Running title: Microbiota predict <i>C. difficile</i> severity in humanized mice
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Abstract

Clostridium difficile causes diarrheal disease when it successfully colonizes a dysbiotic gut microbial community. Current mouse models to study C. difficile infection (CDI) rely on pre-treatment with antibiotics to disrupt the mouse microbiome prior to inoculation. This model does not allow for analysis of human-associated microbial community members that modulate C. difficile colonization and expansion. To study human-associated microbes in the context of CDI, we inoculated germ-free C57BL/6 mice with one of 16 human fecal samples from diarrheal or healthy donors and challenged with C. difficile 14 days later. Five unique donor-mice combinations resulted in severe CDI while the remaining 11 only experienced mild disease. Both healthy and diarrheal donors 11 were susceptible to colonization and severe symptoms of CDI. To determine if specific microbes were associated with disease severity outcomes, we built a classification Random Forest machine learning model based on relative abundance data of the communities prior to infection. The model identified a number of bacterial populations associated with the development of severe 16 CDI, including Bacilliales, Ruminococcaceae, Ruminococcus, Staphylococcus, Streptococcus and Bacteriodetes. Additionally, a regression model accurately 18 predicted colonization levels of *C. difficile* at one to ten days post-infection. This model explained 99% of the variance in the number of CFU isolated from 20 mouse stool. Members of Lachnospiraceae, Parabacteroides, Bacteroidales, 21 Bacteroidetes, Porphyromonadaceae and unclassified Bacteria families were predictive of future C. difficile colonization levels. Finally, challenging these mice with different strains of C. difficile revealed that susceptible human-associated microbial communities were prone to severe disease independent of strain 25 type. Taken together these results suggest that human-associated microbial communities can be recapitulated in germ-free mice and used to characterize

- dynamics of CDI. Because both healthy and diarrheal patients were susceptible
- 29 to severe CDI, machine-learning models are useful to identify bacterial
- populations that allow colonization and contribute to the development of C.
- 31 difficile associated disease in humans.

2 Introduction

- Clostridium difficile is an opportunistic pathogen of the human lower gastrointestinal tract. Disruption of the native microbial community of the gut by antibiotics is the most common risk factor for development of C. difficile infection (CDI) (1). C. difficile is a spore-forming bacteria and can persist on abiotic surfaces and is not readily killed by ethanol-based hand-sanitizers, putting hospital patients particularly at risk. Indeed, ~12% of hospital acquired infections in the United States are due to C. difficile and result in up to 15,000 deaths annually (2). This is a test (@Britton12).
- Murine models to study CDI typically rely on treating conventionally-raised mice
 with antibiotics either in drinking water or by injection to induce susceptibility (3,
 4). This model provides a convenient way to study C. difficile pathogenesis and
 virulence factors. Numerous microbiome studies have been performed using
 this model to determine the antibiotic classes (5), starting microbial community
 (6) and metabolites (7) that impact development and severity of CDI. While
 informative, these studies are somewhat removed from human disease because
 they only examine mouse-associated microbial communities.
- Gnotobiotic or germ-free mouse models have been used for a range of studies of CDI, including assessment of species-specific interactions between C. difficile and competing microbial community members (8), analysis of nutrient restriction (9), in vivo transcriptomics of C. difficile and examination of host immune response to CDI (10). Further, CDI therapeutics such as antibiotics and fecal microbiota transplants have been tested extensively in a gnotobiotic-piglet or piglet-to-gnotobiotic-mouse model of disease (11), (12). Pigs have a longer digestive tract with components more similar to humans than mice and are typically infected by strains typical in human infection (13). However, the murine

and porcine microbiomes typically do not resemble those of the human gut.

The power of the gnotobiotic models to study CDI has been further realized by first inoculating germ-free mice and piglets with human stool microbes. In one study, germ-free piglets were acutely colonized with human feces for one week 61 and then treated with tigecycline. After challenge with C. difficile none of the 62 antibiotic-treated piglets succumbed to infection, while some of the untreated human-colonized pigs did (11). Further, germ-free mice colonized with human feces were bred over several generations to create a cohort of mice with identical human-derived microbiomes (14). These mice were subsequently treated with a 66 five-antibiotic cocktail to induce dysbiosis and then were successfully colonized by C. difficile (14). While informative, these studies were limited in their use of only one human donor as input inoculum. In order to best understand the impact of C. difficile pathogenesis on human disease, we must have a laboratory model that allows for study of a variety of human-derived microbiomes. 71

To test the impact of individual human microbiomes on CDI, we colonized germ-free mice with 16 different human stool donors. We then characterized human-associated microbiome response to C. difficile challenge. Additionally, the use of machine-learning models allowed us to build a predictive model that classified "at-risk" microbiomes prior to infection with C. difficile. These findings show that human-associated microbiomes can be at risk for CDI even in the absence of antibiotics and that study of mice colonized with human feces provides a range of clinical outcomes.

