

# Patterns in the Juan Fernandez fur seal faecal microbiome

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## Abstract

As apex predators, pinnipeds are considered to be useful bioindicators of marine and coastal environments. Endemic to a small archipelago in the South Pacific, the Juan Fernandez fur seal (JFFS) is one of the less-studied members of the pinniped family Otariidae. This study aimed to characterise the faecal microbiome of the JFFS for the first time, in order to establish a baseline for future studies of host-microbial-environment interactions and monitoring programs. During two consecutive reproductive seasons, 57 faecal samples were collected from 7 different JFFS colonies within the Juan Fernandez Archipelago, Chile. Bacterial composition and abundance were characterised by sequencing the V4 region of the 16S rRNA gene. The overall microbiome composition was dominated by five phyla: Firmicutes (40 %  $\pm$  24), Fusobacteria (30 %  $\pm$  17), Bacteroidetes (22 %  $\pm$  10), Proteobacteria (6 %  $\pm$  4) and Actinobacteria (2 %  $\pm$  3). Alpha diversity was higher in Tierras Blancas. However, location was not found to be a dominant driver of microbial composition. Interestingly, the strongest signal in the data was a negative association between the genera *Peptoclostridium* and *Fusobacterium*, which explained 29.7 % of the total microbial composition variability between samples. The genus *Peptoclostridium* has not been reported in other pinniped studies and its role here is unclear, with interpretation challenging due to a lack of information regarding microbiome functionality in marine mammals. As a first insight into the JFFS faecal microbiome, these results contribute towards our understanding of the natural microbial diversity and composition in free-ranging pinnipeds.

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## Abstract

As apex predators, pinnipeds are considered to be useful bioindicators of marine and coastal environments. Endemic to a small archipelago in the South Pacific, the Juan Fernandez fur seal (JFFS) is one of the less-studied members of the pinniped family *Otariidae*. This study aimed to characterise the faecal microbiome of the JFFS for the first time, in order to establish a baseline for future studies of host-microbial-environment interactions and monitoring programs. During two consecutive reproductive seasons, 57 faecal samples were collected from 7 different JFFS colonies within the Juan Fernandez Archipelago, Chile. Bacterial composition and abundance were characterised by sequencing the V4 region of the 16S rRNA gene. The overall microbiome composition was dominated by five phyla: *Firmicutes* (40% ± 24), *Fusobacteria* (30% ± 17), *Bacteroidetes* (22% ± 10), *Proteobacteria* (6% ± 4) and *Actinobacteria* (2% ± 3). Alpha diversity was higher in Tierras Blancas. However, location was not found to be a dominant driver of microbial composition. Interestingly, the strongest signal in the data was a negative association between the genera *Peptoclostridium* and *Fusobacterium*, which explained 29.7% of the total microbial composition variability between samples. The genus *Peptoclostridium* has not been reported in other pinniped studies and its role here is unclear, with interpretation challenging due to a lack of information regarding microbiome functionality in marine mammals. As a first insight into the JFFS faecal microbiome, these results contribute towards our understanding of the natural microbial diversity and composition in free-ranging pinnipeds.

**Keywords**— *Arctocephoca philippii*, scatology, microbiome, pinnipeds

## 1 Introduction

Marine environments are complex and interconnected systems subject to various environmental impacts. Pollution, climate change, disruption of the food network and pathogen dissemination are a few examples of problems currently affecting ocean integrity and function (Halpern et al., 2019). Integrated approaches at the macro- and micro-ecological levels are needed to properly

29 understand and manage environmental threats in these kinds of complex systems. Identifica-  
30 tion and investigation of potential environmental sentinel species such as marine mammals can  
31 provide a better understanding of the deterioration or improvement of ocean health (Bossart,  
32 2011; Hazen et al., 2019). However, to effectively use wild populations as sentinels, it is first  
33 necessary to establish a baseline.

34 In the last couple of decades, the study of the microbiome in wild populations has increased,  
35 due to the profound impact of host-microbial interactions on host physiology and the grow-  
36 ing affordability of sequencing technologies (Redford et al., 2012; Trevelline et al., 2019). The  
37 gastrointestinal tract, especially the colon, is recognised as one of the largest microbial reser-  
38 voirs (O’Hara and Shanahan, 2006). This microbial community fulfils essential functions in  
39 digestion, metabolic activity and immunity, and differences in species composition and abun-  
40 dance can therefore provide much information about the host organism. For example, following  
41 its initial acquisition during birth and lactation, the microbiome is constantly modified by factors  
42 such as age, sex and diet (Ley et al., 2008b,c; Nicholson et al., 2012). Similar factors shaping the  
43 gut microbiome in terrestrial mammals influence that of marine mammals (Nelson et al., 2013b;  
44 Pacheco-Sandoval et al., 2019; Smith et al., 2013; Stoffel et al., 2020). However, studies have  
45 also shown substantial differences between marine and terrestrial mammal gut microbiomes,  
46 even when these two groups share a similar diet (e.g. herbivore, carnivore) (Bik et al., 2016;  
47 Nelson et al., 2013a). Thus, even though research into the microbiome of terrestrial mammals is  
48 at a relatively advanced stage, this information cannot be easily extrapolated to marine mam-  
49 mals whose microbiomes remain poorly understood particularly, those in non-captive, natural  
50 populations. Consistent characterisation of the core microbiome of these populations is there-  
51 fore required as a fundamental baseline before we can attempt to understand its functions, roles,  
52 interactions and possible uses (Shade and Handelsman, 2012).

53 The Juan Fernandez fur seal (*Arctophoca philippii philippii*) (JFFS) is a marine mammal en-  
54 demic to the Juan Fernandez Archipelago, a group of islands located in the middle of the Pacific  
55 Ocean 600 km away from the Chilean continental coast (Fig. 1). The archipelago is a hotspot  
56 for biodiversity with a high number of endemic marine species, including the JFFS (Friedlander  
57 et al., 2016; Pompa et al., 2011). These fur seals are the only native mammals to the archipelago  
58 and like other pinnipeds occupy upper trophic levels in the marine food web (Trites, 2019; Ochoa

59 Acuna and Francis, 1995). Their feeding behaviour, lifespan, fat storage, and their amphibian  
60 lifestyle, which links marine and coastal environments, are some of the characteristics that make  
61 this species a great candidate to act as a marine bioindicator. However, despite showing a sig-  
62 nificant population recovery since the late 1960s and becoming an icon for local tourism, little  
63 is known about this species. This study aimed to characterise the JFFS faecal microbiome for  
64 the first time, as a baseline for understanding the host-microbial interactions in this species. To  
65 investigate, we performed sequencing of the 16S rRNA gene, a highly conserved region of the  
66 bacterial genome, which provides a reliable overview of bacterial community composition.

## 67 **2 Methods**

### 68 **2.1 Ethics statement**

69 All faecal samples were collected from the environment in a non-invasive manner. Disturbance  
70 of the colonies was kept to a minimum and no animal was handled or harmed in the process.  
71 Permits for the collection of samples were given by CONAF (Certificate 009217) and SER-  
72 NAPESCA (R.E.X.N 43). Permission for importation of samples into the United Kingdom was  
73 also obtained (ITIMP16.1158).

