1	Unprepared colonoscopy identifies microbiota specific to
2	the proximal and distal human colon
3	Running title: Unprepared colonoscopy identifies proximal and distal human colonic microbiota
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10

2 Abstract

The human colon contains chemicals and nutrients that change along the proximal to distal gut axis. These gradients create microenvironments that affect the distribution and composition of the gut microbiota. The microbiome has been implicated in the colonic diseases colorectal cancer 15 (CRC) and inflammatory bowel disease (IBD). Further, these diseases exhibit different symptoms 16 depending on the location of the colon they are found in. CRC tumors of the proximal and distal 17 colon are morphologically and genetically distinct. Similarly, bowel diseases such as Crohn's are typically exacerbated in the proximal instestine while ulcerative colitis patients often experience symptoms in the distal colon. Previous analysis of the fecal microbiota from healthy and CRC 20 or IBD patients has revealed different microbial signatures associated with these diseases. We 21 extended these observations of the fecal microbiome to include analysis of the proximal and distal healthy human colon. We used a two-colonoscope approach on subjects that had not undergone standard bowel preparation procedure. This technique allowed us to characterize the native proximal and distal luminal and mucosal microbiome without prior chemical disruption. 16S rRNA gene sequencing was performed on proximal and distal mucosal and luminal biopsies and home-collected stool for 20 healthy individuals. Diversity analysis revealed that each site contained a diverse 27 community, and that a patient's samples were more similar to each other than to that of other individuals. Comparison of all samples to fecal samples taken at exit uncovered that the feces were most similar to samples taken from the distal lumen, likely reflecting the anatomical structure of the colon. Since we could not differentiate sites along the colon based on community structure or community membership alone, we employed the Random Forest machine-learning algorithm to identify key species that distinguish biogeographical sites. Random Forest classification models were built using taxa abundance and sample location and revealed distinct populations that were found in each location. Peptoniphilus, Anaerococcus, Enterobacteraceae, Pseudomonas and Actinomyces were most likely to be found in mucosal samples versus luminal samples (AUC = 0.925). The classification model performed well (AUC = 0.912) when classifying mucosal samples into proximal or distal sides, but separating luminal samples from each side proved more challenging (AUC = 0.755). The distal mucosa was found to have high populations of Finegoldia, Murdochiella and Porphyromonas. Proximal and distal luminal samples were comprised of many of the same taxa. likely reflecting the fact that stool moves along the colon from the proximal to distal end. By
sampling the unprepped human colon, our results have identified distinct bacterial populations
native to the proximal and distal sides. Further investigation of these bacteria may elucidate if and
how these groups contribute to different pathogenesis processes on the respective sides of the colon.

45 Introduction

The human colon is an ecosystem comprised of several different microenvironments inhabited by resident bacterial members of the microbiome. Concentrations of oxygen, water and anti-microbial peptides change along the gut axis and influence what populations of microbes reside in each location. Microenvironments differ not only longitutinally along the colon, but latitudinally from the epithelium to mucosa to intestinal lumen, offering several sites for different microbial communities to flourish. The identity of these specific microbes and communities are important for understanding the etiology of complex colon diseases such as Inflammatory Bowel Disease (IBD) and Colorectal Cancer (CRC). IBD and CRC are known to be preceded or accelerated by perturbations in gut microbes ((1), (2), (3)). The severity, symptoms, morbidity and mortality of these diseases is known to vary based upon the biogeographical location in which they occur. For instance, CRC tumors that arise on the distal side of the colon are infiltrating lesions that present with painful symptoms ((4)). In contrast, 47% of CRCs are caused by proximal-sided colon tumors that are sessile and form along the wall of the colon, often remaining asymptomatic until advanced carcinogenesis ((4)). The distal and proximal sides of the colon differ in the amount of inflammation present and the genomic instability of precancerous cells, respectively, in addition to variation in the previously mentioned chemical gradients ((1), (5), (6)). In IBD patients, disease flares in the distal colon are usually indicative of ulcerative colitis (UC), whereas Crohn's disease (CD) patients typically experience disease in the small intestine, ileum and proximal colon ((2)). UC presents as large and highly inflammed mucosal ulcers, where as CD lesions are often smaller and have areas of normal tissue distributed amongst the flare ((2)). Thus, given the varied physiology of the proximal-distal axis of the colon and known differences in disease patterns at these sites, symbiotic microbes and their metabolites likely vary as well, and may influence the heterogenous disease prognoses of IBD and CRC. Because CRC is a long-term complication of IBD, the distribution of microbes is important to understanding the pathophysiology of both diseases.

