *dacS* and *CD1523* (*vanS*), encoding a two-141 component system sensor histidine kinase that is thought to regulate the vanG operon in 142 response to vancomycin, suggesting that the two pathways to resistance are not entirely 143 mutually exclusive. comR encodes a homologue of the RNA degradosome component 144 PNPase, suggesting that RNA stability may play a role in the vanT-associated mechanism of 145 vancomycin resistance. Consistent with this possibility, in the third clone carrying a vanT 146 mutation we observed coexisting mutations affecting maa, encoding a putative maltose O-147 acetyltransferase, and a 75 bp deletion that completely removed an intergenic region 148 downstream of *rpmH* and before *rnpA*, encoding another predicted component of the RNA 149 degradosome. By contrast, in the 2 clones carrying mutations in *dacS* we did not observe any 150 mutations likely to affect RNA stability: one clone had no additional unique mutations, while 151 the other had nonsynonymous mutations in *bclA3* and *CD3124*. *bclA3* encodes a spore surface 152 protein with no known function in vegetative cells, while CD3124 encodes an orphan histidine kinase of unknown function. Interestingly, four of the hypermutating lineages also had 153 154 mutations in CD3124, an identical frameshift mutation in all four.

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156 Vancomycin resistant clones display reduced fitness

157 To assess the wider consequences of evolved vancomycin resistance for bacterial phenotype, endpoint clones were phenotypically characterised for growth *in vitro*, sporulation efficiency 158 159 and cell morphology. All 10 strains displayed significantly impaired growth in rich liquid media 160 (Supplementary Fig. 1), with particularly severe defects apparent for Bc10 and 11. Several strains were also impaired in sporulation (Supplementary Fig. 2), with a wide range of 161 phenotypes apparent, from a mild defect for Bc2 and 4 and delayed sporulation for Bc8, to 162 more severe defects for Bc7, Bc9 and 10 and complete loss of sporulation for Bc11. These 163 164 growth and sporulation defects were also accompanied by changes in cell length relative to 165 the parental wild type strain (Supplementary Fig. 3). Principal component analysis of the five

wild type-derived endpoint evolved clones (Fig. 1c) revealed all five resistant strains followed similar evolutionary trajectories away from the parental wild type, associated with lower sporulation efficiency and growth defects, albeit with divergence among replicate lines in the extent of defects and cell size. However, there was no apparent sub-clustering by resistance mechanism.

171

172 **Population dynamics in evolving populations**

173 Together the genome sequence data for end-point clones suggests that there are two 174 alternative mechanisms of vancomycin resistance, one involving *dacS* and the other involving 175 vanT. To better understand how selection acted upon these mechanisms we next performed 176 pooled population sequencing at passage 10, 20 and 30 to track mutation frequencies over 177 time (Fig. 2 and Supplementary Figs. 4 and 5). In total, discounting variants found in Bc1 P20, 178 we identified 535 unique variants across the 10 parallel populations and three timepoints. 179 We have removed Bc1 P20 from this analysis as the additional 520 variants identified in that 180 sample alone likely reflect random mutation due to the emergence of a spontaneous hyper-181 mutator phenotype. Impacted genes clustered within 17 distinct functional classes by KEGG 182 analysis (Supplementary Fig. 6), with two component systems and ABC transporters being 183 particularly well-represented. Focusing on the two main routes to resistance identified in 184 endpoint clones, these data revealed highly contrasting selection dynamics, particularly in our wild-type replicate populations: mutations in *dacS* rapidly rose to high frequency, reaching 185 186 fixation by P10. By contrast, mutations in *vanT* arose later and only reached fixation by P20 187 or 30, and were preceded by mutations at other sites which reached high frequency by P10 188 but that ultimately did not survive, being replaced by *vanT* mutants presumably conferring 189 higher levels of vancomycin resistance. The two preceding high frequency mutations in Bc3 190 (both T>TA) were very close together, separated by only 7 bp in an intergenic region 191 downstream of CD0482, encoding a uridine kinase, and approximately 250 bp upstream of *glsA*, encoding a glutaminase. These mutations are outside of the likely promoter region²⁰ but 192 193 it is possible that they affect regulation of *qlsA*. Interestingly, changes in glutamine 194 metabolism have previously been linked to vancomycin resistance in *Staphylococcus aureus*²¹. The single high frequency mutation in Bc5 at P10 is a nonsynonymous substitution in CD3034, 195 196 introducing a Gly255Asp mutation in the encoded D-hydantoinase which may play a role in 197 the synthesis of D-amino acids. Taken together, these data suggest that *vanT* is not required

for first-step resistance; *vanT* mutations either provide higher-level vancomycin resistance,
allowing them to supplant earlier mutations, or they require potentiating mutations to arise
and be selected first. However, no consistent secondary mutations common to all populations
with *vanT* mutations were identified.

