Metagenomic shotgun analyses reveal complex patterns of intra- and interspecific variation in the intestinal microbiomes of codfishes

5
6 Even Sannes Riiser^{1*}, Thomas H.A. Haverkamp^{1,2}, Srinidhi Varadharajan¹,
7 Ørnulf Borgan³, Kjetill S. Jakobsen¹, Sissel Jentoft¹ and Bastiaan Star^{1*}

¹Centre for Ecological and Evolutionary Synthesis, Department of Biosciences,
University of Oslo, PO Box 1066, Blindern, N-0316 Oslo, Norway.

²Present Address: Department of Epidemiology, Norwegian Veterinary Institute,
Oslo, Norway

14

11

4

8

³Department of Mathematics, University of Oslo, PO Box 1053, Blindern, N 0316 Oslo, Norway.

17 18

19 *** Correspondence**

20 Even Sannes Riiser*: <u>e.s.riiser@ibv.uio.no</u>

21 Bastiaan Star*: <u>bastiaan.star@ibv.uio.no</u>

- 22
- 23

24 Abstract

25 The relative importance of host-specific selection or environmental factors in 26 determining the composition of the intestinal microbiome in wild vertebrates 27 remains poorly understood. Here, we use metagenomic shotgun sequencing of 28 individual specimens to compare the intra- and interspecific variation of 29 intestinal microbiome communities in two ecotypes (NEAC and NCC) of 30 Atlantic cod (Gadus morhua) -that have distinct behavior and habitats- and 31 three *Gadidae* species that occupy a range of ecological niches. Interestingly, we 32 find significantly diverged microbiomes amongst the two Atlantic cod ecotypes. 33 Interspecific patterns of variation are more variable, with significantly diverged 34 communities for most species' comparisons, apart from the comparison between 35 coastal cod (NCC) and Norway pout (Trisopterus esmarkii), whose community

36 compositions are not significantly diverged. The absence of consistent species-37 specific microbiomes suggests that external environmental factors, such as 38 temperature, diet or a combination there-off comprise major drivers of the 39 intestinal community composition of codfishes.

40

41 **Importance**

42 The composition of the intestinal microbial community associated with teleost 43 fish is influenced by a diversity of factors, ranging from internal factors (such as 44 host-specific selection) to external factors (such as niche occupation). These 45 factors are often difficult to separate, as differences in niche occupation (e.g. 46 diet, temperature or salinity) may correlate with distinct evolutionary 47 trajectories. Here, we investigate four gadoid species with contrasting levels of 48 evolutionary separation and niche occupation. Using metagenomic shotgun 49 sequencing, we observe distinct microbiomes amongst two Atlantic cod (Gadus 50 morhua) ecotypes (NEAC and NCC) with distinct behavior and habitats. In 51 contrast, interspecific patterns of variation are more variable. For instance, we 52 do not observe interspecific differentiation between the microbiomes of coastal 53 cod (NCC) and Norway pout (Trisopterus esmarkii) whose lineages have 54 evolutionary separated over 20 million years ago. The observed pattern of 55 microbiome variation in these gadoid species is therefore most parsimoniously 56 explained by differences in niche occupation.

57

58 **1. Introduction**

59 Significant research effort has focused on the importance of external, 60 environmental factors (e.g. habitat, geography, microbial biodiversity, diet, 61 water temperature or salinity) and internal, host-related factors (e.g. genetics, 62 physiology or immunity) in driving the composition of the intestinal 63 microbiome in fish (1, 2). That external factors play an important role is well 64 established. For instance, bacterial diversity in the surrounding water influences 65 the intestinal microbiome in fish larvae and fry (3, 4), water temperature is the main driver for the gut microbiome composition in farmed Tasmanian Atlantic 66 67 salmon (Salmo salar) (5) and diet influences the intestinal composition in both 68 experimental (6–9) as well as wild fish populations (10–13). Yet also internal 69 factors influence the composition of these bacterial communities. For instance, 70 observations of a shared (core) microbiome between wild and laboratory-raised 71 zebrafish suggest that distinct selective pressures determine the composition of 72 the microbial communities (14). Moreover, an association between host 73 phylogeny and intestinal microbiome composition has been observed for a range 74 of fishes, marine animals and terrestrial mammals (15–19).

75

76 The adaptive immune system appears especially important for host selection; 77 individual variation of the Major Histocompatibility Complex (MHC) II 78 correlates with the gut microbiome composition in stickleback (20); mucosal 79 IgT depletion causes dysbiosis in rainbow trout (Oncorhynchus mykiss) (21), 80 and lack of a functional adaptive immune system reduces the strength of host 81 selection in knockout zebrafish models (22). Amongst bony fish, gadoid fishes 82 have an unusual adaptive immune system -through the loss of MHC II, CD4 and invariant chain (Ii) and a range of innate (TLR) and MHC I immune-gene 83 84 expansions (23, 24). Moreover, Atlantic cod has high levels of IgM (25) and a 85 minimal antibody response after pathogen exposure (25–27). Gadoids therefore 86 provide an interesting ecological system to study host-microbiome interactions 87 (28).

88

Studies that specifically integrate internal and external influence support a role for both factors driving the microbial community composition (13, 29). Such studies however, remain restricted in both level of taxonomy of fishes (e.g. (30)) as well as taxonomical resolution of the microbial analyses (16S rRNA: 13, 29, 93 31–33). Importantly, it remains often difficult to separate the correlated effects 94 of distinct behavior (i.e. diet) and niche occupation with interspecific selection 95 and no comparative studies exist that use metagenomic shotgun sequencing to 96 investigate fish populations with profound differences in behavior within a 97 single species. It remains therefore unclear, whether the microbial composition 98 for a range of wild fish species is characterized by intra- or interspecific 99 divergence.

100

101 Here, we study intra- and interspecific divergence of intestinal microbial 102 communities within the wide-spread family of Gadidae using a metagenomic 103 shotgun dataset. We compare the microbiomes from Norway pout (Trisopterus 104 esmarkii), poor cod (Trisopterus minutus), northern silvery pout (Gadiculus 105 thori) and two ecotypes of Atlantic cod (Gadus morhua). These four species have overlapping geographical distributions, are dietary generalists, usually 106 107 feeding over sandy and muddy bottoms on pelagic or benthic crustaceans, 108 polychaetas and (small) fish (34, 35) and have evolutionary diverged 109 approximately 20 million years ago (24). Norway pout is benthopelagic, 110 distributed from the English Channel, around Iceland, and up to the Southwest 111 Barents Sea, mostly found between 100 - 200 meters depth. Poor cod is also 112 benthopelagic, distributed from the Trondheim Fjord in Norway to the 113 Mediterranean Ocean, mostly found between 15 - 200 meters. Northern silvery 114 pout (Gadiculus thori) is meso- to bathypelagic (36), distributed in the North 115 Atlantic Ocean, along the coast of Norway and around Iceland and Greenland. It 116 forms large schools, and are usually found between 200 - 400 meters (34, 36, 117 37). Finally, Atlantic cod has a trans-Atlantic distribution, from the Bay of 118 Biscay to the Barents Sea, the Baltic Sea, around Iceland and Greenland, in the 119 Hudson Bay and along the North American coast (34, 38–42). Atlantic cod 120 comprises various subpopulations and "ecotypes" with distinct adaptations, 121 migratory and feeding behavior. For instance, northeast Arctic cod (NEAC)

performs typical spawning migrations from the Barents Sea to the Norwegian 122 123 coast whereas the Norwegian coastal cod (NCC) remains more stationary (34, 124 43). These ecotypes have increased genomic divergence in several large 125 chromosomal inversions (43–47), suggestive of local adaptation. The 126 environments that these two ecotypes encounter are different, and they feed on 127 distinct types of food. NEAC consumes mostly capelin and herring and NCC 128 feeds on a wide range of crustaceans, fish and seaweed (34, 39, 48). During 129 spawning, these ecotypes spatially co-occur, and long-term gene flow between 130 ecotypes is supported by low overall estimates of divergence in most genomic 131 regions, apart from the chromosomal rearrangements (43).