Results

Germ-free mice inoculated with human feces as model for C. difficile 81 infection. To generate mice with human-derived microbiomes, we inoculated one cage of gnotobiotic C57/BL6 mice with one of 16 different human fecal donors. Five donors were patients that had diarrhea that was not attributable to C. difficile infection while 11 donors were healthy at time of donation. Stool from a patient that was colonized with virulent C. difficile was used as a positive 86 control. After inoculation with human stool, mice were allowed to equilibrate for 14 days. Prior to infection, stool samples were taken from each mouse to establish baseline. Then, the C. difficile strain isolated from the positive control patient's sample (strain 430) was used to infect each mouse with 100 90 spores. Mice were monitored for weight loss and clinical signs of disease. 91 Fecal samples were taken to enumerate C. difficile CFU and for microbiome analysis every day for up to 10 days post-infection (Fig 1A). To ensure that 93 the donors we selected represented a diverse array of human microbiomes, we sequenced the 16S rRNA genes from donor fecal inocula. Ordination of 95 the distances between donor communities showed that the donors each had distinctly different communities, independent of whether the sample came from 97 a sick or healthy person (Fig 1B). Likewise, the starting microbial communities of the mice on day 0 were characterizing by sequencing of fecal pellets DNA prior to infection. Ordination of all of the mouse communities on day 0 shows that 100 mice were similar to each other within each cage and donor, but distinct from 101 other donors (Fig 1C). This result confirmed that human-associated microbes 102 were able to colonize gnotobiotic mice and provide distinct initial communities to 103 test C. difficile dynamics. 104

C. difficile infection in mice with human-derived microbiota cause a range

of outcomes. C. difficile colonization was monitored by daily plating of stool pellets for C. difficile CFU. Nearly all of the mice were colonized to $10^5 - 10^7$ 107 CFU by one day post-infection and remained colonized at that level until the end 108 of the experiment (Fig 2A). As one indicator of disease, mouse weights were 109 taken each day post-infection and weight-loss was monitored alongside clinical 110 signs of disease. When mice were judged to be too ill to continue they were humanely euthanized. Overall, disease phenotypes fell into two classes. Mice 112 that became severely ill and lost 20% or more of their starting body weight within 113 one to two days post-infection were classified as "severe" whereas mice that were colonized with C. difficile but did not show signs of disease or severe weight 115 loss were considered to have "mild" disease (Fig 2A, 2B). Interestingly, C. difficile was able to cause severe disease in both mice that had been colonized with 117 healthy stool and those colonized with diarrheal stool, suggesting susceptibility to CDI is dependent on the composition of the starting microbiome and not 119 associated with donor clinical metadata.

Results to be written 1. Microbes present in the gut prior to infection are predictive of C. difficile CFU and severity a. Figure 3: Random forest to predict CFU, predictive OTUs b. Figure 4: Random Forest predicts CDI severity, predictive OTUs 2. Propensity for severe CDI is community-dependent and strain-independent a. Figure 5: Infection of mice with different C. difficile strains.

126 Discussion

Restate results, • caveats about mouse weights • No donors were colonization
 resistant, discuss donor differences • Discuss prediction methods and outcomes
 Discuss potential mechanisms for interesting OTUs • Discuss different strain
 results • Future work blah blah

Materials and Methods

- Mice ULAM number Donor stool ERIN IRB shit Bacteria/plating Sequencing
- Data analysis, code availability Machine learning models

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136 Figure Legends

Figure 1. Germ-free mice inoculated with human feces as a model for C. 137 difficile infection. A) Experimental design. Stool from 16 healthy, diarrheal 138 and CDI patients were independently inoculated into 3-4 germ-free mice by 139 oral gavage. 14 days later mice were orally infected with 100 spores of C. 140 difficile strain 431. Weight and stool CFU were monitored for up to 10 days post 141 infection. B) NMDS ordination of donor stool communities prior to inoculating 142 mice. Each point represents one donor and are colored by clinical diagnosis. C) NDMS ordination of the stool communities on day 0. Each symbol represents 144 one mouse and is colored by donor. Circles represent mice that experienced mild disease and triangles represent those that suffered severe disease. 146

Figure 2. C. difficile infection results in mild or severe disease. A) C. difficile CFU was enumerated by plating of mouse stool pellets daily. Each point 148 represents a mouse and the lines represent the median CFU in each group. Error bars are interquartile ranges. Red lines and points correspond to mice 150 that succumbed to severe disease whereas black lines and points correspond 151 to mice that had mild or no disease. B) Mouse weights were recorded and daily 152 percent weight loss calculated for each mouse. Data is presented as the median 153 of each group and interquartile ranges. Mice that succumbed to severe infection 154 typically lost a significant amount of weight by day 1 or 2 post infection. Red 155 lines correspond to severely ill mice, black to mice with mild disease.

Figure 3. Random Forest prediction of *C. difficile* colonization level. A)

OTUs above 1% relative abundance on day 0 were used to predict median

log₁₀ CFU of *C. difficile* after colonization. OTUs were chosen such that they

were not predictive of cage or donor. Each point is a mouse colored by cage.

B) Partial dependency plots of the top six predictive OTUs. Line displats the

partial dependence of log₁₀ CFU on the relative abundance of each predictive
OUT. Each median log₁₀ CFU is plotted against its relative abundance for each
predictive OTU.

Figure 4. Random Forest prediction of CDI severity. OTUs above 1% relative abundance on day 0 were used to predict disease severity. OTUs were chosen such that they were not predictive of cage or donor. Predictive classification tested via 10-fold (gray), leave-one-cage-out (purple dashed) or leave-one-mouse-out (blue dashed) models are displayed in (A). B) Partial dependency plots of most predictive OTUs. Line displays the partial dependence of log₁₀ CFU on OTU relative abundance. Points are the OTU relative abundance of each mouse colored by outcome (red, severe, black, mild).

Figure 5. Infection of mice with different *C. difficile strains*. 3 strains of *C. difficile* were used to infect mice colonized with susceptible (DA00578) or

resistant (DA00369, DA00430) human donor stool. A) *C. difficile* stool CFU was

enumerated over 10 days. B) Percent weight loss was calculated each day for

each mouse. In both plots, each mouse is a point and lines represent the mean

of each cage.

Supplement Table S1: Mouse day 0 communities by donor genera (avg + stdev of cage)

181 References