202

203 Recapitulation of *dacS* mutations confirms role in resistance

204 As the *dacJRS* cluster had not previously been implicated in vancomycin resistance, we 205 validated the role of DacS in vancomycin resistance by recapitulating individual mutations in a clean genetic background. We chose the variant *dacS* alleles identified in Bc1 (714G>T), as 206 207 this is the sole unique mutation found in that strain, and the variant allele that evolved in 208 parallel in Bc8 and Bc9 (548T>C). Recapitulated strains carrying only the dacS mutation of 209 interest were constructed in the parental R20291\Delta PaLoc by allelic exchange. Introduction of 210 either mutation alone resulted in a 4-fold increase in the vancomycin MIC compared to the 211 parental strain, confirming that DacS is playing a significant role in the evolved resistance we 212 observed (Fig. 3a). AlphaFold prediction of the DacS structure yields a plausible dimer model 213 (Fig. 3b) that is highly similar to previously characterised histidine kinases²². The Bc1 and 214 Bc8/9 mutations described here both result in amino acid changes in the DacS cytoplasmic 215 domain: Bc1 Glu238Asp within the predicted catalytic ATPase (CA) domain and Bc8/9 216 Val183Ala within the dimerization and histidine phosphorylation (DHp) domain. The impact 217 of these mutations on the function of DacS is not clear but we hypothesised that DacS, along 218 with its cognate response regulator DacR, could be regulating the expression of the D,D-219 carboxypeptidase encoded by *dacJ*. To examine this possibility, we extracted RNA from 220 R20291\Delta Paloc, R20291\Delta Paloc dacSc.714G>T and R20291\Delta Paloc dacSc.548T>C, both in the 221 absence and presence of 0.5 µg/ml vancomycin, and assessed the expression of dacS, dacR and *dacJ* by qRT-PCR (Fig. 3c). Either point mutation resulted in a dramatic 4.8-94.6-fold 222 223 upregulation of expression of all three genes and this effect was independent of vancomycin. It is highly likely, therefore, that the overexpression of DacJ and the resulting reduction in 224 225 vancomycin binding sites within the cell wall accounts for the reduction in vancomycin 226 susceptibility in both strains with mutated *dacS*. The genomic organisation in this region (Fig. 3d) and previous global transcription site mapping²⁰ suggests that *dacS* and *dacR* are 227 228 transcribed from a single promoter upstream of *dacR* and that *dacJ* is transcribed from its 229 own separate promoter. Our data demonstrate that both of these promoters are subject to

- regulation by the DacS/DacR two component system, although it is not clear if the observed
- effects are a result of constitutive activation of a TCS that positively regulates these promoters
- 232 or deactivation of a repressor (Fig. 3d).

233 Discussion

Vancomycin is one of the few antibiotics in routine use for treatment of CDI worldwide and is 234 235 now the frontline drug of choice in the UK⁷. High level vancomycin resistance is widespread 236 in *Enterococcus* spp. and in *S. aureus* but has yet to be reported in *C. difficile*, where there are few verifiable reports of reduced susceptibility in clinical strains, despite anecdotal reports of 237 vancomycin treatment failure²⁶. However, it is not clear if this apparent lack of resistance 238 239 reflects an underlying constraint upon the emergence of resistance in this species or is simply 240 an artefact of a lack of routine monitoring in the clinic. Here we show using laboratory experimental evolution that C. difficile can rapidly evolve high-level vancomycin resistance via 241 242 two alternative mechanisms, but that increased resistance is associated with severe 243 pleiotropic effects, including growth and sporulation defects, which may act to limit the 244 emergence of resistance in clinical settings.