### 74 **2.2 Sample collection**

75 Faecal samples were collected from seven reproductive colonies of Juan Fernandez fur seals situ-  
76 ated throughout the Juan Fernandez archipelago, Chile (coordinates: 33°38'29"S 78°50'28"W)  
77 (Fig. 2). Six of the seven colonies included in this study were located on Robinson Crusoe is-  
78 land: El Arenal (EA), Bahia El Padre (BP), Piedra Carvajal (PC), Punta Trueno (PT), Tierras  
79 Blancas (TB) and Vaquera (V). One colony was located on Santa Clara island (SC). Samples  
80 were collected during two consecutive reproductive seasons (2017 and 2018), which take place  
81 between mid-January to the end of February. Collection of samples took place before noon to  
82 limit sun exposure. A disposable wooden spatula was used to expose the centre of the faeces. .  
83 Using a sterile Copan FLOQSwab, a sample from the core of the faeces was placed into RNAlater  
84 (Sigma-Aldrich) (Blekhman et al., 2016; Vlčková et al., 2012). No distinction of sex and age

85 was made at the time of sample collection. Samples were stored at  $-20^{\circ}\text{C}$  within 32 hours post  
86 collection for 1-2 months until arrival in the laboratory, where they were transferred to  $-80^{\circ}\text{C}$   
87 until further analysis.

## 88 **2.3 DNA extraction and sequencing**

89 Samples were processed in two batches according to the year of collection (2017 and 2018 re-  
90 spectively). Due to the possible batch effect introduced by processing samples in different years,  
91 comparisons between years of collection will not be explored in this study. Samples were thawed  
92 on ice and centrifuged at  $10,000 \times g$  for 15 min to pellet the sample out of RNAlater. Genomic  
93 DNA was extracted from each pelleted sample (approx. 180 micrograms) using the MO BIO  
94 PowerSoil DNA Isolation kit (QIAGEN) according to the manufacturers instructions. Isolated  
95 DNA was quantified on a Qubit fluorometer (Invitrogen). The bacterial 16S rRNA gene was  
96 PCR amplified targeting a 250 bp region covering the V4 variable region. PCR amplification,  
97 barcode tagging and library preparation was performed according Kozich *et al.* (Kozich *et al.*,  
98 2013). Libraries were constructed using the TrueSeq DNA kit and sequenced on a MiSeq plat-  
99 form (Illumina). The read length target changed between the two sampling years. Sequencing  
100 was performed using the v2 chemistry producing  $2 \times 250$  bp paired-end reads in the 2017 samples  
101 while the 2018 sequences were  $2 \times 150$  bp paired-end reads.

## 102 **2.4 Sequence data analysis and taxonomic classification**

103 Raw sequence quality was manually assessed with FastQC v. 0.11.5 (Simon Andrews, 2010).  
104 All 57 samples contained reads of consistent length (respective to the sequencing year) and the  
105 average read quality score was above 30. . A drop in base quality was observed at the ends of  
106 reads (4 - 5 and 8 - 10 respectively). Demultiplexed raw sequences were imported into QIIME2-  
107 2019.10 (Bolyen *et al.*, 2019) where quality control, de-replication, read truncation and paired  
108 read merging was performed using the DADA2 (Divisive Amplicon Denoising Algorithm) qiime2  
109 plugin (Callahan *et al.*, 2016). Instead of generating operational taxonomic units (OTUs) by  
110 clustering sequences based on similarity, the final output of DADA2 is a table with exact sequence  
111 variants also known as amplicon sequence variants (ASVs), which are generated by modelling

112 and correcting Illumina sequencing errors. This step was carried out separately according to  
113 the year of collection. However, in order to normalise between datasets, the 250 bp reads  
114 produced from 2017 samples were truncated so that the paired reads matched the length of the  
115 paired reads from 2018 samples. To confirm consistency in paired read lengths between the  
116 two years, representative sequences generated from both years were aligned in Geneious Prime  
117 2020.0.5 (<https://www.geneious.com>) by Multiple Alignment using the Fast Fourier Transform  
118 (MAFFT) plug-in with default settings (Kato and Standley, 2013) and then assessed by eye.

119 Next, a mid-point rooted, approximately-maximum-likelihood phylogenetic tree for diver-  
120 sity analysis was generated using the qiime2 phylogeny plug-in which uses MAFFT and the  
121 FastTree program (Price et al., 2010). Finally, taxonomies were assigned to the ASVs using  
122 a 16S-V4-specific classifier trained against the Silva132 database clustered at 99% sequence  
123 similarity (Quast et al., 2013).

## 124 **2.5 Data processing and statistical analysis**

125 Statistical analysis was performed in duplicate, once using all available data and again with data  
126 corresponding to the core microbiome only. The core microbiome was defined here as all the  
127 ASVs present in at least 50 percent of the samples. Data processing and statistical analysis were  
128 carried out in R version 3.6.0 (R Core Team, 2019). To prepare the data by identifying unas-  
129 signed ASVs and removing contaminants and samples with insufficient depth of sampling prior  
130 to analysis, multiple filtering steps were applied to the data using the phyloseq package (Mc-  
131 Murdie and Holmes, 2013). 1) Unassigned ASVs at the Kingdom level, were manually inspected  
132 with the Basic Local Alignment Search Tool (BLAST) (Altschul et al., 1990) before filtering  
133 based on both BLAST results (those with non-bacterial matches) and prevalence (ambiguous  
134 taxonomy at the phylum level with a prevalence of 1 and total abundance less than 5 reads).  
135 2) Based on the rarefaction curve (Supplementary Fig. 1), 3 samples were identified as hav-  
136 ing insufficient depth of sampling and were therefore removed from the statistical analysis. A  
137 threshold of 13,980 reads was used as a cut-off. Removed samples were identified as 17JFFS16  
138 (BP, 4463 counts), 17JFFS23 (TB, 2602 counts) and 17JFFS23 (EA, 2042 counts). 3) Possible  
139 contaminant signals were also removed by running a correlation analysis and comparing clus-  
140 ters with a list of previously identified reagent contaminants (Salter et al., 2014). 4) Finally,

141 the data was rarefied using the same threshold used for filtering samples (Supplementary Table  
142 1) (McKnight et al., 2019).

143 The overall microbiota composition was characterised by summing the non-normalized read  
144 counts and obtaining the relative abundance at different taxonomic levels.

### 145 **2.5.1 Alpha diversity**

146 Estimates of within-sample diversity (alpha diversity) were calculated using the phyloseq pack-  
147 age. Three indices were included: a richness estimator, which estimates the total number of  
148 species in each sample (Chao1) and two different diversity estimators (Shannon-Weiner and  
149 Simpson index). The latter two approaches consider richness as well as abundance. However,  
150 the effect of richness and rare species strongly impact the Shannon-Weiner index, whereas the  
151 Simpson index is mainly influenced by evenness and common species.