Several recent findings have shown that development and progression of IBD or CRC can be attributed to specific molecular events as a result of interactions between the gut microbiota and 71 human host ((1), (3)). For instance, comparison of the bacteria present on CRC tumors with those 72 found on nearby healthy tissue has identified specific species that are tumor-associated ((7)). These species include the oral pathogens Fusobacterium nucleatum and Porphyromonas asacharolytica. Interestingly, these periodontal pathogens have been highly predictive of whether a patient had CRC tumors or not in our human stool classification studies ((8)). F. nucleatum has also been found to be elevated in the stool and biopsies of patients with IBD as compared to healthy controls (Strauss2011). Furthermore, studies of F. nucleatum isolated from mucosal biopsies showed that more invasive F. nucleatum positively correlates with IBD disease level (Strauss2011). Like many 79 intestinal pathogens, the bacteria appear to have a high-impact despite being lowly-abundant in the community ((2)). The molecular capabilities of these rare taxa may contribute to the colonic disease state. These studies often examined only shed human stool or the small intestine, preventing fine-resolution analysis of paired samples from the proximal and distal sides of the colon. Similarly, comparisons of on- or off-tumor/lesion bacteria rarely have matched tissue from the other side of the colon from the same patient, limiting what conclusions can be drawn about the colonic microbiome overall, let alone at that specific site. Due to these limitations, the contribution of the gut microbiota to IBD and CRC disease location in the colon is largely undefined. Characterizing these communities could provide needed insight into disease etiology, including how the disruption of the healthy community could promote the initiation or proliferation of the distinct proximal and distal CRC tumors or IBD flares. The few existing profiles of the microbial biogeography of the gut have been limited by sample 91 collection methods. The majority of human gut microbiome studies have been performed on whole shed stool or on samples collected during colonoscopy procedures. While the latter method allows investigators to acquire samples from inside the human colon, typically this procedure is preceded

by the use of bowel preparation methods such as the consumption of laxatives to cleanse the bowel.

Bowel preparation is essential for detecting cancerous or precancerous lesions in the colon, but

complicates microbiome profiling as the chemicals strip the bowel of contents and disrupt the

mucosal layer ((9), (10)). As such, what little information we do have about the biogeographical distribution of the microbes in the proximal and distal colon is confounded by the bowel preparation procedure.

Here we aimed to address the limitations of previous studies and effectively identify the microbes 101 specific to the lumen and mucosa of the proximal and distal healthy human colon. Our design used 102 an unprepared colonoscopy technique to sample the natural community of each location of the gut 103 without prior disruption of the native bacteria in 20 healthy volunteers. To address the inherent inter-individual variation in human microbiomes, we used a machine-learning classification algorithm 105 trained on curated 16S rRNA sequencing reads to identify microbes specific to each location. We 106 found that our classification models were able to separate mucosal and lumenal samples as well as differentiate between sides of the colon based on populations of specific microbes. By identifying 108 the specific microbes we are poised to ask if and how the presence or disruption of the microbes at 100 each site contribute to the development of the specific tumor subtypes of CRC in the proximal and 110 distal human colon. 111

Results

Microbial membership and diversity of the proximal and distal colon

Lumenal and mucosal samples were collected from the proximal and distal colon of 20 healthy 114 humans that had not undergone bowel preparation (Figure 1). Participants also collected stool 115 at home one week prior to the procedure. To characterize the bacterial communities present at 116 these sites, 16S rRNA gene sequencing was performed on extracted DNA from each sample. Each 117 site was primarily dominated by Firmicutes and Bacteriodetes (Figure 2A), consistent with known 118 variability in human microbiome research ((11)). Likewise, samples had varying levels of diversity 119 at each site, irrespective of the individual (Figure 2B). For example, the proximal mucosa was more diverse than the distal for some individuals while the opposite was true for others. Therefore we 121 could not identify a clear pattern of changes in microbial diversity along the gut axis. 122

To compare similarity between sides (proximal or distal) or sites (lumen or mucosa), we calculated θ YC distances from OTU abundances and compared these distances for all individuals. Again,

across all patient samples we observed a range of θ YC distances when comparing sample locations (Figure 3A) and again those ranges did not follow a clear pattern on an individual basis. However, when comparing median distances between the proximal lumen and mucosa, the proximal versus distal lumen, the proximal versus distal mucosa, and the distal lumen and mucosa, we found that the proximal lumen and mucosa were most similar to each other than the other samples (P < 0.005, Wilcoxon, BH adjustment).