245

246 Under ramping vancomycin selection, resistance emerged rapidly in all 10 replicate lines 247 reaching 16 to 32-fold higher MIC within approximately 250 generations. Whole genome 248 sequencing of individual clones from each population at the end of the evolution revealed 249 two evolutionary pathways to resistance, centring around mutations in vanT, encoding the 250 Serine/Alanine racemase component of a VanG-type cluster, and mutations in a cluster of 251 genes encoding a two component system and a D,D-carboxypeptidase (dacJRS). The VanG 252 cluster is common in C. difficile strains but its potential role in vancomycin resistance had 253 been disputed^{14,15}. More recently however, mutations in the VanRS two-component system 254 that derepress the rest of the cluster and reduce vancomycin susceptibility have been 255 identified in both laboratory evolution experiments and in clinical isolates^{16,17}. These 256 observations confirm that changes to the expression of genes in the VanG cluster can indeed contribute to reduced vancomycin susceptibility. Interestingly we detected vanS mutations 257 258 only transiently in our evolution and none became fixed, suggesting they had only a limited 259 contribution to resistance or were accompanied by severe fitness defects (Fig. 4). In contrast, 260 mutations to vanT were common, fixing in four of ten replicate lines. Mutations in the dacJRS cluster were also extremely commonly observed and often rose to high frequency, with 261 variant dacS alleles fixing in three populations (Bc1, 2 and 9), an identical dacRc.532A>G 262 263 variant fixing in two populations (Bc7 and 10) and a *dacJ* variant fixing in Bc9. Mutations in 264 dacS and dacR also transiently fixed in two further populations (Bc8 and 10 respectively) at 265 P20 before being subsequently outcompeted. Interestingly the *dacS* mutation identified in Bc8 at P20 (548T>C) is identical to that observed in endpoint isolates from Bc9. None of these 266 267 genes had been previously linked to vancomycin resistance but the predicted D,D-268 carboxypeptidase activity of DacJ points to a plausible mechanism through removal of the terminal D-Ala residue in nascent peptidoglycan^{27,28} (Fig. 3d). We hypothesised that the two-269 270 component system encoded by *dacS* and *dacR* was regulating expression of *dacJ*, providing a 271 mechanistic link for all of these mutations. This was confirmed by recapitulation of two 272 distinct *dacS* mutations (from Bc1 and Bc8/9) in a clean genetic background, leading to 273 derepression of both *dacJ* and the *dacSR* bicistronic operon. Importantly this effect was 274 independent of vancomycin, leaving open the possibility that the wild type system could still 275 contribute to vancomycin resistance in the appropriate permissive environmental condition. 276

277 Analysis of evolutionary dynamics over the course of the evolution also revealed intriguing 278 differences in the timing of dacJRS vs vanT mutations (summarised in Fig. 4). dacJRS 279 mutations had fixed by P10 in nearly every population in which they persisted to the end of 280 the evolution, the exception being Bc9 *dacJ* which first appeared and fixed at P20. However, 281 that population also had a superseding *dacS* mutation that had fixed by P10. By contrast, 282 mutations in *vanT* typically didn't fix until P20 or P30, although sometimes present at lower 283 frequency at earlier timepoints. Delayed emergence of vanT variants may also indicate a reliance on preexisting potentiating mutations, although no consistent secondary mutations 284 285 were found in all *vanT* lineages. The two resistance mechanisms also seemed to be mutually exclusive, *vanT* mutations did not co-occur with *dacJRS* in any population or endpoint isolate. 286 It is possible that early emergence of mutations in *dacJRS* precludes subsequent mutations to 287 288 *vanT* and commits the population to that pathway.

289

In all populations, emergence of resistance in our experiment was accompanied by severe, if somewhat variable, fitness costs, importantly including severe sporulation defects in two lineages. As the spore is the infectious form of *C. difficile* and an absolute requirement for patient to patient transmission²⁹, these sporulation defects would likely have serious consequences for the infectivity and onward transmission of evolved resistant isolates. The genetic basis of the sporulation defects in these isolates is not clear, as sporulation is an extremely complex and poorly understood cell differentiation process^{30,31}, but if a similar trait

297 were to emerge in a patient during vancomycin treatment this would be an evolutionary dead 298 end. These fitness defects may well explain delays to the emergence of vancomycin resistance 299 in the clinic. However, it is also possible that the accumulation of refining mutations would 300 eventually lead to gradually improving fitness as has been seen in long term evolution experiments^{32,33}. Indeed, we saw extensive evidence of succession here, with early high 301 302 frequency mutations that conferred moderate increases in MIC being completely supplanted 303 by later variants, of presumably improved fitness. The potential for further evolution in our 304 experiment was also clear from the continual emergence of new variants, even at the final 305 timepoint. It remains to be seen how far towards full resistance C. difficile can go given the 306 enough time and the right conditions but, given the increasing reliance on vancomycin in the 307 treatment of CDI, it is crucial that we begin to understand the possible routes to resistance. 308 The question of how likely *C. difficile* vancomycin resistance is in the real world remains open 309 but this work, and other efforts towards understanding possible mechanisms of resistance, 310 will hopefully provide a roadmap to guide genomic surveillance efforts.

311 Methods

312 Strains and Growth Conditions

313 All strains used or generated in the course of this study are described in Supplementary Table 314 1. C. difficile was routinely cultured on brain heart infusion (BHI) agar and in tryptone yeast 315 (TY) broth. C. difficile was grown in an anaerobic cabinet (Don Whitley Scientific) at 37°C, with 316 an atmosphere composed of 80% N₂, 10% CO₂ and 10% H₂. Media was supplemented with 317 thiamphenicol (15 μ g/mL) (Sigma) and colistin (50 μ g/mL) (Sigma) as appropriate. For counter selection against plasmids bearing the codA gene, C. difficile differential media with 5-318 319 fluorocytosine (CDDM 5-FC) was used as described previously³⁴. *E. coli* was routinely cultured 320 in Luria-Bertani (LB) broth or agar at 37°C, and supplemented with chloramphenicol (15 321 μ g/mL) (Acros Organics) or kanamycin (50 μ g/mL) (Sigma) as appropriate.