132

133 We hypothesize that if *interspecific selection* (indicative of host-selection) is the 134 main driver for the intestinal communities in the Gadidae, most differences will 135 be found *between the different species*, and not between the different ecotypes 136 within Atlantic cod. In contrast, if environmental factors are the main drivers for 137 the intestinal communities, we expect significant compositional differences 138 between the ecotypes of Atlantic cod as well as varying levels of differentiation between the species. We use taxonomic profiling of metagenomic shotgun reads 139 140 to classify these microbiomes -obtained from various locations around the 141 Norwegian coast- at order and species-level resolution and analyze within-142 species differentiation of the most abundant members by genome-wide single 143 variation. Finally, differences in gut bacterial community nucleotide 144 composition among the species and ecotypes are assessed using multivariate 145 statistics.

146

147 **2. Methods**

148 **2.1 Sample collection**

149 Northeast Atlantic cod (Gadus morhua) (NEAC, 10 individuals) were collected 150 in Lofoten (N68.0619167, E13.5921667) in March 2014, and Norwegian coastal 151 cod (Gadus morhua) (NCC, 10 individuals) at the same location in August 2014 152 (Fig. 1a, Table S1). NCC (2 individuals) were also collected in the Oslo Fjord 153 (N58.9125100, E9.9202624 & N59.8150006, E10.5544914). Norway pout 154 (Trisopterus esmarkii, 4 individuals), poor cod (Trisopterus minutus, 5 155 individuals) and northern silvery pout (Gadiculus thori, 3 individuals) were collected in the inner Oslo Fjord in May 2015 (Table S1). All fish specimens 156 157 were collected from wild populations. A 3 cm long part of the hindgut 158 (immediately above the short, wider rectal chamber) was aseptically removed 159 post-mortem by scalpel and stored on 70% ethanol. The samples were frozen (-160 20°C) for long-term storage. Relevant metadata such as length, weight, sex and 161 maturity were registered. We always strive to reduce the impact of our sampling 162 needs on populations and individuals. Therefore, samples were obtained as a 163 byproduct of conventional business practice. Specimens were caught by 164 commercial vessels, euthanized by local fishermen and were intended for human 165 consumption. Samples were taken post-mortem and no scientific experiments have been performed on live animals. This sampling follows the guidelines set 166 167 by the "Norwegian consensus platform for replacement, reduction and 168 refinement of animal experiments" (49) and does not fall under any specific 169 legislation in Norway, requiring no formal ethics approval.

170 2.2 Sample preparation and DNA extraction

171 Intestinal samples were split open lengthwise, before the combined gut content 172 and mucosa was gently removed using a sterile disposable spatula. Each 173 individual sample was washed in 500 µl 100% EtOH and centrifuged before the 174 ethanol was allowed to evaporate, after which dry weight was measured before 175 proceeding to DNA extraction. DNA was extracted from between < 10 and 300 176 mg dry weight of gut content using the MoBio Powersoil HTP 96 Soil DNA 177 Isolation Kit (Qiagen, Valencia, CA, USA) according to the DNA extraction 178 protocol (v. 4.13) utilized by the Earth Microbiome Project (50). DNA was 179 eluted in 100 µl Elution buffer, and stored at -20° Celsius. Due to high 180 methodological consistency between biological replicates in previous 181 experiments, only one sample was collected per fish (32).

182 **2.3** Sequence data generation and filtering

183 Quality and quantity of the DNA was measured using a Qubit fluorometer (Life 184 Technologies, Carlsbad, CA, USA), and normalized by dilution. DNA libraries 185 were prepared using the Kapa HyperPlus kit (Roche Sequencing, Pleasanton, 186 CA, USA) and paired-end sequenced (2x125 base pairs) on an Illumina 187 HiSeq2500 using the HiSeq SBS V4 chemistry with dual-indexing in two 188 independent sequencing runs. Read qualities were assessed using FastQC (51), 189 before adapter removal, singleton read identification, de-duplication and further 190 read quality trimming was performed using *Trimmomatic* (ver. 0.36) (52) and 191 *PRINSEQ-lite* (ver. 0.20.4) (53) (Table S2). PhiX-, host- and human sequences 192 were removed by mapping reads to the phiX reference genome 193 [GenBank: J02482.1], the Atlantic cod genome assembly (gadMor 2) (this 194 applied to all the fish species) (54), and a masked version of the human genome 195 (HG19) (55) using BWA (ver. 0.7.13) (56) or BBMap (ver. 37.53) (JGI) with 196 default parameters, and discarding matching sequences using seqtk (ver. 197 2012.11) (58). All sequence data have been deposited in the European 198 Nucleotide Archive (ENA) under study accession number PRJEB31095.

199 2.4 Taxonomic profiling

200 Taxonomic classification of quality trimmed and filtered metagenomic paired-201 end reads was performed using Kaiju (ver. 1.5.0) (59) ("greedy" heuristic 202 approach, -e 5), with the NCBI nr database (rel. 84) (incl. proteins from fungal 203 and microbial eukaryotes) as reference (60). Counts of sequences successfully 204 assigned to orders and species were imported into *RStudio* (ver. 1.1.383) (61) 205 based on R (ver. 3.4.2) (62) for further processing. Filtering of the most 206 abundant bacterial orders for visualization was based on a minimum relative 207 abundance threshold of 1% of the total number of sequences per library 208 (threshold ranging from 5933 - 95,146 depending on sample size). Similarly, 209 filtering of the most abundant bacterial species was based on a minimum 210 relative abundance threshold of 2% of the total number of sequences per library 211 (threshold ranging from 6548 - 190,294 depending on sample size). Any taxon 212 not exceeding this threshold in at least one (order level)/two (species level) 213 sample(s) was removed. All filtering was based on the *R* package *genefilter* (ver. 214 1.62.0) (63). Final results were visualized using the R package ggplot (ver. 215 2.2.1) (64). Note: Based on a recent reclassification (65), we refer to the Photobacterium phosphoreum 216 reference strain ANT-2200 (acc. nr. 217 GCF_000613045.2) as *Photobacterium kishitanii* (Table S3).

218 **2.5 Sequence variation analysis**

In order to assess the heterogeneity of the most abundant bacteria in the fish species, we analyzed the sequence variation in the two genomes with the highest mean relative abundance over all fish species and ecotypes; *Photobacterium kishitanii* and *Photobacterium iliopiscarium*. Paired-end reads from each individual fish were mapped to the reference genomes (Table S3) using the *Snakemake* workflow (66) of *anvi'o* (ver. 5.1) (67) with default parameters in the "all-against-all" modus (with *anvi-profile --min-coverage-for-variability 20*).