152 Non-rarefied data was used to explore the alpha diversity. To compare locations, a one-way  
153 analysis of variance test (ANOVA) or a non parametric Kruskal Wallis test were performed for  
154 each estimate. ANOVA assumptions were tested by visualisation of the data and statistical test-  
155 ing. A Shapiro-Wilk test was used to confirm normality and a Levenes test for heteroscedasticity.  
156 When exploring Shannon-Weiner and Simpson indices sample 18JFFS23 (SC) was identified as  
157 an outlier and was removed for these indices only. Finally, data visualisation suggested samples  
158 collected from TB differed from the other locations thus, a post-hoc analysis was performed with  
159 Dunnetts or the non-parametric Dunns test to compare each location to TB. Samples from PC,  
160 PT and V were not included in the location comparison due to their limited sample size (n =  
161 1).

### 162 **2.5.2 Beta diversity**

163 To investigate variation between samples (beta diversity) two different distances were calculated  
164 using the rarefied full as well as the core datasets. Bray-Curtis dissimilarity distance was use  
165 to look at the differences between samples based on the ASVs abundances. Weighted UniFraq  
166 distance was used to explore the phylogentic divergence between ASVs by also taking into  
167 account the abundance of these (with an emphasis on dominant ASVs). Respective distance  
168 matrices were visualised using principal coordinate analysis plots (PCoA).

169 To further explore the clustering of samples (Cluster 1 versus Cluster 2) observed in the  
170 Bray-Curtis PCoA, a permutational multivariate ANOVA (PERMANOVA) was computed with  
171 999 permutations to test for statistically significant differences between the clusters. Finally, a  
172 Similarity Percentages breakdown analysis (SIMPER) was performed between the clusters to  
173 identify the genera that most contributed to the difference between clusters. Genera that highly  
174 contributed to dissimilarities between groups were further explored with the non parametric  
175 Mann-Whitney U test.

176 Spearman rank correlation coefficient ( $\rho$ ) was used to explore any possible associations be-  
177 tween the different taxa and also between the first two components of the Bray-Curtis ordination  
178 analysis. Correlations were visualised in a correlation matrix plot and only those significantly  
179 and strongly correlated (Rho ( $\rho \geq |0.6|$ )) were explored further. For this method, only the core  
180 microbiome dataset was used at the genus level.

## 181 **3 Results**

182 Following removal of low quality sequences and merging the 2017 and 2018 datasets, a total  
183 of 2,074,038 paired reads, grouped into 595 ASVs were imported into R studio for statistical  
184 analysis. A total of 54 samples, with 2,062,763 sequences clustered into 558 ASVs remained after  
185 the filtering steps (Supplementary Table 1). Three samples were removed from the analysis due  
186 to rarefaction analysis indicating insufficient depth of sequencing. The rarefied dataset ended  
187 up with 518 ASVs and a total of 754,974 reads.

### 188 **3.1 Composition of the Juan Fernandez fur seal faecal micro-** 189 **biome**

190 A total of 10 bacterial phyla were detected in the faeces of the JFFSs. From the total ASV  
191 counts *Firmicutes* (41.9%), *Fusobacteria* (28.2%), *Bacteroidetes* (22.1%), *Proteobacteria* (5.5%)  
192 and *Actinobacteria* (1.5%) dominated the bacterial composition. The total ASV counts from  
193 individual samples were very similar to the average relative abundance: *Firmicutes* (40% *pm*  
194 24), *Fusobacteria* (30% *pm* 17), *Bacteroidetes* (22% *pm* 10), *Proteobacteria* (6% *pm* 4) and  
195 *Actinobacteria* (2% *pm* 3) (Supplementary Table 2). Eighty-two bacterial families could be

196 assigned, of which 14 had a relative abundance  $\geq 1\%$  of the total ASV count. Five bacte-  
197 rial families accounted for 78.5% of all read counts: *Fusobacteriaceae* (28.2%) belonging to the  
198 phylum *Fusobacteria*, *Bacteroidaceae* (15.5%) from the phylum *Bacteroidetes*, and *Ruminococ-*  
199 *caceae* (15.0%), *Lachnospiraceae* (10.4%) and *Peptostreptococcaceae* (9.4%) from the phylum  
200 *Firmicutes* (Fig. 3A and 3B, Supplementary Table 3). Forty-six ASVs were present in at least  
201 50% of the samples (Supplementary Table 4). While fourteen ASVs were present in  $> 90\%$  of  
202 samples, only three ASVs were present in all the samples, all of which were assigned to the genus  
203 *Fusobacterium* (14.9%, 6.5% and 3.7% of the total reads respectively) (Table 1).

## 204 **3.2 Alpha diversity**

205 Three alpha diversity indices (Chao1, Shannon-Weiner and Simpson) were used to compare  
206 within-sample diversity between locations (Supplementary Table 5). Despite a clear trend, the  
207 one-way ANOVA results showed no significant differences between locations according to Chao  
208 1 index ( $F(3/47) = 2.45$ ,  $p = 0.07$ ,  $ges = 0.08$ ) and Shannon-Weiner index ( $F(3/46) = 2.65$ ,  $p =$   
209  $0.06$ ,  $ges = 0.09$ ). The Simpson index ( $\chi^2 = 8.26$ ,  $p < 0.05$ ,  $ges =$  not provided) on the  
210 other hand, showed a significant difference between locations. Post-hoc Dunnett's and Dunnett's tests  
211 consistently showed that samples from TB had higher mean and mean rank values (respectively)  
212 than the other locations, especially when compared to Tierras Blancas. Differences in sample  
213 group sizes could explain the lack of statistical power (Fig. 4, Supplementary Fig. 2).

## 214 **3.3 Beta diversity**

215 Based on weighted Unifrac dissimilarity distance, 51.0% (full dataset) and 53.8% (core dataset)  
216 of the total variation between samples could be explained by the first principal component (PC1).  
217 No clustering of individual samples by location or year of collection was observed. Similarly,  
218 Bray-Curtis dissimilarity, which quantifies the differences in ASV abundance, found that the first  
219 principal components in both the full and core datasets explained 23.9% and 29.8% of the total  
220 variation respectively. In both data sets, a group of samples (cluster 2) were clearly separated  
221 from the main cluster (cluster 1) along PC1 (Fig. 5, Supplementary Fig 3). Based on the  
222 relative average abundance of the dominant phyla, evident differences in the overall microbial

223 composition were visualised between the two clusters (Fig. 6). PERMANOVA evidenced a  
224 significant difference in the microbial composition between the two clusters. This was consistent  
225 in both full ( $F = 10.1$ ,  $\text{Pr}(>F) = 0.001$ ,  $R^2 = 16.3\%$ ) and core datasets ( $F = 13.6$ ,  $\text{Pr}(>F) =$   
226  $0.001$ ,  $R^2 = 20.88\%$ ). SIMPER analysis identified five genera that together contributed 71% to  
227 the observed compositional difference between the clusters. As expected, both *Fusobacterium*  
228 and *Peptoclostridium* were the largest contributors (24 and 25% respectively). Furthermore, the  
229 abundance of *Fusobacterium* and *Peptoclostridium* were significantly different between clusters.  
230 Full results of the SIMPER and Mann-Whitney U-tests are summarised in Table 2.