322

323 Molecular biology

324 All oligonucleotides and plasmids are described in Supplementary Tables 2 and 3 respectively. 325 Plasmid miniprep, PCR purification and gel extractions were performed using GeneJET kits 326 (Thermo Fisher). High-fidelity PCR amplification was performed using Phusion polymerase 327 (NEB), and standard PCR amplification was performed using Tag mix red (PCRBIO), according 328 to manufacturers' instructions. Gibson assembly primers were designed using NEBBuilder 329 (NEB) and restriction digestion and DNA ligation was performed using enzymes supplied by 330 NEB. Plasmids were transformed into NEB5a (NEB) or CA434 competent E. coli cells according to the NEB High Efficiency Transformation Protocol. The sequences of cloned fragments were 331 332 confirmed using Sanger sequencing (Genewiz, Azenta Life Sciences, Germany).

333

334 C. difficile mutagenesis

335 *C. difficile* strain R20291 was modified for use in evolution experiments and to recapitulate 336 individual mutations. Homology arms for introducing mutations onto the *C. difficile* 337 chromosome by allelic exchange were generated either by Gibson assembly of PCR products or synthesised by Genewiz (Azenta Life Sciences) and then subsequently cloned between BamHI and SacI sites in pJAK112²⁰, a derivative of pMTL-SC7215³⁴. Following confirmation by Sanger sequencing, plasmids were transformed into *E. coli* CA434 and transferred to *C. difficile* by conjugation³⁵. Homologous recombination was performed as previously described³⁴ and mutations were confirmed by PCR and Sanger sequencing of mutated regions.

344 R20291 was rendered avirulent via deletion of the complete pathogenicity locus (encoding 345 toxins A and B) using plasmid pJAK143³⁶, yielding strain R20291 Δ PaLoc. This strain was then 346 further modified by deletion of DNA repair operon *mutSL* to create a hyper-mutator strain. 347 1.2 kb up and downstream of the *mutSL* operon was amplified by PCR using oligonucleotides 348 RF2066 and RF2067, and RF2068 and RF2069 (Supplementary Table 2), respectively, and cloned between BamHI and SacI sites in plasmid pJAK112. The resulting plasmid, pJEB002, 349 was then conjugated into R20291\Delta PaLoc and allelic exchange performed as described above, 350 yielding R20291∆*PaLoc*∆*mutSL*. Enumeration of CFUs on BHI agar containing rifampicin (0.015 351 µg/mL) suggested this strain has a mutation rate approximately 20-fold higher than 352 353 R20291 Δ PaLoc. Five barcoded variants of both R20291 Δ PaLoc and R20291 Δ PaLoc Δ mutSL 354 were then generated by insertion of unique sequencing barcodes. Briefly, pJAK081 (identical 355 to pJAK080²⁰ but in the pMTL-SC7215 backbone) was modified by introduction of a synthetic 356 DNA fragment containing a multiple cloning site and a 9 bp sequencing barcode (Barcode 1), 357 flanked by the *fdx* and *slpA* terminators, between the existing homology arms for insertion of 358 DNA between CD0188 (pyrE) and CD0189 in the R20291 genome. A second plasmid (pJAK202) 359 containing Barcode 2 was constructed in the same manner and the rest (pJAK203-205 and 360 207-211) were generated via site directed mutagenesis using pJAK201 as a template. These ten plasmids were then used to generate R20291 Δ PaLoc Bc1-5 and R20291 Δ PaLoc Δ mutSL 361 362 Bc7-11. Recapitulated strains were generated by introducing point mutations into 363 R20291 Δ PaLoc. An approximately 2 kb synthetic DNA fragment, centred on the mutation of interest, was cloned between BamHI and SacI sites in pJAK217, generating pJEB019 364 365 (dacSc.548T>C) and pJEB026 (dacSc.714G>T). The resulting plasmids were conjugated into R20291 Δ PaLoc, and allelic exchange was performed as described above, generating 366 R20291∆PaLoc dacSc.548T>C and R20291∆PaLoc dacSc.714G>T respectively. 367

369 Evolution

Directed evolution of *C. difficile* was performed using a broth-based gradient approach in which 10 individually barcoded parallel lines were evolved for a period of 30 passages. A 6well plate for each of the parallel lines was prepared for each passage using 4 mL of TY broth with vancomycin, spanning a gradient of 0.25 to 8x the current MIC, as determined from the most recent passage for each line, allowing the gradient to rise with increasing resistance.