131 Stool at exit most resembles lumenal samples from the distal colon

Next, we calculated θ YC distances to examine how each sample compared to the home-collected exit 132 stool. Amidst variability between patients, we did identify significantly smaller θYC distance between the distal lumenal sample and the exit stool (Figure 3B, P < 0.05, Wilcoxon, BH adjustment). 134 Furthermore, there was an even larger difference in the comparisons of the distal mucosa to the 135 exit stool, indicating that the mucosa is different from the stool as compared to lumen (P < 0.0005, Wilcoxon, BH adjustment). To determine what factors may be driving the differences seen among 137 the samples, we compared thetaYC distances between samples from all subjects (interpersonal) 138 versus samples from within one subject (intrapersonal). We found that samples from one individual 139 were far more similar to each other than to other study subjects (Figure 3C), consistent with 140 previous human microbiome studies that have sampled multiple sites of the human colon (???, (12), 141 (13)). Thus interpersonal variation between subjects drives the differences between samples more 142 than sample site or location. Overall, the results comparing the structure of the communities suggest that the contents of the distal lumen are most representative of stool at exit, and the microbes 144 remaining on the mucosa are much different. 145

146 Random Forest classification models identify important OTUs on each side

To identify OTUs that were distinct at each biogeographical site, we constructed several Random
Forest models trained using OTU abundances. We built the first model to classify the lumen versus
mucosal samples for the proximal and distal sides, independently (Figure 4A). The constructed
model used 14 features for the proximal and 6 for the distal. The models performed well when
classifying these samples (0.8 and 1.0, respectively). The OTUs that were most predictive of each

site are identified by their greatest mean decrease in accuracy when removed from the model. For distinguishing the proximal lumen and mucosa, OTUs from the Bacteriodes, Actinomyces, Psuedomonas and two OTUs from the Enterobacteraceae genera were differentially abundant 154 (Figure 4B). The model classifying the distal lumen and mucosa identified OTUs from Turicibacter, 155 Finegoldia, Peptoniphilus and two OTUs from the Anaerococcus genera that could distinguish lumen from mucosal samples (Figure 4C). These results indicate that there are fine differences between the 157 different sites of the colon, and that these can be traced down to specific OTUs on each side. 158 Next, we built a model to differentiate the proximal and distal lumenal samples. The model 159 performed best when distinguishing the proximal versus distal mucosa (Figure 5A, AUC = 0.9) 160 compared to the proximal versus distal lumen (AUC = 0.8). OTUs that were differentially abundant between the distal and proximal mucosa included members of the Porphyromonas, Murdochiella, 162 Finegoldia, Anaerococcus and Peptoniphilus genera (Figure 5B). Differentially abundant OTUs of 163 the proximal and distal lumen included three OTUs of the Bacteroides genus, a Clostridium IV OTU and an Oscillibacter OTU (Figure 5C). This analysis found that some of the same OTUs that 165 are distinct between the mucosa and lumen also helped to differentiate between the two sides- such 166 as Anaerococcus and Finegoldia. 167

168 Bacterial OTUs associated with cancer are found in healthy individuals

Given that specific bacterial species have been associated with colorectal cancer and IBD, we 169 probed our sample set for these OTUs. Among our 100 samples, the most frequent sequence 170 associated with the Fusobacterium genus was OTU179, which aligns via BLASTn to Fusobacterium 171 nucleatum subsp animalis (XX% over full length). This is the only species of Fusobacterium known 172 to have oncogenic properties and be found on the surfaces of colorectal cancer tumors. ((14)). 173 The Fusibacterium positive samples were located in x\% of the the proximal and X\% of the distal 174 mucosa and represented as much as 1% of any sample (Figure 6A). OTU152 was similar to the 175 members of the *Porphyromonas* genus and the most frequent sequence in that OTU aligned to 176 Porphyromonas asacharolytica (X\% over full length), another bacterium commonly detected and 177 isolated from colorectal tumors. OTU152 was only detected on the distal mucosa, and in fact was one of the OTUs the classification model identified as separating distal and proximal sides (Figure