### 231 3.4 Correlation analysis

232 Spearman correlation analysis revealed that the genera *Bacteroides*, *Fusobacterium* and *Pep-*  
233 *toclostridium* were strong drivers of PC1 in both Bray-Curtis and Weighted Unifrac PCoA  
234 analyses. In addition, the genera *Ruminoclostridium 9* and Ruminococcaceae NK4A214 were  
235 also found to be influential for PC1 in Bray-Curtis analysis (Fig. 7, Supplementary Table 6).  
236 PCoA analyses showed strong negative correlations between PC1 and *Bacteroides* (Bray-Curtis,  
237  $\rho = -0.67$ ,  $p \leq 0.001$ ); and between PC1 and *Fusobacterium* (Bray-Curtis,  $\rho = -0.92$ ,  $p \leq 0.001$   
238 and weighted Unifrac,  $\rho = -0.94$ ,  $p \leq 0.001$ ). *Peptoclostridium*, on the other hand, was posi-  
239 tively correlated with PC1 (Bray-Curtis,  $\rho = 0.81$ ,  $p \leq 0.001$ , and weighted Unifrac,  $\rho = -0.75$ ,  
240  $p \leq 0.001$ ).

## 241 4 Discussion

242 Marine mammal microbiome studies of free-ranging, wild populations are rare, with many of  
243 these studies being limited to a small number of individuals. Instead, most studies of marine  
244 mammals have relied on data from dead or captive animals. To our knowledge, this is one of  
245 the most extensive studies of the faecal microbiome in free-ranging pinnipeds and the first of  
246 JFFS. Our approach focused on characterising the core members of the JFFS faecal microbiome,  
247 identified at the genus level, providing a baseline for understanding host-microbial interactions  
248 in this species. However, interpreting unexpected phenomena in a dataset such as ours is made  
249 difficult by a lack of consistent literature, as well as the various uncontrollable factors influencing

250 wild populations.

251 Consistent with previous reports in other pinniped species, five phyla dominated the JFFS  
252 faecal microbiome: *Firmicutes*, *Fusobacteria*, *Bacteroidetes*, *Proteobacteria* and *Actinobacte-*  
253 *ria11* (Nelson et al., 2013b; Pacheco-Sandoval et al., 2019; Stoffel et al., 2020; Bik et al., 2016;  
254 Numberger et al., 2016; Kim et al., 2020). Overall, pinniped gut microbiomes are very variable  
255 between and within species, possibly due to differences in their geographic range (e.g. polar  
256 versus subtropical), diet (benthic vs pelagic hunters, generalist versus specialist), or mating sys-  
257 tems. One or more of *Fusobacteria*, *Firmicutes* and *Bacteroides* (all three in the case of JFFS  
258 and harbour seals), have been found to consistently dominate the overall microbial composi-  
259 tion of pinnipeds, followed by *Proteobacteria* and *Actinobacteria* (Pacheco-Sandoval et al., 2019;  
260 Nelson et al., 2013b). The latter two are usually at lower abundance and *Actinobacteria*, in  
261 particular, has not been described in every pinniped species studied. Another interesting obser-  
262 vation, common to all the studies reviewed, including ours, is that when *Firmicutes* dominates,  
263 the abundance of *Fusobacteria* and *Bacteroidetes* decreases, suggesting some degree of compe-  
264 tition. The *Firmicutes* : *Bacteroidetes* ratio has been well documented in human and mice.  
265 In these land mammals, the ratio increases in response to diets high in lipids and decreases in  
266 response to large amounts of protein (Pu et al., 2016; Hildebrandt et al., 2009; Turnbaugh et al.,  
267 2006). We also observed changes in the relative abundance of *Fusobacteria* were similar to those  
268 observed in *Bacteroidetes*. This suggest some functionally redundant roles.

269 The phylum *Firmicutes* is common in mammalian gut microbiomes (Ley et al., 2008d,a).  
270 Members of this taxonomic group are well known for their role in obesity in humans and mice,  
271 which is associated with an increase in *Firmicutes* and a decrease in *Bacteroidetes* (Pu et al.,  
272 2016; Hildebrandt et al., 2009; Turnbaugh et al., 2006). The energy harvesting role of *Firmicutes*  
273 has also been identified in the zebrafish gut microbiome, where these bacteria are associated with  
274 an increase in lipid droplet numbers in epithelial cells (Semova et al., 2012). Fat is fundamental  
275 for marine mammal survival, as it is needed for energy storage and thermoregulation (Guerrero  
276 and Rogers, 2019) and may explain why *Firmicutes* is consistently among the most dominant  
277 phyla across all pinniped species.

278 The phylum *Fusobacteria* consists of facultative or strict anaerobes that produce various  
279 organic acids from amino acids or carbohydrates fermentation (Olsen, 2014). This phylum is

280 usually found at high relative abundance in the gut microbiomes of strict carnivores adapted  
281 to diets rich in proteins, purines and polyunsaturated fatty acids (Zhu et al., 2018; Guo et al.,  
282 2020). Similar to other marine carnivores, *Fusobacteria* was one of the most abundant phyla in  
283 JFFS (Pacheco-Sandoval et al., 2019). Most of the knowledge generated around the specific role  
284 *Fusobacteria* may play in mammalian intestinal tracts is based on human-centred research. Even  
285 though some genus members seem to play a beneficial role in the human gut microbiome, the  
286 presence of relatively high levels of the genus *Fusobacterium* is more often associated with health  
287 issues (Huh and Roh, 2020; Garrett and Onderdonk, 2014; Potrykus et al., 2008). Conversely, the  
288 high relative abundance of this bacterial genus in the gut of carnivores suggests a rather symbiotic  
289 relationship where *Fusobacterium* is likely to play a role in protein metabolism (Potrykus et al.,  
290 2008).

291 Similar to *Fusobacteria*, the phylum *Bacteroidetes*, especially members of the genus *Bac-*  
292 *teroides* are associated with diets high in animal proteins (Zhu et al., 2018; Guo et al., 2020).  
293 This genus, known for its capacity to degrade animal-derived glycans (Eilam et al., 2014), was  
294 the most abundant *Bacteroidetes*. Similar to previous reports, JFFS samples high in *Firmi-*  
295 *cutes* contained lower relative abundances of *Bacteroidetes* and *Fusobacteria*. This phenomenon  
296 suggests differences in nutritional needs and will be discussed later in the text.