The evolution was initiated using overnight cultures from single colonies, adjusted to OD_{600nm} 0.1. 10 µL was added to each well, before incubating for 48 h at 37°C. Plates were visually inspected after 48 h and 10 µL of the well with the highest antibiotic concentration supporting growth was used to inoculate the wells of the subsequent passage. For each parallel line, a control well was passaged without antibiotic. 1 mL of a population, and the corresponding control, was frozen at -80°C in 15% glycerol whenever the MIC increased; and after passages 10, 20 and 30.

382

383 gDNA Extraction

384 gDNA was obtained from *C. difficile* cultures using the phenol-chloroform method as 385 described previously³⁰. DNA concentration was quantified using Qubit, and purity was 386 assessed via microvolume spectrometry.

387

388 Sequencing

End-point (P30) isolates and respective controls were gained through plating P30 populations and culturing single colonies. Vancomycin MICs were determined by agar dilution (described below), and gDNA extraction and quantification were performed as above. Library prep (Nextera XT Library Prep Kit (Illumina, San Diego, USA)) and 30x illumina sequencing (NovaSeq 6000, 250 bp paired end protocol) of isolates was performed at MicrobesNG (Birmingham, UK). Reads were trimmed at MicrobesNG using Trimmomatic (v0.30)³⁷ with a sliding window guality cut-off of Q15. 396 Pooled sequencing of the 10 evolved and 10 control populations was performed at 3 397 timepoints – P10, P20 and P30. 1 mL of the well with highest antibiotic concentration 398 supporting growth was taken after 48 h, harvested via centrifugation, and frozen at -20°C. 399 gDNA extraction and quantification were performed as above. Library prep (Nextera DNA Flex 400 Library Prep Kit (Illumina, San Diego, USA)) and 250x Illumina sequencing (NovaSeq 6000, 150bp paired end protocol) was performed at SNPsaurus (Oregon, USA). Reads were trimmed 401 using Trimmomatic (v0.39) with the following criteria: 402 leading:3; trailing:3; 403 slidingwindow:4:15; minlen:36.

404

405 Sequencing Analysis Pipeline: Isolates

Trimmed reads were checked using FastQC (v0.11.9)³⁸ to ensure sufficient quality for analysis. 406 Analysis was primarily performed using a custom script, built based on a resistant mutant 407 408 analysis pipeline³⁹. Reads were aligned to the reference (*C. difficile* R20291, accession number: FN545816) using BWA-mem (v0.7.17)⁴⁰ and sorted using SAMtools (v1.43)⁴¹. 409 Coverage across the genome was inferred using Bedtools (v2.30.0)⁴² genomecov and map 410 411 functions. PCR duplicates removed Picard (v2.25.2) were via 412 (http://broadinstitute.github.io/picard/). The mpileup utility within SAMtools (v1.43) was 413 used to generate the mpileup file required for Varscan. Variants were then called using Varscan (v2.4.3-1)¹⁸ mpileup2cns (calling SNPs and indels) using the following parameters: 414 415 min-coverage 4; min-reads2 4; min-var-freq 0.80; p-value 0.05; variants 1; output-vcf 1. Here, 416 a minimum of 4 reads were required to support a variant, and the cut-off for variant calling was 80%. Vcf files were annotated using snpEff (v5.0)⁴³. Variants were also called using Breseq 417 (v0.35.5)¹⁹, using default parameters, and putative variants were retained if detected in both 418 419 analysis pipelines. All variants were manually verified using IGV (v2.8.6)⁴⁴. Variants that were 420 also called in control lines were removed using Varscan (v2.4.3-1) compare, generating a list 421 of variants unique to vancomycin resistant lines.

422 Nonsynonymous mutations occurring within genes were plotted using a previously published
 423 custom script in RStudio (v4.1.0) utilising the Plotrix package⁴⁵ to visualise parallel evolution.

424

425 Sequencing Analysis Pipeline: Populations

426 The 10 evolved and 10 control populations, sequenced at P10, P20 and P30, were evaluated using a population analysis pipeline that followed the same custom script as for isolate 427 428 analysis, with modifications to SNP calling parameters and filtering. InSilicoSeq (v1.5.4)⁴⁶ was 429 used to rapidly simulate realistic sequencing data at a range of coverage depths (80x, 100x, 430 300x) based on the R20291 genome, with SNPs seeded at 5% frequency. Simulated sequences 431 were analysed using the custom pipeline, and a range of Varscan parameters (P-values, min-432 coverage, min-reads2) were trialled to generate a set of parameters to accurately call variants 433 without false positives.