6B). Among the samples that were positive for *P. asacharolytica*, the relative abundances for this
OTU ranged from 0.01% - 16%. Thus, disease-associated OTUs could be found in our sample set of
lead to be abundances for this

183 Discussion

Here we identified bacterial taxa that were specific to the lumen and mucosa of the proximal and 184 distal sides of the human colon from samples collected during unprepared colonoscopy. We found 185 that all locations contained a range of phyla and a range of diversity, but that there was a wide 186 variability between subjects. Pairwise comparisons of each of the sites revealed that the proximal 187 mucosa and lumen were most similar to each other. Further, comparison of colonoscopy-collected 188 samples with samples collected from stool at home showed that the distal lumen is most similar to 189 stool at exit. Random Forest models built on OTU relative abundances from each sample identified 190 microbes that are particular to each location of the colon. Finally, we were able to detect some 191 bacterial OTUs associated with colonic disease in our healthy patient cohort. Using unprepped 192 colonoscopy and machine learning, we have identified bacterial phyla specific to the healthy proximal 193 and distal human colon. 194

When examining the relative abundance of the different phyla at each site, there was a wide amount 195 of variation for each phyla with communities primarily dominated by the *Bacteriodes* and *Firmicutes*. 196 This likely reflects not only the variability between human subjects, casued by differences in age, gender, diet, but also reports of biogeographical "patchiness" in the gut microbiome. Several studies 198 have noted that the bacteria recoverable from the same mucosal sample location can be vastly 199 different when the samples are taken just 1 cm away from each other (15). Similar patchiness is 200 also observed in lumenal contents and fecal samples themselves; there is observed separation of 201 different interacting microbes along the length of a stool sample, for instance (16). That said, across 202 our samples the mucosal samples harbor more Proteobacteria, consistent with previous studies 203 comparing mucosal swabs to lumenal content in humans (5). Hence, the conclusions we can draw 204 from phyla analysis are likely impacted by patchiness between subjects. 205

To get around the noisiness from a diverse set of samples, we built a Random Forest model to identify

microbes specific to each side. For each comparison we identified top X OTUs that were strongly 207 predictive of one site or another. Generally, OTUs identified in each location were consistent with known physiological gradients along the gut axis (6). For instance, the proximal mucosa contains 209 the highest oxygen concentrations of the colon and harbored mucosa-associated facultative anaerobes 210 such as Actinomyces and Enterobacteraceae and aerobic Psuedomonas. The distal mucosa was far 211 more likely to host strictly anaerobic species such as Porphyromonas, Anaerococcus, Finegoldia 212 and Peptoniphilus. The model was less effective at classifying the proximal and distal lumenal 213 contents, probably because the samples are arguably composed of the same bacteria but differ in 214 water content. 215

We detected F. nucleatum and P. asacharolytica in 8 and 5 of our subjects, respectively. These 216 bacteria have been shown to be predictive of colorectal cancer in humans (8) and have oncogenic 217 properties in cell culture and in mice (17). Interestingly, while F. nucleatum was found on both sides 218 of the colon, P. asacharolytica was only detected in the distal mucosa. Not much is known about the 219 distribution of P. asacharolytica but given its documented anaerobic characteristics and asacharolytic 220 metabolism, it might not be surprising that it resides in the less-oygen-rich and proteinaceous distal 221 mucosa ((5)). In studies examining bacteria on colorectal cancer tumors, F. nucleatum is more 222 commonly detected on proximal-sided tumors, and distribution of F. nucleatum decreases along 223 the colon to rectum ((18)). Of the 8 (40%) individuals positive for F. nucleatum in this study, the 224 bacterium was spread across the proximal mucosa, distal lumen and distal mucosa. The presence of 225 F. nucleatum in a healthy individual is not necessarily linked to the development of future colorectal 226 cancers (cite). Data examining bacterial biofilms on CRC tumors suggests that Fusobacteria species 227 are more commonly found on proximal tumors and in biofilms, indicating that it is not only 228 the presence of the bacteria but the organization of the tumor community that contributes to 229 Fusobacterium's role in tumorigenesis ((7)). Finally, Fusobacterium and Porphyromonas species have 230 been known to not only co-occur on CRC tumors but also to synergistically promote tumorigenesis 231 in an oral cancer model ((19), (20)). Thus, further analysis of the distribution and activities of these 232 pathogens may elucidate a mechanism for development of IBD or CRC subtypes in the proximal or 233 distal colon. 234