## 297 **4.1 Within sample diversity**

298 Initially, we hypothesised that the alpha diversity of samples collected from BP, a key access  
299 point to Robinson Crusoe island, was going to be different from other colonies. BP is the most  
300 transited area in this study; it connects the airfield with the town and is a popular leisure location  
301 for the local community (Fig. 1). We found instead that BP did not differ from other less-visited  
302 locations such as EA and SC. Therefore, this finding is different to a previous report showing an  
303 association between exposure to anthropogenic stressors and reduced alpha diversity in harbour  
304 seals (Pacheco-Sandoval et al., 2019). The colony at TB was the only location with higher alpha  
305 diversity, indicating that samples collected from TB had a richer and more evenly distributed  
306 microbial composition than other samples. Bacterial richness has been previously associated  
307 with population density due to the increase in microbial sharing (Li et al., 2016). Alternative  
308 studies have suggested that overcrowding might also negatively affect microbial diversity due to

309 higher levels of stress (Bharwani et al., 2016; Partrick et al., 2018). Population density of JFFS  
310 and its effects on the microbiome has not been studied. However, superficial observations from  
311 the field did not suggest differences in population density between the colonies. It may therefore  
312 be that other stressors were limiting alpha diversity in the other locations. For instance, the  
313 colony on TB was relatively sheltered compared to the other colonies, as it was situated on an  
314 open platform a few meters above sea level; in contrast, the other colonies were on narrow strips  
315 of land with greater exposure to sea storms, rockfalls and landslides. Additionally, the colony on  
316 TB is rarely visited by humans due to the complicated access. However, the effects of location  
317 on alpha diversity were marginal. Nevertheless, the stress hypothesis could be tested in future  
318 studies by measuring markers of stress (e.g. cortisol) in the faeces (Wasser et al., 2000).

## 319 4.2 Variation between samples

320 The Bray-Curtis dissimilarity PCoA revealed two distinct clusters. Seventy-five per cent of the  
321 samples clustered together in what we named cluster 1. The remaining samples were grouped as  
322 cluster 2. This variation between clusters was mostly explained by the differences in the relative  
323 abundance of the genera *Fusobacterium* and *Peptoclostridium*. Samples in cluster 1 had a high  
324 relative abundance of *Fusobacterium* and very low *Peptoclostridium* relative abundance, whilst  
325 samples in cluster 2 showed the opposite pattern: increased *Peptoclostridium* and a significant  
326 drop in *Fusobacterium* relative abundance. To our knowledge, this is the first time the genus  
327 *Peptoclostridium* (phylum *Firmicutes*, class *Clostridia*) has been reported in a pinniped gut  
328 microbiome. The family *Peptostreptococcaceae*, to which *Peptoclostridium* belongs, has been  
329 reported in previous studies, but representing no more than 8% of the total composition, and  
330 more often less than 4% (Pacheco-Sandoval et al., 2019; Nelson et al., 2013b; Delpont et al.,  
331 2016). On average, *Peptoclostridium* represented 29% of the microbial composition observed in  
332 Cluster 2 versus the average 3% observed in Cluster 1.

333 The genus *Peptoclostridium* was initially proposed in 2013 and validated in 2016 (Galperin  
334 et al., 2016). This poorly characterised taxonomic group is believed to metabolize amino  
335 acids and oligopeptides and has been isolated from both waste water-mud and marine sedi-  
336 ments (Galperin et al., 2016). The SILVA 132 taxonomy reference database used in this study  
337 included 144 members in the *Peptoclostridium* clade from which only 11 were classified within

338 the four known species of this genus (*P. litorale*, *P. acidaminophilum*, *P. paradoxum* and *P.*  
339 *thermoalcaliphilum*). The remaining clade members were classified as uncultured bacteria. It  
340 should be noted that depending on the taxonomic reference database used, the taxonomic clas-  
341 sification regarding members of the genus *Peptoclostridium* may differ between studies. For  
342 instance, some studies may refer to species such as *Clostridoides difficile* (previously known as  
343 *Clostridium*) as *Peptoclostridium difficile* (Pereira et al., 2016). All four species included in the  
344 SILVA 132 database have been isolated from environments with little or no oxygen (Galperin  
345 et al., 2016). Despite these species being linked to environmental samples, *Peptoclostridium* was  
346 found in at least 90% of the samples. The particular condition required for this bacterial species  
347 to thrive makes it unlikely that the *Peptoclostridium* members found in JFFS faeces originated  
348 from sample contamination by surrounding environmental bacteria. Such high prevalence may  
349 be a sign of a deeper relationship between this uncharacterised bacteria and the host.

350 The microbiome is constantly reshaping through an individuals lifetime. Most of the changes  
351 occur within symbiotic margins responding to factors such as diet, reproductive state and age,  
352 but some changes may also result in dysbiosis and disease (Ley et al., 2008c; Nicholson et al.,  
353 2012). Despite the limited information available on free-range pinnipeds, a few hypotheses may  
354 be suggested to explain the significant changes observed between the two clusters reported in  
355 our study.

356 There is evidence that the mammalian gut microbiota changes over time. This difference is  
357 particularly evident between suckling and post-weaning stages, possibly due to dietary changes  
358 (milk vs solids). As discussed earlier, *Firmicutes* are known for their capacity to regulate  
359 lipid absorption (Semova et al., 2012). Juan Fernandez fur seal milk composition contains  
360 a higher proportion of lipids in comparison to many pinnipeds (~41%) (Ochoa-Acuña et al.,  
361 1999). Thus, if the faecal samples from Cluster 2 were collected from pre-weaning pups (7-  
362 10 months old), it may be expected to that a higher relative abundance of members of the  
363 phylum *Firmicutes* would be found. Similar to the microbial pattern observed in Cluster 2,  
364 samples analysed from Australian fur seal were dominated by the class *Clostridia* in six and  
365 nine months old pups (Smith et al., 2013). In the same study, the families *Lachnospiraceae* and  
366 *Ruminococcaceae* were the most dominant family within this Class, while the overall relative  
367 abundance of *Peptostreptococcaceae*, was less than 4%. Despite age (pre-weaning diet) being a

368 reasonable explanation for the difference observed in our dataset, this hypothesis arrives with  
369 a critical bias. Samples were collected between February and March, and at this point, pups  
370 would be no older than four months. At this stage, pup faeces are still distinguishable from  
371 older individuals in colour and consistency. Individuals from the previous reproductive season  
372 would be older than a year and milk would no longer form a part of their diet. This suggests  
373 that pre-weaning diet is not the explanation for the abundance of *Peptoclostridium*.