434 Varscan parameters used to call population variants were dependent on average coverage 435 depth, allowing a strong evidence base whilst calling low frequency variants. Samples were 436 separated by average coverage: the following Varscan (v2.4.3-1) mpileup2cns parameters 437 were used for samples with 100x or higher average coverage: min-coverage 80; min-var-freq 0.05; p-value 0.05; variants 1; output-vcf 1. A minimum of 4 reads were required to support 438 439 a variant, and the cut-off for variant calling was 5%. For samples with average coverage below 440 100x, the following Varscan (v2.4.3-1) mpileup2cns parameters were used: min-coverage 4; 441 min-reads2 4; min-var-freq 0.05; p-value 0.05; variants 1; output-vcf 1. As with high coverage 442 samples, a minimum of 4 reads were required to support a variant, meaning the minimum 443 variant frequency which could be called was inversely scaled (4/coverage at position), with a minimum frequency of 5%. This process identified a list of variants within each line at each 444 time point, and their respective frequencies. 445

As in the isolate pipeline, population variants were filtered using Varscan (v2.4.3-1) compare, removing variants that were also present in control lines to generate a list of variants unique to vancomycin resistant lines. Further filtering of this dataset, again using Varscan (v2.4.3-1) compare, discarded variants that never reached above 10% frequency by the end of the evolution. Variants were also manually inspected to ensure no variants remained the same frequency across all time points.

452 Mutations occurring within genes (except those in Bc1 P20) were coloured according to their
 453 barcoded replicate line, and displayed in KEGG Mapper - Color⁴⁷ to view KEGG pathways.

Nonsynonymous mutations occurring within genes and rRNAs were plotted using a previously
 published custom script in RStudio (v4.1.0) utilising the Plotrix package⁴⁵ to visualise
 population dynamics, where point size indicated mutation frequency in the population.

457

458 Strain fitness analysis

Growth curves were performed anaerobically in 96 well plates using a Stratus microplate reader (Cerillo). Overnight cultures of *C. difficile* were diluted to OD_{600nm} 0.05 and incubated at 37°C for 1 h. Plate lids were treated with 0.05% Triton X-100 + 20% ethanol. Plates were prepared using 200 µL of equilibrated culture. Samples were measured at minimum in biological and technical triplicate. The OD_{600nm} was measured every 3 min over a 24 h period. Data was plotted in Graph pad Prism (v9.0.2), and was analysed in RStudio (v4.1.0) using the GrowthCurver package (v0.3.0)⁴⁸.

To assess sporulation efficiency, triplicate overnight *C. difficile* cultures were adjusted to OD_{600nm} 0.01 and grown for 8 h. Cultures were then adjusted again to OD_{600nm} 0.01, subcultured 1:100 into 10 mL BHIS broth, and grown overnight to obtain early stationary phase spore-free cultures (T = 0). At T = 0, and the following 5 days, total viable counts were enumerated by spotting 10-fold dilutions in technical triplicate onto BHIS agar with 0.1% sodium taurocholate. Colonies were counted after 24 h incubation. Spore counts were enumerated using the above method following heat treatment (65°C for 30 min).

To visualise cell morphology, *C. difficile* samples were harvested via centrifugation, washed
twice in PBS, and fixed in 4% paraformaldehyde; before harvesting and resuspension in dH₂O.
Samples were mounted in 80% glycerol, and imaged using a 100x Phase Contrast objective on
a Nikon Ti eclipse widefield imaging microscope using NIS elements software. Images were
analysed in Fiji (v2.9.0) using MicrobeJ (v5.13l)⁴⁹.

478

479 Vancomycin MICs

480 MICs were obtained via standard agar dilution methods⁵⁰. Briefly, overnight cultures were 481 adjusted to OD_{600nm} 0.1. 2.5 µL of sample was spotted in biological triplicate and technical 482 duplicate onto BHI plates with ranging antibiotic concentrations. MICs were determined after 483 48 h incubation, and plates were imaged using a Scan 4000 colony counter (Interscience).

484

485 Principal Component Analysis

WT (Bc1-5) P30 resistant isolates were characterised in terms of their sporulation efficiency, growth rate, MIC and cell length. These were compared, along with the WT ancestor from the start of the evolution, in a principal component analysis (PCA). The PCA was computed using the prcomp() function in Base R (http://www.rproject.org/), and visualised using the factoextra package. The first 2 PC were plotted, as these accounted for 93% variance. The loadings were added in their respective locations, and isolates were coloured based on resistance mechanism.

493

494 **qRT-PCR**

495 Total RNA was extracted using the FastRNA pro blue kit (MP Biomedicals). Cells were grown to an OD_{600nm} of approx. 0.4, and 2 volumes of RNA protect (Qiagen) were added. Cells were 496 497 incubated for a further 5 min, before harvesting via centrifugation. Cell pellets were stored at 498 -80°C. Pellets were resuspended in RNA Pro solution, and transferred to a tube containing lysing matrix B (MP Biomedicals). Cells were lysed via FastPrep (2 cycles of 20 s, 6 m/s), and 499 500 centrifuged (16,200 x g, 4°C, 10 min) to remove insoluble cell debris. The supernatant was 501 transferred to a microfuge tube, and 300 µL chloroform (Sigma) was added. Samples were 502 centrifuged again (13,000 rpm, 4°C, 15 min), and the supernatant was precipitated at 20°C 503 overnight after addition of 500 μ L 100% ethanol. After precipitation, RNA was harvested by centrifugation (13,000 rpm, 4°C, 15 min), washed with 70% ethanol, and dried. Precipitated 504 505 RNA was resuspended in 50 μ L nuclease-free water, residual DNA was removed using the 506 Turbo DNA-free kit (Invitrogen) and the RNA was cleaned and concentrated with the RNeasy 507 Minelute cleanup kit (Qiagen), as per manufacturers' instructions.