The Fusobacterium species nucleatum and varium have been commonly isolated from mucosal

biopsies of patients with IBD (cite). Laboratory experiments with these isolates have shown that disease-isolated F. nucleatum are more invasive and stimulate more TNF-a production than strains 237 from healthy individuals ((21)), suggesting the bacteria may increase inflammation in the gut as well 238 (cite - Dharmani 2011). F. varium isolated from UC patients caused colonic ulcers in an experimental 239 mouse model (Ohkusa 2003). F. varium was only detected in 3 of our study participants and two of those samples were isolated form the proximal mucosa (cite figure). Although there is low 241 occurrance in our study, F. varium is most commonly isolated from UC patient biopsies from the 242 ileum or cecum (cite), suggesting this species may exhibit preference for the different environmental 243 conditions of these gastrointestinal sites. Further work will assess how gut environment may select for species which then cause localized disease. 245

Specific comparisons of our findings to previously published gut biogeography studies are additionally 246 confounded by the use of bowel preparation methods in most other studies. A rare report of a 247 matched-colonoscopy study sampled 18 patient's colonic mucosa and lumenal contents prior to 248 and after bowel cleansing ((22)). This group found that mucosal and lumenal samples were 240 distinguishable prior to bowel cleansing, but that bowel preparation resulted in an increase in shared 250 OTUs between each site ((22)). Bowel cleansing not only made the samples harder to distinguish, 251 it resulted in decreases in diversity across sites. Further, the differences were not great enough to 252 overcome interpersonal differences between subjects. Bowel preparation clearly induces bias into 253 the microbes recovered from sampling the lumen or mucosa of a prepared bowel. Thus our findings 254 in this study are strengthened by the lack of bowel preparation.

By revealing specific differences in microbial populations at each location in the gut via sampling an unprepared bowel, we can begin to form hypothesies about how specific host-microbe interactions can affect disease progression of proximal and distal CRC and IBD subtypes. To this point, 16S rRNA gene sequecing community profiling studies do not provide enough information to fully probe these questions. In particular, 16S sequencing cannot not profile the host characteristics at each site. Combining the unprepared colonoscopy approach with analysis of multi-omic sequencing data may be useful in further characterizing host-microbiome interactions along the gut axis for both health and disease.

264 Acknowledgments

We thank all the individuals who volunteered for the study. This work was supported by the Rose and
Lawrence C. Page Foundation (DKT). We would also like to thank Brian Kleiner, Chelsea Crofoot,
and Kirk Herman for their roles in study coordination, subject recruitment, sample collection and
sample processing.

269 Methods

270 Human subjects

The procedures in this study and consent were approved by the Institutional Review Board at the
University of Michigan Health System with protocol number XXXX. Subjects were recruited using
the online recruitment platform and were pre-screened prior to enrollment in the study. Exclusion
criteria included: use of asprin or NSAIDs within 7 days, use of antibiotics within 3 months,
current use of anticoagulants, known allergies to Fentanyl or Benadryl, prior history of colon disease,
diabetes, abdominal surgery, respiratory, liver, kidney or brain impairments, undergoing current
chemotherapy or radiation treatment and subjects that were pregnant or trying to conceive. 20
subjects that met the criteria were selected and provided signed informed consent prior to the
procedure. There were 13 female and 7 male subjects ranging in age from 25 to 64.

280 Sample collection

At a baseline visit, subjects were consented and given a home collection stool kit (Source of kit supplies). At least one week prior to the scheduled colonoscopy, subjects were to collect whole stool at home and ship the samples to the University on ice. Notably, subjects did not undergo any bowel preparation method prior to sampling. On the procedure day, subjects reported to the Michigan Clinical Research Unit at the University of Michigan Health System. Patients were consciously sedated using Fentanyl, Versed and/or Benadryl as appropriate. A flexible sigmoidoscope was first inserted about 25cm into the colon and endoscopy brush used to collect lumenal/stool contents. Two lumenal samples were collected and the contents immediately deposited into RNAlater (source) and flash-frozen in liquid nitrogen. The brushes were withdrawn and biopsy forceps were used to

collect mucosal biopsies on sections of the colon that were pink and free of stool matter. Three
mucosal biopsies were collected and flash-frozen in RNAlater. These samples comprised the distal
or distal colon samples. The sigmoidoscope was then withdrawn and a pediatric colonoscope was
inserted to reach the ascending colon. Samples were then collected as in the distal colon and the
colonoscope withdrawn. All samples were stored at -80 C until study completion.