374 Differences between genders may also be an explanation of the difference in samples. Otariids  
375 and Phocids such as northern and southern elephant seals exhibit an important degree of sexual  
376 size dimorphism (Ralls and Mesnick, 2009). Gender differences in foraging behaviour and prey  
377 selection have also been reported (Ochoa Acuna and Francis, 1995; Lewis et al., 2006; Andersen  
378 et al., 2013). Based on the differences in diets, it is not surprising to find studies in gut micro-  
379 bial composition also showing gender-based differences. Samples collected from adult Southern  
380 elephant seals evidenced significant differences between adult males and females (Kim et al.,  
381 2020; Nelson et al., 2013b). The same studies did not find differences in leopard or Weddel  
382 seals, less sexually dimorphic phocids. Adult southern elephant seal females showed a signif-  
383 icantly higher relative abundance of *Firmicutes* and less *Fusobacteria* and *Bacteroidetes* than  
384 males (Kim et al., 2020; Nelson et al., 2013b). The proportional changes are very similar to the  
385 one observed between clusters 1 and 2 here. Cluster 2 shows patterns similar to those observed  
386 in females. It seems that the microbial community diverges early in life based on gender as  
387 reported in northern elephant seal pups under naturally controlled diet (Stoffel et al., 2020).  
388 Sexual dimorphism is a common mating strategy in otariids. Thus, it is possible that otariids  
389 such as JFFS, show similar differences as the ones observed in elephant seals. This hypothesis  
390 could be confirmed by using molecular methods for gender identification.

391 A commonality between the gender and age hypotheses is their relationship to the diet.  
392 Differences in diet have been identified as one of the main drivers of gut microbiome diver-  
393 sity (Ley et al., 2008a; Nishida and Ochman, 2018; Nelson et al., 2013c). While pups rely on  
394 lipid-rich milk, fish from the family *Myctophidae* are the most important prey of adult female  
395 JFFS (Francis et al., 1998). Myctophids are known to be rich in fatty acids (Baby et al., 2014;  
396 Lea et al., 2002). Pacheco-Sandoval et al. (2019) showed that harbour seal faecal samples con-  
397 taining more lipid-rich preys had a much higher abundance of *Firmicutes* and lower *Fusobacteria*

398 and *Bacteroidetes*. Molecular identification of prey species in faecal samples, may therefore help  
399 to determine whether diet is the driving factor behind the microbial differences observed here.

400 This study characterised the faecal microbiome of the Juan Fernandez fur seal for the first  
401 time, including colonies from two of the three islands of the Juan Fernandez archipelago to  
402 which the species is endemic. Our findings showed that the overall microbiome composition was  
403 similar to compositions described for other pinnipeds. However, some of the samples showed  
404 a very different microbial composition pattern. This difference was mostly explained by an  
405 inverse relationship between *Peptoclostridium* and *Fusobacterium* abundance. Gender, and its  
406 relationship to foraging behaviour, seems to be the most likely explanation of this phenomenon.  
407 However, additional studies investigating the relationship between gender, age and prey are  
408 required to test this hypothesis. Overall, the results of this study provide a good baseline from  
409 which future hypothesis-based studies can be carried out and it contributes to the understanding  
410 of host-microbial interaction in free-ranging, wild populations of pinnipeds. We highlight the  
411 need to expand knowledge in this field, particularly on microbial functionality, to understand  
412 its different members roles and compare microbial patterns between and within species.

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## 615 Data availability

616 Raw reads data are publicly available in the European Nucleotide Archive (ENA) under the  
617 study accession PRJEB36555. All the scripts used in this study can be accessed in [https:](https://github.com/Cotissima/JFFS_microbiome_first_characterisation)  
618 [//github.com/Cotissima/JFFS\\_microbiome\\_first\\_characterisation](https://github.com/Cotissima/JFFS_microbiome_first_characterisation).

619 **Author contributions**

620 **Conceptualisation:** Constanza Toro-Valdivieso, Barbara Blacklaws.

621 **Formal analysis:** Constanza Toro-Valdivieso, Barbara Blacklaws, Sam Stubbs.

622 **Fieldwork:** Constanza Toro-Valdivieso.

623 **Laboratory work:** Constanza Toro, Frederick Toro.

624 **Funding acquisition:** Constanza Toro-Valdivieso, Barbara Blacklaws, Eduardo Castro-Nallar.

625 **Supervision:** Barbara Blacklaws.

626 **Writing – original draft:** Constanza Toro-Valdivieso.

627 **Writing – review & editing:** Barbara Blacklaws, Eduardo Castro-Nallar, Samuel Stubbs.



Figure 1: Juan Fernandez fur seal (*Arctocephoca philippii philippii*)

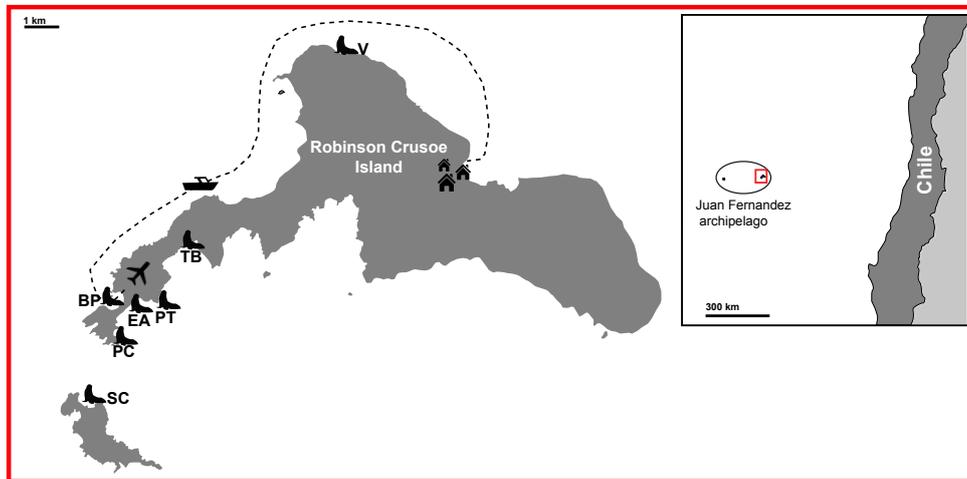
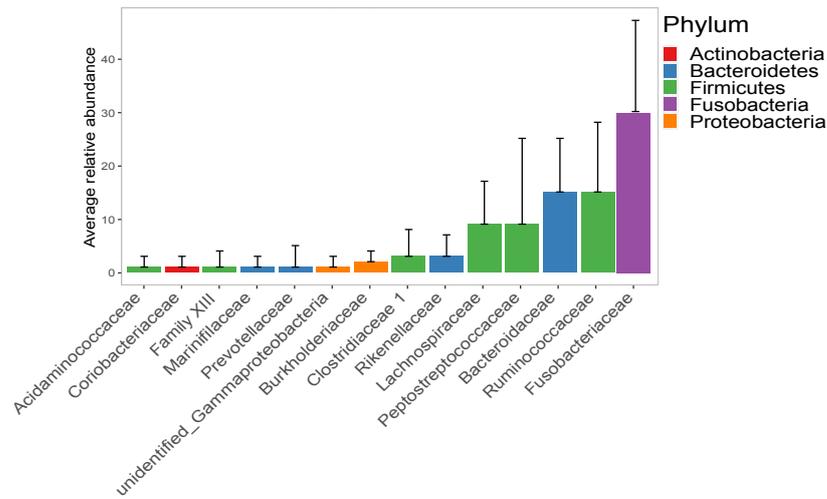


Figure 2: **Simplified map of Robinson Crusoe and Santa Clara islands.** The plane indicates the airfield and the dotted line the access route from the airfield to San Juan Bautista Village (the only settlement on the island). Fur seal icons show the sampling locations. El Arenal (EA) (n = 9), Bahia El Padre (BP) (n = 23), Piedra Carvajal (PC) (n = 1), Punta Trueno (PT) (n = 1), Santa Clara (SC) (n = 12), Tierras Blancas (TB) (n = 10) and Vaqueria (V) (n = 1). 57 samples in total.