cDNA was generated using Superscript III (Invitrogen). 5 μg RNA was mixed with 2 μL dNTP
mix and 1 μL 100 mM random primer (Eurofins), heated at 65°C for 5 min, and cooled on ice.
8 μL 5X buffer, 2 μL 0.1M DTT (Invitrogen), 1 μL RiboLock RNase inhibitor (Thermo Scientific)
and 2 μL Superscript III were added, before incubation at 25°C (5 min), 50°C (30 min) and 70°C
(15 min). cDNA was adjusted to 40 ng/μL. RT negative controls were made as above, without
presence of Superscript III.

514 Expression was measured against an exact copy number control, via standardisation with a 515 plasmid containing one copy of each target DNA sequence⁵¹. *rpoA* was used as the 516 housekeeping gene to normalise results, and *rnpA* was used as an additional unrelated 517 control. A plasmid (pJEB029) was synthesised with the pUC-GW-Kan backbone (Genewiz), 518 containing approximately 200 bp target gene fragments of *dacS*, *dacR*, *dacJ*, *rpoA* and *rnpA*. This was purified using the GeneJET plasmid miniprep kit (Thermo Fisher), and linearised using 519 NdeI (NEB) and diluted to known copy number (2 x 10^8 per μ L). A qPCR mastermix was 520 521 assembled, containing 25 µL SYBR Green JumpStart Taq ReadyMix (Sigma), 7 µL MgCl2 in buffer (Sigma), forward and reverse primers (concentration determined by prior 522 523 optimisation) and nuclease-free water up to 45 μ L. 5 μ L of cDNA (40 ng/ μ L), 5 μ L of RT 524 negative control, or 5 µL gPCR plasmid pJEB029 (diluted serially in lambda DNA (Promega)) 525 were added, and gPCR was performed (BioRad CFX Connect Real Time System). Copy 526 numbers were calculated using BioRad CFX manager (v3.1), and data were analysed in 527 Microsoft Excel (2016) to generate copies per 1000 copies of *rpoA*. Data were graphed and 528 statistically analysed in Graph pad Prism (v9.0.2).

529

530 Structural modelling of DacS

DacS was modelled as a dimer using AlphaFold²³ and the resulting output files were visualised
 using ChimeraX⁵².

533

534 Statistics

535 Statistical analysis was performed in Graphpad Prism (v9.0.2), and P < 0.05 was considered significant. Data are presented as mean ± SD, unless otherwise stated. Differences in growth 536 537 were analysed using the R package GrowthCurver outputs. A cross-correlation was performed in RStudio using Hmisc and corrplot packages^{53,54}, which determined area under curve (AUC-538 E) to be the most representative measure of growth. The differences in AUC-E, compared to 539 the control strain curves, were calculated using t-tests with Welch's correction. To test 540 differences in sporulation efficiency, AUC was chosen as a representative measure of 541 542 sporulation across all timepoints. AUC was calculated in Graphpad Prism (v9.0.2), and the difference between AUC for P30 isolates was compared with the WT using a one-way ANOVA 543 544 with Brown-Forsythe and Welch's correction. Statistical analysis of cell length was performed 545 in Graphpad Prism (v9.0.2). Fiji (v2.9.0) MicrobeJ (v5.13l) cell length outputs for P30 isolates 546 were compared to the WT using a 1-way ANOVA with Brown-Forsythe and Welch's correction to test for differences in cell length. To determine whether differences in expression of dacS, 547 548 dacR, dacJ and rnpA were significant between R20291\Daloc, R20291\Daloc dacSc.548T>C 549 and R20291 Δ *PaLoc dacSc*.714G>T, a two-way ANOVA with Dunnett's multiple comparisons 550 was performed in Graphpad Prism (v9.0.2).

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705 Data Availability

Genome sequence data for parental strains, P30 resistant isolates and respective controls, as
well as pooled population sequencing data for resistant populations and controls at P10, 20
and 30 is deposited with the European Nucleotide Archive (ENA). The accession numbers for
these may be viewed in Supplementary Table 4.