226 Samples of low coverage, restricting detection of SNVs in anvi'o, were 227 excluded from the variation analysis. For each individual sample, variable sites 228 were identified, and the mean number of these per 1000 bp calculated (variation 229 density). A variable site required a minimum coverage of 20X. Next, variable 230 sites with a minimum of 20X coverage in *all* samples were defined as single 231 nucleotide variants (SNVs, anvi-gen-variability-profile --min-occurrence 1 --232 *min-coverage-in-each-sample 20*). Coverage, variation density and SNV profiles 233 were plotted in RStudio following the R script provided by anvi'o (68). The 234 anvi'o SNV output was converted to .vcf format using a custom-developed 235 script (https://github.com/srinidhi202/AnvioSNV_to_vcf), and the resulting .vcf 236 files were used in a principal component analysis (PCA) to test for population 237 differences as implemented in *smartpca* (ver. 6.1.4) (*EIGENSOFT*) (69).

238 **2.6 Statistical analysis**

Although included in data visualization, the Oslo Fjord NCC samples were 239 240 excluded from statistical analysis, due to low sample size (n = 2). Within-sample 241 diversity (alpha diversity) was calculated using the *diversity* function in the R package vegan (ver. 2.4-1) (70) based on Shannon, Simpson and Inverse 242 243 Simpson indices calculated from non-normalized order-level read counts (Table 244 S4). Differences in alpha diversity were studied using linear regression. The 245 "optimal model" (the model that best describes the individual diversity) was 246 identified through a "top-down" strategy including all covariates (Table S5), 247 except weight, which highly correlated with length (r = 0.95), and selected 248 through *t*-tests. Model assumptions were verified through plotting of residuals. 249 Differences in bacterial community structure (beta diversity) between the fish 250 species or ecotypes were visualized using non-metric multidimensional scaling 251 (NMDS) plots based on the Bray-Curtis dissimilarity calculated from order-level 252 sequence counts. Next, pairwise differences in beta diversity between the fish 253 species or ecotypes were tested using Permutational Multivariate Analysis of 254 Variance (PERMANOVA) in the *R* package *pairwise.adonis* (ver. 0.1) (71), a 255 wrapper for the *adonis* functions in *vegan* (ver. 2.4-1), based on Bray-Curtis 256 dissimilarity calculated from order-, genus- and species-level sequence counts. 257 *pairwise.adonis* was run with 20,000 permutations, and p values adjusted for 258 multiple testing using the Holm method (72). Adjusted *p*-values < 0.05 indicate 259 statistical significance. PERMANOVA assumes the multivariate dispersion in 260 the compared groups to be homogeneous; this was verified (p > 0.05) using the 261 *betadisper* function (*vegan*) (Table S6). Similarity percentage (SIMPER) 262 procedure implemented in *vegan* was used to quantify the contribution of 263 individual orders to the overall Bray-Curtis dissimilarity between the 264 species/ecotypes. All beta diversity analyses were based on sequence counts 265 normalized using a common scaling procedure, following McMurdie & Holmes 266 2014 (73). This involves multiplying the sequence count of every unit (e.g. 267 order) in a given library with a factor corresponding to the ratio of the smallest 268 library size in the dataset to the library size of the sample in question. 269 Normalization using this procedure effectively results in the library scaling by 270 averaging an infinite number of repeated sub-samplings. We used Chi-squared 271 statistics, as implemented in *smartpca* (69), to test for significant differences in 272 the distribution of SNVs per reference genome, while correcting for multiple 273 testing using sequential Bonferroni (72).

274

275 **3. Results**

276

3.1 Taxonomical composition of the intestinal microbiomes

We analyze a dataset of 422 million paired-end reads, with a median sample size of 11.9 million reads (8.0 - 19.6 million reads per sample) (Table 2, Table S7). 280 Following filtering, order level classification could be obtained for 93% of all 281 sequences (Table 2). Based on non-normalized order-level sequence counts, we 282 observe clear patterns of separation between species and ecotypes in a 283 multivariate NMDS plot (Fig. 1b), with NEAC and northern silvery pout 284 forming distinct clusters, whereas the NCC populations encompasses the 285 Norway pout and poor cod populations. Vibrionales is the most abundant order 286 in the intestinal microbiomes of NCC specimens at both coastal locations (mean 287 relative abundance (MRA): 76%) as well as Norway pout (MRA: 79%) and 288 poor cod (MRA: 44%) (Fig. 2a, Table 3), with the remainder of each gut 289 community consisting of a mix of orders with low relative abundance. The 290 intestinal microbiome of the NEAC and northern silvery pout specimens have a 291 significantly more diverse community composition (Fig. 2a, Fig. 3). NEAC is 292 dominated by *Bacteroidales* (MRA: 21%), *Vibrionales* (MRA: 17%), 293 Clostridiales (MRA: 12%) and Brevinematales (MRA: 7%) and northern silvery 294 pout has a high relative abundance of orders Brachyspirales (MRA: 16%) and 295 Clostridiales (MRA: 14%). Distinct from the gut community of the other fish 296 populations, northern silvery pout has a low abundance of *Vibrionales*. Finally, 297 the amount of sequences in the "Others" category, as well as sequences 298 classified above order level (mean all samples: 7.8%), vary slightly between the 299 fish species (Table S8). A species-level classification was obtained for 66% of 300 all sequences. Overall, species of the genus Photobacterium comprise on 301 average 40.6% of the classified sequences, ranging from 0.2% in northern 302 silvery pout to 74.3% in Norway pout (Fig. 2b). In particular, P. kishitanii and 303 P. iliopiscarium represent on average 43% and 36% of all Photobacterium 304 species, although the ratio differs in the different fish species (e.g. 49% vs. 41%) 305 in NCC, 16% vs. 56% in NEAC and 55% vs. 12% in Norway pout).

306

The NCC Lofoten intestinal microbiome is dominated by *P. iliopiscarium* (MRA: 21%) and *P. kishitanii* (MRA: 20%), followed by different species of

309 Aliivibrio (wodanis, logei and fischeri) (MRA: 13%) (Fig. 2b). Similarly, the 310 bacterial gut community of Norway pout is also dominated by *Photobacterium* 311 species, in particular *P. kishitanii* (MRA: 17%). The intestinal microbiome of 312 poor cod is dominated by *Photobacterium* species (MRA: 18%), followed by 313 different Vibrio spp. (MRA: 8%). The gut bacterial community of NEAC is 314 more diverse, with high relative abundance of a Brevinema sp. (MRA: 31%) and 315 different species in the genera Photobacterium (MRA: 34%), Clostridium 316 (MRA: 12%) and Aliivibrio (MRA: 9%). The high abundance of Bacteroidales 317 observed at the order level (Fig. 2a) is not reflected at the species level, as this 318 order represents a high number of *Bacteroidales* species with low abundance. 319 Consequently, no *Bacteroidales* species are among the 15 most abundant species 320 in the NEAC intestinal microbiome (Fig. 2b). The NEAC samples also contain a 321 Mucispirillum sp. (MRA: 4%) and two Brachyspira spp. (MRA: 2%). In 322 northern silvery pout, the gut microbiome is quite evenly distributed between 323 the Brevinema sp., the Mucispirillum sp., Brachyspira pilosicoli, Brachyspira 324 sp. CAG:700 and a group of different *Clostridium* species in two of three 325 samples. The third sample contains the same species, but has an even higher 326 relative abundance of the Brevinema sp. (64%) (Fig. 2b).

327 3.2 Variation in bacterial community composition among species and 328 ecotypes

Significant differences in within-sample diversity (alpha diversity) at the order level are observed among all species and within-species ecotypes, except between NCC and Norway pout (Table 4, Table S5). None of the other covariates have a significant effect on alpha diversity. Similar to the results from the within-sample diversity, significant differences in community structure (beta diversity) are observed among the gadoid species at order-, genus- and species level (Table 5, Table S6). At the order level, the NEAC intestinal community 336 has a different structure than what is observed in all the other gadoids (0.05 337 significance level). The NCC intestinal microbiome is also different from that of 338 both poor cod and northern silvery pout. In agreement with results of within-339 sample (alpha) diversity, no differences in community structure are observed 340 between the microbiomes of NCC and Norway pout. Finally, no differences are 341 observed between the gut microbiome of poor cod vs. Norway pout, poor cod 342 vs. northern silvery pout or Norway pout vs. northern silvery pout (p = 0.074 for 343 all). Beta diversity analysis also demonstrate that community differences at the 344 genus and species level are similar to those observed at the order level (Table 345 S6).