### 3.A



### 3.B

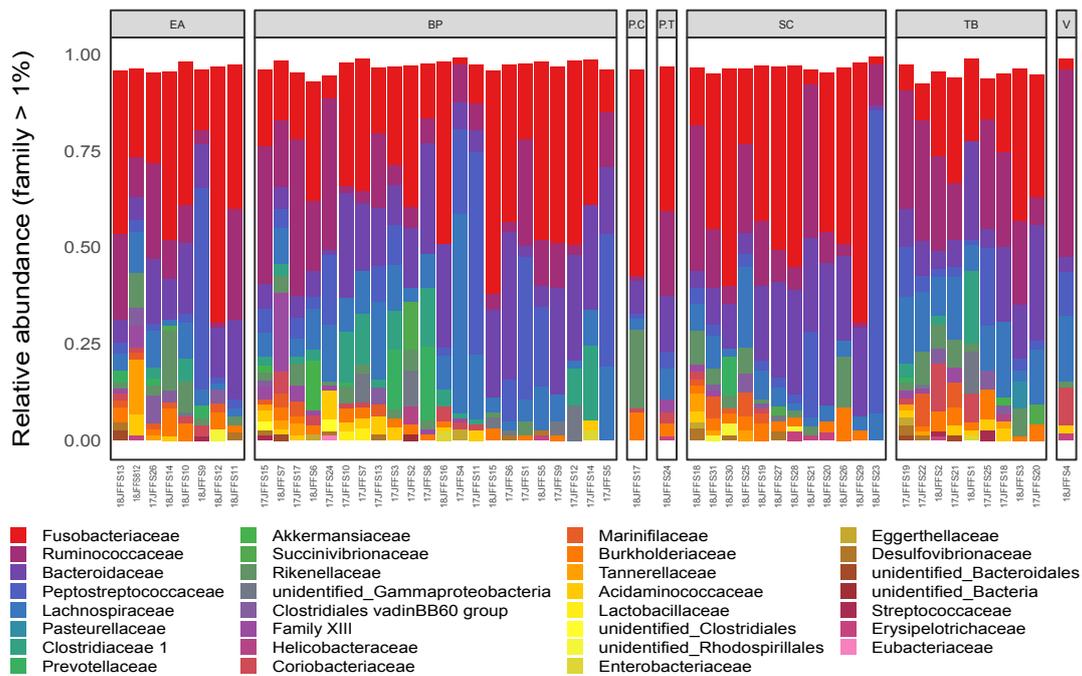


Figure 3: **Figure 3. Compositon of the Juan Fernandez fur seal faecal microbiome at the family level.** Only families with 1% relative abundance are shown. A) Average relative abundance across all samples with standard deviations. B) Relative abundance per sample grouped by location: EA= El Arenal, BP= Bahia El Padre, PC = Piedra Carvajal, PT= Punta Truenos, SC= Santa Clara, TB= Tierras Blancas, V= Vaqueria.

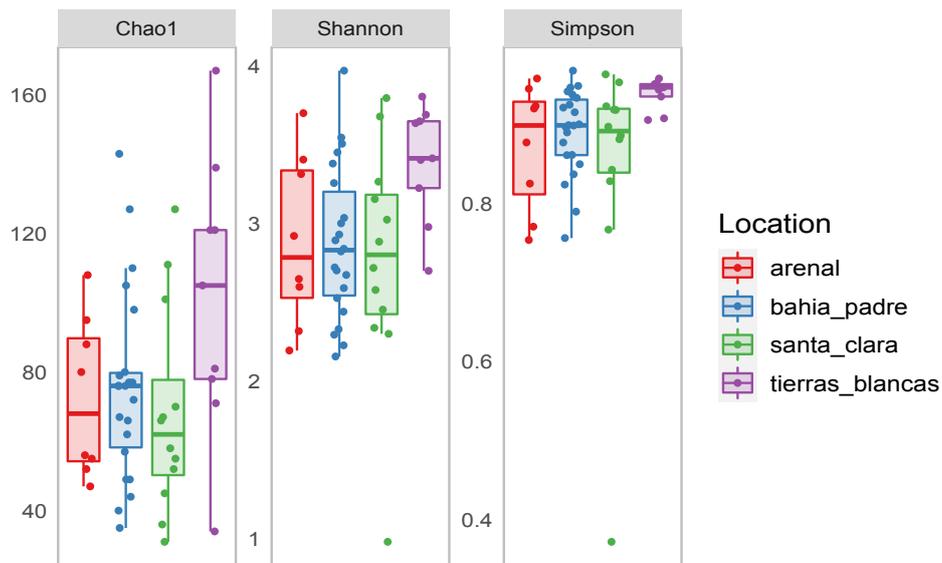


Figure 4: Comparison of three different alpha diversity indices between the four reproductive colonies in the Juan Fernandez archipelago. Samples collected from Tierras Blancas show a tendency to have higher levels of alpha diversity. Filtered rarefied data was used to calculate the diversity estimates.

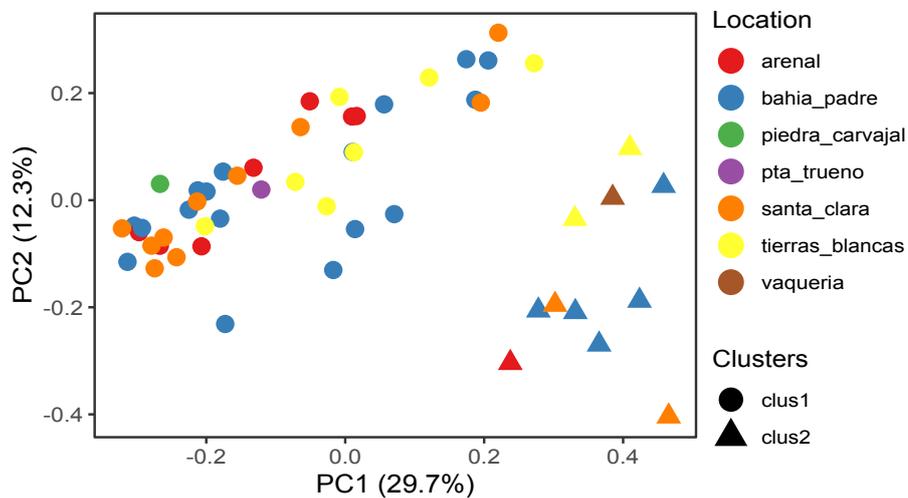


Figure 5: PCoA using Bray-Curtis dissimilarity distance matrix using the filtered rarefied core dataset. Samples clustered in two groups. (circles = cluster 1, triangles = cluster 2). Location is not driving the clustering.