295 Sample processing, sequencing and analysis

DNA extraction was performed using the PowerMicrobiome DNA/RNA Isolation Kit (MO BIO 296 Laboratories). For tissue biopsies, Bond-Breaker TCEP solution (Fisher) and 2.8mm ceramic beads 297 (MO BIO Laboratories) were added to the bead beating step to enhance DNA recovery from mucosal samples. The resulting DNA was used as template for amplification of the V4 region of the 16S 299 rRNA gene and fragments were sequenced on an Illumina MiSeq as previously described ((23)). 300 Sequences were curated using the mothur software as described previously ((24)). The sequences were assigned taxonomic classification using a naive Bayesian classifier trained using a 16S rRNA 302 gene training set from the Ribosomal Database Project (RDP) ((25)) and clustered into operational 303 taxonomic units (OTUs) based on a 97% similarity cutoff. Sequencing and analysis of a mock community revealed the error rate to be X%. Samples were rarefied to 4231 sequences per sample 305 in order to reduce uneven sampling bias. 306 Diversity analysis was performed using the Simpson diversity calculator and theta YC calculator 307 metrics in mother version $\langle 3.2? \rangle$ ((24)). ThetaYC distances were calculated to determine the 308 dissimilarity between two samples. Random Forest classification models were built using the randomForest R package and resultant models were used to identify the OTUs that were most 310 important for classifying each location ((26)). To get species-level information about sequences of 311 interest, sequences were aligned using blastn and the species name was only used if the identity 312 score was $\geq 99\%$. 313

314 Statistical analysis

Differences in community membership at the phyla level were tested using the analysis of molecular variance (AMOVA) metric in mothur. Differences in thetaYC distances by location were tested

- using the Wilcoxon rank-sum test adjusted for multiple comparisons using the Benjamini-Hochberg procedure.
- Data availability
- 320 Github repository

Figures Figures

Figure 1

Sampling strategy. A flexible sigmoidoscope was used to sample the distal colonic luminal contents and mucosa. The scope was inserted ~ 25cm into the subject and endoscopy brushes were used to sample the luminal contents (green star). A separate set of biopsy forceps was used to sample the distal mucosa (blue star). The sigmoidoscope was removed. A pediatric colonoscope was inserted and used to access the proximal colon. Biopsies were taken of the proximal luminal contents and mucosa as described. One week prior to the procedure stool was collected at home and sent into the laboratory. Representative images from one individual are shown.

Microbial membership and diversity of the proximal and distal human colon. A) Relative abundance of the top five bacterial phyla in each sampling site. Each box represents the median and confidence intervals. B) Simpson diversity of the microbial communities at each location. The lines represent the median values.

- $_{336}$ Distances of microbial community structure between sites of the gut. ThetaYC distances are shown
- for interpersonal similarities between two sites each point represents one individual. In (A),
- comparisons of the proximal and distal mucosal and lumen are shown. In (B), comparisons of each
- 339 site to the exit stool are shown.

Random Forest classifies the mucosa and lumen of each side of the colon. A) Receiver Operator
Characteristic curves are shown for the 10-fold cross validation of the Random Forest model
classifying lumen and mucosal samples for the distal and proximal sides of the colon. (B) Top five
OTUs that are most important for the classification model for the distal mucosa and lumen (B) and
the proximal mucosa and lumen (C).

Random Forest classifies the distal and proximal sides of the colon. A) Receiver Operator Characteristic curves are shown for the 10-fold cross validation of the Random Forest model classifying
distal lumen versus proximal lumen (orange) and distal mucosa vs proximal mucosa (green). (B)
Top five OTUs that are most important for the classification model for the distal and proximal
mucosa (B) and the distal and proximal lumen (C).

- Location and abundance of cancer-associated OTUs. Relative abundance was calculated and plotted
- by sample site for each OTU of interest: (A) Fusobacterium nucleatum and (B) Porphyromonas
- 355 asacharolytica

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