346

347 Differences in the intestinal community composition between these gadoids are 348 predominantly explained by changes in the relative abundance of a limited 349 number of orders. For example, different proportions of Vibrionales contribute 350 29% to the (Bray-Curtis) dissimilarity between the NCC and NEAC (p = 0.001), 351 followed by differences in the relative abundance of *Bacteroidales*, explaining 352 10% of the dissimilarity (p = 0.001) (Table S9). Together, 80% of the observed 353 dissimilarity between NCC and NEAC is explained by differences in their 354 relative abundance of the top six orders. Similarly, 60% of the dissimilarity 355 between NCC and northern silvery pout are driven by Vibrionales, 356 Brachyspirales and Clostridiales.

357 3.3 Bacterial within-species variation of Single Nucleotide Variant heterogeneity

We investigated bacterial within-species variation of *P. iliopiscarium* and *P. kishitanii* – with sufficient read coverage across all samples– among the different gadoids by mapping sequencing reads to their respective reference genomes (GCF_000949935.1, GCF_000613045.2). In the samples used for SNV analysis,

363 the mean percentage of the reference genomes with minimum 20-fold coverage 364 (coverage breadth) after mapping were 63% for *P. iliopiscarium* and 19% for *P.* 365 kishitanii. Hence, the variation analysis of the two species is based on different 366 proportions of the reference genomes. The two reference genomes greatly vary 367 in the number of SNVs observed in all samples, from 84,866 in *P. iliopiscarium* 368 to 1229 in P. kishitanii, Fig. 4a). The density of variable sites within each 369 individual sample shows varying levels of heterogeneity in the bacterial 370 populations (Fig. 4b). This heterogeneity is particularly clear in *P. kishitanii*, 371 with sites-density varying from 0.5 to 45.4 variant positions per Kbp per 372 individual specimen. Further, the heatmap shows gadoid specific SNV patterns 373 (Fig. 4c), in particular for *P. iliopiscarium*, where Norway pout contains a 374 distinct pattern compared to the other gadoids, indicating the presence of 375 specific P. iliopiscarium strain(s). Statistical analyses of SNV variation reveals 376 that NEAC has a significantly different SNV pattern from Norway pout (Chi-377 square, p = 0.017) and poor cod (p = 0.028) for P. kishitanii, and from NCC (p = 0.028) 378 0.033) and Norway pout (p = 0.000) for *P. iliopiscarium* (Fig. 4d, Table S10). 379 NCC has a significantly different SNV pattern from Norway pout (p = 0.003) 380 for P. iliopiscarium. (Fig. 4d, Table S10). The relative abundance of P. 381 kishitanii and P. iliopiscarium vary greatly among the fish specimens used in the 382 variation analysis (Fig. 4e).

383

384 **4. Discussion**

385

Using metagenomic shotgun sequencing, we show the composition of the intestinal microbiomes of two Atlantic cod ecotypes (NEAC, NCC) to be at least as divergent as those found between the different codfish species investigated here. Our findings have several implications for our understanding of the composition of the intestinal microbiome in wild fish populations.

391

392 Although species-specific selection has been proposed as a factor driving the 393 composition of the intestinal community in fish in a variety of settings (13, 14, 394 16–19, 29, 33), our results show that this may not be the most important driver 395 among gadoid species in wild populations. First, we observe highly significant 396 differences in the intestinal microbiomes at order-, species- and within-species 397 bacterial level between the NEAC and NCC ecotypes. Despite showing different 398 migratory behavior, these ecotypes co-occur during seasonal spawning in 399 northern Norway (Lofoten), from where most of the samples are collected (43– 400 45). Second, we observe no significant bacterial order- or species-level 401 differences in the intestinal microbiome between different gadoid species, 402 Atlantic cod (ecotype NCC) and Norway pout, which are sampled from different 403 geographical locations (Lofoten and Oslo fjord). We visually do not observe any 404 differentiation between the NCC sampled from Lofoten and the Oslo fjord 405 (although statistical certainly is low), which reflects an earlier observed lack of 406 geographical structure for this ecotype (30). The similar microbial composition 407 of the NCC and Norway pout is striking, as these are distinctly different genetic lineages with an evolutionary separation of at least 20 million years (24). These 408 409 results suggest that NCC and Norway pout occupy an environmental niche that 410 allows bacterial members with a broad geographical distribution to colonize 411 their intestinal communities. Overall, the observation of a significant 412 differentiation between microbiomes from ecotypes of the same species and a 413 lack of differentiation between microbiomes from two distinct species, suggest 414 that the intestinal microbiome in these gadoid species and ecotypes is not driven 415 by species-specific selection alone.

416

There are several factors that may underlie the compositional differences in the
NCC and NEAC intestinal microbiomes. First, for more than 10 months during
the year, the two populations encounter different habitats, as the NEAC ecotype

420 is distributed in the pelagic waters of the Barents Sea, while NCC remains more 421 stationary in coastal waters (74). Although several 16S rRNA-based studies 422 have reported limited effects of geographic location on the composition and 423 diversity of the fish intestinal microbiome (32, 75), the Barents Sea has 424 significantly lower temperatures (76) compared to Norwegian coastal waters 425 (77). Temperature has been shown to have a significant impact on the intestinal 426 microbiome in several studies (e.g., Senegalese sole (Solea senegalesis), 427 Tasmanian Atlantic Salmon (Salmo salar) and mummichog (Fundulus 428 heteroclitus) (5, 78, 79) but not in all cases (e.g., Atlantic salmon (80). Second, 429 the ecotypes were sampled during different seasons; NCC Lofoten during 430 summer (August) and NEAC during winter/early spring (March). Nonetheless, a 431 lack of difference between NCC Lofoten (August) and NCC Oslo fjord (May) 432 suggests that seasonality is unlikely to fully explain the observed differences 433 between NEAC and NCC. Third, the ecotypes show different feeding behavior; 434 while the NEAC during foraging and spawning migrations from the Barents Sea 435 may perform vertical movements down to 500 meters (42, 81, 82), NCC mainly 436 occupy shallow and warmer coastal and fjord waters (83). This behavior exposes the two ecotypes to different sources of food, with NEAC predominantly eating 437 438 capelin and herring (48), and NCC living of a more diverse diet, including 439 crustaceans, fish and even seaweeds (34, 39). Diet has been shown to influence 440 the composition of the intestinal microbiome in several fish species (9, 10, 13, 441 78, 84, 85). Finally, Barents Sea has a high microbial biodiversity compared to 442 coastal areas (86). The specific bacterial load in the surrounding waters also 443 influences the intestinal microbiome composition in fish, including Atlantic cod 444 (3, 4). Nonetheless, because these different environmental and behavioral factors 445 are correlated, it is unclear which of these parameters contributes the most to the 446 observed differences in the intestinal microbiome composition between these 447 ecotypes.