710

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- 719

720 Author contributions

JEB carried out all experiments, collected and analysed data, revised the sequencing analysis pipelines, wrote and revised the manuscript. RCTW provided guidance on sequencing analysis. CET designed experiments, provided guidance and provided tools for qRT-PCR experiments and analysis. RRC provided guidance on bioinformatics analysis, provided bioinformatics scripts, and analysed KEGG data. MAB designed the study, interpreted evolution data, supervised the study, and revised the manuscript. RPF designed the study, analysed data, supervised the study, wrote and revised the manuscript.

728

729 Competing interests

Summit Therapeutics Inc were industrial partners for JEB's MRC DiMeN iCASE PhD
studentship but had no input in study design, interpretation or manuscript preparation. The
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734 Materials and correspondence

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Fig. 1 Evolution of vancomycin resistance. a Changes in apparent vancomycin MIC over the
course of a 30 transfer experimental evolution. MIC was determined as the well with the
lowest vancomycin concentration showing no clear growth. b Shown are the means and

standard deviations of the apparent MIC for five wild type (open squares) and five hyper-742 743 mutating (closed circles) populations. Linear regressions fitted to each data set, blue and pink 744 respectively, are significantly different by ANCOVA, P=0.0008. c Shown are the chromosomal 745 locations of non-synonymous variant alleles in isolated wild type (top) and hypermutating 746 (bottom) C. difficile end-point clones, excluding any mutations that were also identified in any of the control strains. Each circle represents a single C. difficile genome, colour coded 747 according to population as indicated in the key on the left. A full list of all variants shown here 748 749 and including synonymous and intergenic mutations is included in Supplementary Data 2. d 750 Principal Component Analysis (PCA) of P30 isolates from populations Bc1-5 (coloured points) 751 vs the ancestral strain (black triangle), with PC1 versus PC2 plotted, accounting for 93% variance. The loadings (sporulation efficiency, growth, MIC, cell length) are shown in 752 753 respective locations. Arrows show the evolutionary trajectories of wild-type replicate lines from their ancestor in multivariate phenotype space. 754





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Fig. 2 Genomic location of gene variants over time. Accumulation of variants in the wild type *C. difficile* lineages. Each circle plot represents the 4.2 Mb genome of a single evolving population after 10 (inner ring), 20 (middle ring) and 30 passages (outer ring), with the locations of non-synonymous within gene variants indicated with black circles and the penetrance of each mutation in the population indicated by the size of the circle. The line

graphs show the frequency of all variants (intergenic, synonymous, non-synonymous, frameshifts and nonsense) in each population. The vancomycin MIC for each population is also indicated by the shaded region. Mutations also identified in the respective end point clone (Fig. 1b) are highlighted by the coloured lines. Note population Bc1 evolved an apparent hypermutator phenotype prior to P20, with 520 variants identified at that time point. For simplicity only variants present in P10 and P30 are labelled. A full list of all variants shown here, including those in Bc1 P20, is included in Supplementary Data 3.





Fig. 3 *dacS* mutations lead to dysregulation of *dacJRS*. a Vancomycin MICs of R20291Δ*PaLoc*,
 R20291Δ*PaLoc dacS*c.714G>T and R20291Δ*PaLoc dacS*c.548T>C as determined by agar
 dilution. Assays were performed in biological triplicate and technical duplicate, four

representative spots are shown for each strain. **b** AlphaFold model of DacS as a dimer²³. The 776 777 transmembrane domains were identified using DeepTMHMM²⁴ and the Catalytic ATPase and 778 Dimerization and Histidine Phosphorylation domains were predicted using InterProScan²⁵. 779 The locations of Val183 and Glu238 are highlighted in purple on one chain. **c** qRT-PCR analysis 780 of dacJRS expression in R20291\DataPaLoc (black bars), R20291\DataPaLoc dacSc.714G>T (yellow bars) and R20291 Δ PaLoc dacSc.548T>C (green bars). rnpA, which is implicated in the vanT 781 resistance pathway, was included as an additional unrelated control. Expression was 782 783 quantified against a standard curve and normalised relative to the house-keeping gene rpoA. Assays were performed in biological and technical triplicate. Statistical significance was 784 785 calculated using a two-way ANOVA with Dunnett's multiple comparison, ** = P<0.001. **d** Two 786 mechanisms by which DacS mutations could alter expression of *dacJRS*. Phosphorylated-DacR 787 could be an activator or repressor of the two promoters in the *dacJRS* locus, with DacS Glu238Asp or Val183Ala substitutions either constitutively activating or inhibiting the 788 789 function of the TCS respectively. In either scenario, the consequence is over-expression of 790 DacJ which is then translocated to the cell surface where it can cleave the terminal D-Ala 791 residue in nascent peptidoglycan, thereby preventing vancomycin binding.



- 794 Fig. 4 Two distinct pathways to resistance. Relative frequencies and time of emergence of
- 795 mutations in genes *dacS*, *dacR*, *dacJ*, *vanT* and *vanS* across all ten evolving populations.