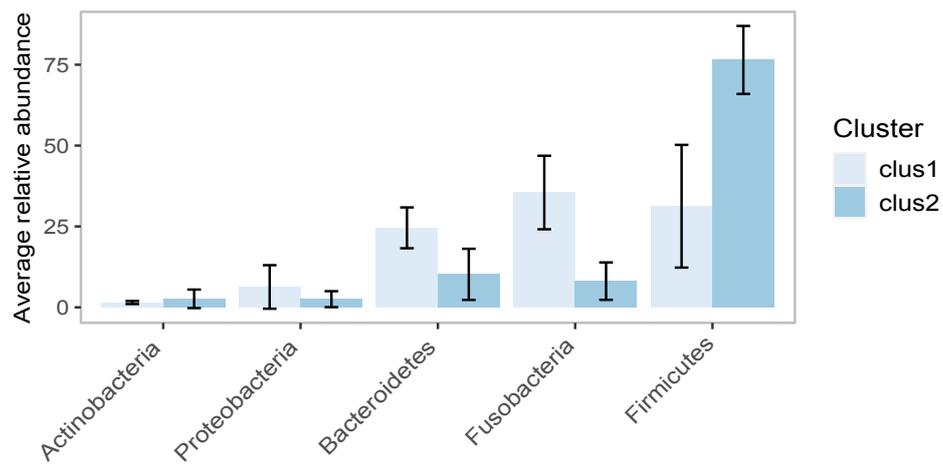


Figure 6: **Relative average abundance of the dominant phyla according to the clusters identified with Bray Curtis dissimilarity.** Showing only phyla with an average relative abundance  $\geq 1\%$ . The differences in microbial patterns can be identified from high taxonomic levels.

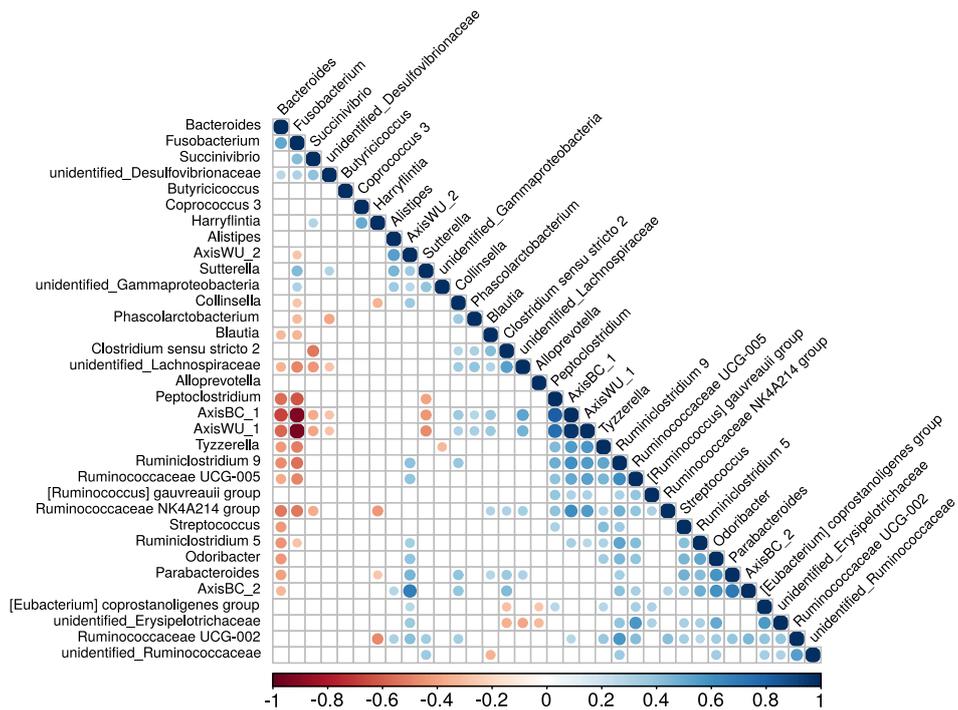


Figure 7: Spearman rank correlation correlogram between bacterial genera and the first two principal components generated from Unifrac and Bray-Curtis distances. The plot shows the direction (blue = positive, red = negative) and the strength (larger = stronger) of the correlation between each pair combination. Only significant correlations ( $p \leq 0.05$ ) are represented with circles.

**Table 1.** Amplicon sequence variants present in at least 90% of the samples. Only three were present in all the samples. Unrarefied data were used to build this table. Abundances was calculated based on the total ASVs count

ASV	Phylum	Family	Genus	Abundance (%)
<b>Present in all the samples</b>				
57729b2b058d8d5253d3e56e4f6386ca	Fusobacteria	Fusobacteriaceae	Fusobacterium	14.93
e8b1922518029c50c69add839142db03	Fusobacteria	Fusobacteriaceae	Fusobacterium	6.52
c0dc53aad260a1b951b7f99966251c7c	Fusobacteria	Fusobacteriaceae	Fusobacterium	3.73
<b>Present in at least 90% of the samples</b>				
f347c63fc5e4aeb97531e656e3765e2a	Firmicutes	Peptostreptococcaceae	Peptoclostridium	8.29
57f9edc6542ce6b78ff352942d6774c6	Bacteroidetes	Bacteroidaceae	Bacteroides	4.28
31984a302fdfe46b5e852fa473e682a4	Bacteroidetes	Bacteroidaceae	Bacteroides	4.26
1153942c5cc40d6ba5609222ded586fe	Firmicutes	Lachnospiraceae	Coprococcus 3	2.98
65dd9f625700a97a1cce9f5eefe4e6cb	Firmicutes	Lachnospiraceae	Blautia	2.18
435975b6d032d4b05233d8b94193b2ad	Firmicutes	Lachnospiraceae	[Ruminococcus] gauvreauii group	1.93
03f74c0ea1f0654719b21d2701e9fa30	Proteobacteria	Burkholderiaceae	Sutterella	1.30
8e10797dedc288dbc0be61fe4b5a5dfb	Actinobacteria	Coriobacteriaceae	Collinsella	1.16

**Table 2.** SIMPER analysis comparing the faecal microbiota composition of Juan Fernandez fur seal at the genus level. The table is showing up to a cumulative contribution of 70%. Cluster averages were calculated based on total counts. Kruskal-Wallis results are only shown when reaching a significant difference

Genus	av_cluster 1	av_cluster 2	Av.Diss	Contrib %	cum%	w	P-value
Peptoclostridium	3%	29%	17%	25	25	3	<0.001
Fusobacterium	34%	8%	17%	24	49	456	<0.001
Bacteroides	14%	6%	7%	10	59	365.5	0.006
[Ruminococcus] gauvreauii group	1%	6%	4%	5	70	124	0.06
Ruminococcaceae UCG-005	4%	7%	4%	6	65		