448

449 Comparing two spatially separated coastal Atlantic cod populations, 450 metagenomic shotgun data revealed no strain-level differentiation (30). In this 451 study, we find specific SNV variants amongst the most abundant bacterial 452 species that are associated with either species or Atlantic cod ecotype. This 453 indicates that NEAC harbor different strains of P. iliopiscarium than those 454 identified in the NCC ecotype and the other gadoid species. Our current study 455 encompasses a significantly greater geographical area and taxonomical samples 456 than the earlier coastal comparison (30–32), and is indicative of strain-level 457 variation at such larger comparative scales. In line with Riiser et al. 2019 (30), 458 this study shows that such strain-level differences cannot be detected using 16S 459 rRNA techniques alone, and that metagenomic shotgun sequencing is currently 460 the most accurate approach to detect strain-level spatial variation in the marine 461 environment.

462

463 Most striking amongst the comparisons of gadoid species are the microbiome 464 differences observed in NEAC, northern silvery pout and poor cod compared to 465 NCC and Norway pout. Several bacterial species that drive this differentiation 466 are of particular interest. First, two bacterial species, Mucispirillum sp. and 467 Brevinema sp., are almost exclusively detected in the intestinal microbiomes of 468 NEAC and northern silvery pout. Nonetheless, these genera are represented by a 469 single species in the *RefSeq* database (60) (accessed 10.01.19) and hence little is 470 known. B. andersonii (order Brevinematales) was originally identified in short-471 tailed shrews (Blarina brevicauda) and white-footed mice (Peromyscus 472 leucopus), and were found unable to grow below 25°C (87). Brevinema sp. has 473 previously been identified in Atlantic cod (32) and in Atlantic salmon (88). 474 Mucispirillum schaedleri (order Deferribacterales) is a mucosa-associated 475 member of the intestinal microbiome in terrestrial animals as pigs, goats and 476 rodents, where it is thought to be involved in mucus production through 477 expression of lectins, important components in the innate immune response (89,

478 90). Nevertheless, the distant relationship between Atlantic cod and these 479 terrestrial hosts, and the availability of only single reference genomes for 480 Mucispirillum and Brevinema, strongly suggests that the representatives found 481 here represent related, but novel species with a different intestinal ecology and 482 physiology. Second, both NEAC and northern silvery pout contain significant 483 fractions of *Brachyspira* spp., previously identified as dominant members in the 484 gut of the carnivorous marine fish species mahi mahi (*Coryphaena hippurus*) 485 (12, 91). *Brachyspira* spp. are known as intestinal pathogens in pigs and humans 486 (92, 93), although recent studies show that *Brachyspira* spp. are more 487 widespread in the wildlife community than previously thought, including in 488 freshwater (94). The ecology of *Brachyspira* in the marine environment is unclear, although an association with the carnivorous diet of mahi mahi and 489 490 NEAC may suggest that the diet of northern silvery pout also has a considerable 491 carnivorous component. Third, poor cod is the only species with considerable 492 abundance of *Enterovibrio norvegicus* (Table S11). This bacterium within the 493 Vibrionaceae family was isolated from the intestines of cultured turbot 494 (Scophthalmus maximus) larvae in Norway, and classified as a novel species 495 phenotypically similar to the Vibrio genus (95). Interestingly, poor cod are also 496 host to the highest abundance of *Vibrio* spp. among the fish species in this study 497 (Table S11). Other *Enterovibrio* species have been found in association with 498 diseased corals (96) and internal organs of cultured fish species in the 499 Mediterranean Ocean (97–99). However, little is known about the function of 500 this relatively novel genus in fish intestines.

501

502 Given the observations of species specific selection for a similar microbiome in 503 various teleosts and range of habitats (13, 14, 16–19, 29, 33), the diverse 504 microbiomes *within* and *among* gadoid species may suggest that their intestinal 505 communities could be more easily modulated by external factors. At this stage, 506 limited sampling across various fish taxa and the lack of comparative 507 approaches leave reasons for such diverse communities speculative. 508 Nonetheless, it is interesting to note that all gadoids have unusual adaptive 509 immune system -through the loss of MHC II, CD4 and invariant chain (Ii) and a 510 range of innate (TLR) and MHC I immune-gene expansions (23, 24). There are 511 significant correlations between immune genes and the vertebrate microbiome 512 (100, 101), and it has been hypothesized that adaptive immunity has evolved to 513 help maintain complex community of beneficial commensal bacteria (102). 514 Indeed, studies of wild-type zebrafish and knockout zebrafish without a 515 functional adaptive immune system suggested that adaptive immunity increases 516 the strength of host filtering of potential fish-associated microbes (22). The 517 unusual adaptive immune system of gadoids may therefore affect the strength of 518 co-evolutionary associations with their microbiome.

519

520 **5.** Conclusion

521 Based on metagenomic shotgun sequencing, we here characterize the intra- and 522 interspecific community composition among two ecotypes of Atlantic cod and 523 three related fish species in the Gadidae family. Several of these fish species 524 harbor unique, and possibly novel bacterial species. We identify a complex 525 pattern of diversity with significant differences between the Atlantic cod ecotypes, and variable interspecific patterns of variation. Although most species 526 527 and ecotypes yield different communities, those found in coastal cod (NCC) and 528 Norway pout are not significantly diverged, indicating that ecological niche 529 plays an important role in determining the intestinal microbiomes in these 530 gadoid species.

531

532 **Conflict of interest**

533 The authors declare that the research was conducted in the absence of any 534 commercial or financial relationships that could be construed as a potential 535 conflict of interest.

536

537 Authors' contributions

538 SJ, BS and THA conceived and designed the experiments. KSJ provided the 539 initial framework for the study. ESR and SJ sampled the specimens. ESR 540 performed the laboratory work. ESR and THA performed data analysis. SV 541 created the Python script to convert the anvi'o format to VCF. ØB, THA, ESR 542 and BS interpreted the results. ESR and BS wrote the paper with input of all 543 authors. All authors read and approved the final manuscript.

544

545 Funding

This work was funded by a grant from the Research Council of Norway (project
no. 222378) and University of Oslo (Strategic Research Initiative) – both to
KSJ.

549

550 Acknowledgements

We thank Børge Iversen and Helle Tessand Baalsrud for their kind help in sampling Atlantic cod specimens in Lofoten, and Martin Malmstrøm, Paul Ragnar Berg and Monica Hongrø Solbakken for sampling at Sørøya. We are grateful for the metagenome sequencing performed at the Norwegian Sequencing Centre (NSC: https://www.sequencing.uio.no).

556

557 Availability of data and materials

558 The data set generated and analyzed for this study is available in the European

559 Nucleotide Archive (ENA), study accession number PRJEB31095.

- 560
- 561 **References**

5	67
J	02

562		
563	1.	Wang AR, Ran C, Ringø E, Zhou ZG. 2017. Progress in fish
564		gastrointestinal microbiota research. Rev Aquac 0:1–15.
565	2.	Romero J, Ringø E, Merrifield DL. 2014. The Gut Microbiota of Fish.
566		Aquac Nutr 75–100.
567	3.	Nayak SK. 2010. Role of gastrointestinal microbiota in fish. Aquac Res
568		41:1553–1573.
569	4.	Olafsen JA. 1983. Ingestion of bacteria by cod (Gadus morhua L.) larvae,
570		p. 627–643. In The Propagation of Cod Gadus morhua L.: an international
571		symposium.
572	5.	Neuman C, Hatje E, Zarkasi KZ, Smullen R, Bowman JP, Katouli M.
573		2016. The effect of diet and environmental temperature on the faecal
574		microbiota of farmed Tasmanian Atlantic Salmon (Salmo salar L.). Aquac
575		Res 47:660–672.
576	6.	Zarkasi KZ, Taylor RS, Abell GCJ, Tamplin ML, Glencross BD, Bowman
577		JP. 2016. Atlantic Salmon (Salmo salar L.) Gastrointestinal Microbial
578		Community Dynamics in Relation to Digesta Properties and Diet. Microb
579		Ecol 71:589–603.
580	7.	Zhou Z, Olsen RE. 2012. Culturable autochthonous gut bacteria in Atlantic
581		salmon (Salmo salar L.) fed diets with or without chitin. Characterization
582		by 16S rRNA gene sequencing, ability to produce enzymes and in vitro
583	_	growth inhibition of four fish pathogens.
584	8.	Schmidt V, Amaral-Zettler L, Davidson J, Summerfelt S, Good C. 2016.
585		Influence of fishmeal-free diets on microbial communities in Atlantic
586		salmon (<i>Salmo salar</i>) recirculation aquaculture systems. Appl Environ
587	0	Microbiol 82:4470–4481.
588	9.	Desai AR, Links MG, Collins SA, Mansfield GS, Drew MD, Van Kessel
589		AG, Hill JE. 2012. Effects of plant-based diets on the distal gut
590		microbiome of rainbow trout (<i>Oncorhynchus mykiss</i>). Aquaculture 350–
591	10	353:134–142. Balaish DL Saesshara LK, Uisash DE, Lashar CL, Ora E, Darla D, Lasia
592	10.	Bolnick DI, Snowberg LK, Hirsch PE, Lauber CL, Org E, Parks B, Lusis
593		AJ, Knight R, Caporaso JG, Svanba R. 2014. Individual diet has sex-
594 505	11	dependent effects on vertebrate gut microbiota. Nat Commun 5.
595 506	11.	Bolnick DI, Snowberg LK, Hirsch PE, Lauber CL, Knight R, Caporaso
596		JG, Svanbäck R. 2014. Individuals' diet diversity influences gut microbial
597 508		diversity in two freshwater fish (threespine stickleback and Eurasian perch). Ecol L att 17:070, 87
598 500	10	perch). Ecol Lett 17:979–87. Givens CE, Ransom B, Bano N, Hollibaugh JT. 2015. Comparison of the
599 600	12.	gut microbiomes of 12 bony fish and 3 shark species. Mar Ecol Prog Ser
600 601		518:209–223.
601 602	13.	Miyake S, Ngugi DK, Stingl U. 2015. Diet strongly influences the gut
602 603	13.	microbiota of surgeonfishes. Mol Ecol 24:656–672.
603 604	14.	Roeselers G, Mittge EK, Stephens WZ, Parichy DM, Cavanaugh CM,
004	17.	Robberts G, mille Lix, Stephens WZ, I arony Divi, Cavanaugh Civi,

605		Guillemin K, Rawls JF. 2011. Evidence for a core gut microbiota in the
606		zebrafish. ISME J 5:1595–608.
607	15.	Ley RE, Ley RE, Hamady M, Lozupone C, Turnbaugh PJ, Ramey RR,
608		Bircher JS, Schlegel ML, Tucker TA, Schrenzel MD, Knight R, Gordon JI.
609		2008. Evolution of Mammals and Their Gut Microbes. Science 1647.
610	16.	Zhang M, Sun Y, Liu Y, Qiao F, Chen L, Liu WT, Du Z, Li E. 2016.
611	10.	Response of gut microbiota to salinity change in two euryhaline aquatic
612		animals with reverse salinity preference. Aquaculture 454:72–80.
613	17.	Li X, Yu Y, Feng W, Yan Q, Gong Y. 2012. Host Species as a Strong
614	1,1	Determinant of the Intestinal Microbiota of Fish Larvae. J Microbiol
615		50:29–37.
616	18.	Li J, Ni J, Li J, Wang C, Li X, Wu S, Zhang T, Yu Y, Yan Q. 2014.
617	101	Comparative study on gastrointestinal microbiota of eight fish species with
618		different feeding habits. J Appl Microbiol 117:1750–1760.
619	19.	Navarrete P, Magne F, Araneda C, Fuentes P, Barros L, Opazo R, Espejo
620		R, Romero J. 2012. PCR-TTGE analysis of 16S rRNA from Rainbow trout
621		(Oncorhynchus mykiss) gut microbiota reveals host-specific communities
622		of active bacteria. PLoS One 7:1–10.
623	20.	Bolnick DI, Snowberg LK, Caporaso JG, Lauber C, Knight R, Stutz WE.
624		2014. Major Histocompatibility Complex class IIb polymorphism
625		influences gut microbiota composition and diversity. Mol Ecol 23:4831–
626		45.
020		+ J.
620 627	21.	
	21.	Salinas I, Casadei E, Takizawa F, Shibasaki Y, Sunyer OJ. 2018.
627	21.	
627 628	21. 22.	Salinas I, Casadei E, Takizawa F, Shibasaki Y, Sunyer OJ. 2018. Interactions between microbiota and the teleost immune system in health
627 628 629		Salinas I, Casadei E, Takizawa F, Shibasaki Y, Sunyer OJ. 2018. Interactions between microbiota and the teleost immune system in health and disease. J Immunol 200.
627 628 629 630		Salinas I, Casadei E, Takizawa F, Shibasaki Y, Sunyer OJ. 2018. Interactions between microbiota and the teleost immune system in health and disease. J Immunol 200. Stagaman K, Burns AR, Guillemin K, Bohannan BJM. 2017. The role of
627 628 629 630 631		Salinas I, Casadei E, Takizawa F, Shibasaki Y, Sunyer OJ. 2018.Interactions between microbiota and the teleost immune system in health and disease. J Immunol 200.Stagaman K, Burns AR, Guillemin K, Bohannan BJM. 2017. The role of adaptive immunity as an ecological filter on the gut microbiota in
627 628 629 630 631 632	22.	 Salinas I, Casadei E, Takizawa F, Shibasaki Y, Sunyer OJ. 2018. Interactions between microbiota and the teleost immune system in health and disease. J Immunol 200. Stagaman K, Burns AR, Guillemin K, Bohannan BJM. 2017. The role of adaptive immunity as an ecological filter on the gut microbiota in zebrafish. ISME J March:1–10.
627 628 629 630 631 632 633	22.	 Salinas I, Casadei E, Takizawa F, Shibasaki Y, Sunyer OJ. 2018. Interactions between microbiota and the teleost immune system in health and disease. J Immunol 200. Stagaman K, Burns AR, Guillemin K, Bohannan BJM. 2017. The role of adaptive immunity as an ecological filter on the gut microbiota in zebrafish. ISME J March:1–10. Star B, Nederbragt AJ, Jentoft S, Grimholt U, Malmstrøm M, Gregers TF,
627 628 629 630 631 632 633 634	22.	 Salinas I, Casadei E, Takizawa F, Shibasaki Y, Sunyer OJ. 2018. Interactions between microbiota and the teleost immune system in health and disease. J Immunol 200. Stagaman K, Burns AR, Guillemin K, Bohannan BJM. 2017. The role of adaptive immunity as an ecological filter on the gut microbiota in zebrafish. ISME J March:1–10. Star B, Nederbragt AJ, Jentoft S, Grimholt U, Malmstrøm M, Gregers TF, Rounge TB, Paulsen J, Solbakken MH, Sharma A, Wetten OF, Lanzén A,
627 628 629 630 631 632 633 634 635	22.	 Salinas I, Casadei E, Takizawa F, Shibasaki Y, Sunyer OJ. 2018. Interactions between microbiota and the teleost immune system in health and disease. J Immunol 200. Stagaman K, Burns AR, Guillemin K, Bohannan BJM. 2017. The role of adaptive immunity as an ecological filter on the gut microbiota in zebrafish. ISME J March:1–10. Star B, Nederbragt AJ, Jentoft S, Grimholt U, Malmstrøm M, Gregers TF, Rounge TB, Paulsen J, Solbakken MH, Sharma A, Wetten OF, Lanzén A, Winer R, Knight J, Vogel J-H, Aken B, Andersen O, Lagesen K,
627 628 629 630 631 632 633 634 635 636	22.	 Salinas I, Casadei E, Takizawa F, Shibasaki Y, Sunyer OJ. 2018. Interactions between microbiota and the teleost immune system in health and disease. J Immunol 200. Stagaman K, Burns AR, Guillemin K, Bohannan BJM. 2017. The role of adaptive immunity as an ecological filter on the gut microbiota in zebrafish. ISME J March:1–10. Star B, Nederbragt AJ, Jentoft S, Grimholt U, Malmstrøm M, Gregers TF, Rounge TB, Paulsen J, Solbakken MH, Sharma A, Wetten OF, Lanzén A, Winer R, Knight J, Vogel J-H, Aken B, Andersen O, Lagesen K, Tooming-Klunderud A, Edvardsen RB, Tina KG, Espelund M, Nepal C,
627 628 629 630 631 632 633 634 635 636 637	22.	 Salinas I, Casadei E, Takizawa F, Shibasaki Y, Sunyer OJ. 2018. Interactions between microbiota and the teleost immune system in health and disease. J Immunol 200. Stagaman K, Burns AR, Guillemin K, Bohannan BJM. 2017. The role of adaptive immunity as an ecological filter on the gut microbiota in zebrafish. ISME J March:1–10. Star B, Nederbragt AJ, Jentoft S, Grimholt U, Malmstrøm M, Gregers TF, Rounge TB, Paulsen J, Solbakken MH, Sharma A, Wetten OF, Lanzén A, Winer R, Knight J, Vogel J-H, Aken B, Andersen O, Lagesen K, Tooming-Klunderud A, Edvardsen RB, Tina KG, Espelund M, Nepal C, Previti C, Karlsen BO, Moum T, Skage M, Berg PR, Gjøen T, Kuhl H,
627 628 629 630 631 632 633 634 635 636 637 638	22.	 Salinas I, Casadei E, Takizawa F, Shibasaki Y, Sunyer OJ. 2018. Interactions between microbiota and the teleost immune system in health and disease. J Immunol 200. Stagaman K, Burns AR, Guillemin K, Bohannan BJM. 2017. The role of adaptive immunity as an ecological filter on the gut microbiota in zebrafish. ISME J March:1–10. Star B, Nederbragt AJ, Jentoft S, Grimholt U, Malmstrøm M, Gregers TF, Rounge TB, Paulsen J, Solbakken MH, Sharma A, Wetten OF, Lanzén A, Winer R, Knight J, Vogel J-H, Aken B, Andersen O, Lagesen K, Tooming-Klunderud A, Edvardsen RB, Tina KG, Espelund M, Nepal C, Previti C, Karlsen BO, Moum T, Skage M, Berg PR, Gjøen T, Kuhl H, Thorsen J, Malde K, Reinhardt R, Du L, Johansen SD, Searle S, Lien S,
627 628 629 630 631 632 633 634 635 636 637 638 639	22.	 Salinas I, Casadei E, Takizawa F, Shibasaki Y, Sunyer OJ. 2018. Interactions between microbiota and the teleost immune system in health and disease. J Immunol 200. Stagaman K, Burns AR, Guillemin K, Bohannan BJM. 2017. The role of adaptive immunity as an ecological filter on the gut microbiota in zebrafish. ISME J March:1–10. Star B, Nederbragt AJ, Jentoft S, Grimholt U, Malmstrøm M, Gregers TF, Rounge TB, Paulsen J, Solbakken MH, Sharma A, Wetten OF, Lanzén A, Winer R, Knight J, Vogel J-H, Aken B, Andersen O, Lagesen K, Tooming-Klunderud A, Edvardsen RB, Tina KG, Espelund M, Nepal C, Previti C, Karlsen BO, Moum T, Skage M, Berg PR, Gjøen T, Kuhl H, Thorsen J, Malde K, Reinhardt R, Du L, Johansen SD, Searle S, Lien S, Nilsen F, Jonassen I, Omholt SW, Stenseth NC, Jakobsen KS. 2011. The
627 628 629 630 631 632 633 634 635 636 637 638 639 640 641 642	22.	 Salinas I, Casadei E, Takizawa F, Shibasaki Y, Sunyer OJ. 2018. Interactions between microbiota and the teleost immune system in health and disease. J Immunol 200. Stagaman K, Burns AR, Guillemin K, Bohannan BJM. 2017. The role of adaptive immunity as an ecological filter on the gut microbiota in zebrafish. ISME J March:1–10. Star B, Nederbragt AJ, Jentoft S, Grimholt U, Malmstrøm M, Gregers TF, Rounge TB, Paulsen J, Solbakken MH, Sharma A, Wetten OF, Lanzén A, Winer R, Knight J, Vogel J-H, Aken B, Andersen O, Lagesen K, Tooming-Klunderud A, Edvardsen RB, Tina KG, Espelund M, Nepal C, Previti C, Karlsen BO, Moum T, Skage M, Berg PR, Gjøen T, Kuhl H, Thorsen J, Malde K, Reinhardt R, Du L, Johansen SD, Searle S, Lien S, Nilsen F, Jonassen I, Omholt SW, Stenseth NC, Jakobsen KS. 2011. The genome sequence of Atlantic cod reveals a unique immune system. Nature 477:207–10. Malmstrøm M, Matschiner M, Tørresen OK, Star B, Snipen LG, Hansen
627 628 629 630 631 632 633 634 635 636 637 638 639 640 641 642 643	22. 23.	 Salinas I, Casadei E, Takizawa F, Shibasaki Y, Sunyer OJ. 2018. Interactions between microbiota and the teleost immune system in health and disease. J Immunol 200. Stagaman K, Burns AR, Guillemin K, Bohannan BJM. 2017. The role of adaptive immunity as an ecological filter on the gut microbiota in zebrafish. ISME J March:1–10. Star B, Nederbragt AJ, Jentoft S, Grimholt U, Malmstrøm M, Gregers TF, Rounge TB, Paulsen J, Solbakken MH, Sharma A, Wetten OF, Lanzén A, Winer R, Knight J, Vogel J-H, Aken B, Andersen O, Lagesen K, Tooming-Klunderud A, Edvardsen RB, Tina KG, Espelund M, Nepal C, Previti C, Karlsen BO, Moum T, Skage M, Berg PR, Gjøen T, Kuhl H, Thorsen J, Malde K, Reinhardt R, Du L, Johansen SD, Searle S, Lien S, Nilsen F, Jonassen I, Omholt SW, Stenseth NC, Jakobsen KS. 2011. The genome sequence of Atlantic cod reveals a unique immune system. Nature 477:207–10. Malmstrøm M, Matschiner M, Tørresen OK, Star B, Snipen LG, Hansen TF, Baalsrud HT, Nederbragt AJ, Hanel R, Salzburger W, Stenseth NC,
627 628 629 630 631 632 633 634 635 636 637 638 639 640 641 642 643 644	22. 23.	 Salinas I, Casadei E, Takizawa F, Shibasaki Y, Sunyer OJ. 2018. Interactions between microbiota and the teleost immune system in health and disease. J Immunol 200. Stagaman K, Burns AR, Guillemin K, Bohannan BJM. 2017. The role of adaptive immunity as an ecological filter on the gut microbiota in zebrafish. ISME J March:1–10. Star B, Nederbragt AJ, Jentoft S, Grimholt U, Malmstrøm M, Gregers TF, Rounge TB, Paulsen J, Solbakken MH, Sharma A, Wetten OF, Lanzén A, Winer R, Knight J, Vogel J-H, Aken B, Andersen O, Lagesen K, Tooming-Klunderud A, Edvardsen RB, Tina KG, Espelund M, Nepal C, Previti C, Karlsen BO, Moum T, Skage M, Berg PR, Gjøen T, Kuhl H, Thorsen J, Malde K, Reinhardt R, Du L, Johansen SD, Searle S, Lien S, Nilsen F, Jonassen I, Omholt SW, Stenseth NC, Jakobsen KS. 2011. The genome sequence of Atlantic cod reveals a unique immune system. Nature 477:207–10. Malmstrøm M, Matschiner M, Tørresen OK, Star B, Snipen LG, Hansen TF, Baalsrud HT, Nederbragt AJ, Hanel R, Salzburger W, Stenseth NC, Jakobsen KS, Jentoft S. 2016. Evolution of the immune system influences
627 628 629 630 631 632 633 634 635 636 637 638 639 640 641 642 643 644 645	22.23.24.	 Salinas I, Casadei E, Takizawa F, Shibasaki Y, Sunyer OJ. 2018. Interactions between microbiota and the teleost immune system in health and disease. J Immunol 200. Stagaman K, Burns AR, Guillemin K, Bohannan BJM. 2017. The role of adaptive immunity as an ecological filter on the gut microbiota in zebrafish. ISME J March: 1–10. Star B, Nederbragt AJ, Jentoft S, Grimholt U, Malmstrøm M, Gregers TF, Rounge TB, Paulsen J, Solbakken MH, Sharma A, Wetten OF, Lanzén A, Winer R, Knight J, Vogel J-H, Aken B, Andersen O, Lagesen K, Tooming-Klunderud A, Edvardsen RB, Tina KG, Espelund M, Nepal C, Previti C, Karlsen BO, Moum T, Skage M, Berg PR, Gjøen T, Kuhl H, Thorsen J, Malde K, Reinhardt R, Du L, Johansen SD, Searle S, Lien S, Nilsen F, Jonassen I, Omholt SW, Stenseth NC, Jakobsen KS. 2011. The genome sequence of Atlantic cod reveals a unique immune system. Nature 477:207–10. Malmstrøm M, Matschiner M, Tørresen OK, Star B, Snipen LG, Hansen TF, Baalsrud HT, Nederbragt AJ, Hanel R, Salzburger W, Stenseth NC, Jakobsen KS, Jentoft S. 2016. Evolution of the immune system influences speciation rates in teleost fishes. Nat Genet 48:1204–1210.
627 628 629 630 631 632 633 634 635 636 637 638 639 640 641 642 643 644	22. 23.	 Salinas I, Casadei E, Takizawa F, Shibasaki Y, Sunyer OJ. 2018. Interactions between microbiota and the teleost immune system in health and disease. J Immunol 200. Stagaman K, Burns AR, Guillemin K, Bohannan BJM. 2017. The role of adaptive immunity as an ecological filter on the gut microbiota in zebrafish. ISME J March:1–10. Star B, Nederbragt AJ, Jentoft S, Grimholt U, Malmstrøm M, Gregers TF, Rounge TB, Paulsen J, Solbakken MH, Sharma A, Wetten OF, Lanzén A, Winer R, Knight J, Vogel J-H, Aken B, Andersen O, Lagesen K, Tooming-Klunderud A, Edvardsen RB, Tina KG, Espelund M, Nepal C, Previti C, Karlsen BO, Moum T, Skage M, Berg PR, Gjøen T, Kuhl H, Thorsen J, Malde K, Reinhardt R, Du L, Johansen SD, Searle S, Lien S, Nilsen F, Jonassen I, Omholt SW, Stenseth NC, Jakobsen KS. 2011. The genome sequence of Atlantic cod reveals a unique immune system. Nature 477:207–10. Malmstrøm M, Matschiner M, Tørresen OK, Star B, Snipen LG, Hansen TF, Baalsrud HT, Nederbragt AJ, Hanel R, Salzburger W, Stenseth NC, Jakobsen KS, Jentoft S. 2016. Evolution of the immune system influences

648 and affinity. Comp Biochem Physiol Part B 154:309-316. 649 26. Pilström L, Warr G, Strömberg S. 2005. Why is the antibody response of 650 Atlantic cod so poor? The search for a genetic explanation. Fish Sci 651 71:961-971. 652 27. Solem ST, Stenvik J. 2006. Antibody repertoire development in teleosts — 653 a review with emphasis on salmonids and Gadus morhua L. Dev Comp 654 Immunol 30:57–76. 655 28. Star B, Jentoft S. 2012. Why does the immune system of Atlantic cod lack 656 MHC II? Bioessays 34:648–51. 657 29. Sullam KE, Essinger SD, Lozupone CA, O'Connor MP, Rosen GL, 658 Knight R, Kilham SS, Russell JA. 2012. Environmental and ecological 659 factors that shape the gut bacterial communities of fish: a meta-analysis. 660 Mol Ecol 21:3363–78. 661 Riiser ES, Haverkamp THA, Varadharajan S, Borgan Ø, Jakobsen KS, 30. 662 Jentoft S. 2019. Switching on the light: using metagenomic shotgun sequencing to characterize the intestinal microbiome of Atlantic cod. 663 664 Environ Microbiol 21:2576–2594. 665 31. Star B, Haverkamp TH, Jentoft S, Jakobsen KS. 2013. Next generation 666 sequencing shows high variation of the intestinal microbial species 667 composition in Atlantic cod caught at a single location. BMC Microbiol 668 13:248. 669 32. Riiser ES, Haverkamp THA, Borgan Ø, Jakobsen KS, Jentoft S, Star B. 670 2018. A single Vibrionales 16S rRNA oligotype dominates the intestinal 671 microbiome in two geographically separated Atlantic cod populations. 672 Front Microbiol 9:1–14. 673 Sevellec M, Laporte M, Bernatchez A, Derome N, Bernatchez L. 2019. 33. 674 Evidence for host effect on the intestinal microbiota of whitefish 675 (Coregonus sp.) species pairs and their hybrids. Ecol Evol 9:11762–11774. 676 Cohen DM, Inada T, Iwamoto T, Scialabba N. 1990. Gadiform fishes of 34. 677 the world (Order Gadiformes). An annotated and illustrated catalogue of 678 cods, hakes, grenadiers and other gadiform fishes known to date.FAO 679 Species Catalogue. 680 35. Froese R, Pauly D. 2019. fishbase.org, version (06/2018). fishbase.org. 681 36. Gaemers PAM, Poulsen JY. 2017. Recognition and Distribution of Two 682 North Atlantic Gadiculus Species, G. argenteus and G. thori (Gadidae), 683 Based on Otolith Morphology, Larval Pigmentation, Molecular Evidence, 684 Morphometrics and Meristics. Fishes 2:1–24. 685 Albert OT. 1993. Distribution, population structure and diet of silvery pout 37. 686 (Gadiculus argenteus thori J. Schmidt), poor cod (Trisopterus minutus 687 minutus (L.)), four-bearded rockling (Rhinonemus cimbrius (L.)), and 688 Vahl's eelpout (Lycodes vahlii gracilis Reinhardt) in the Norwegian Deep. 689 Sarsia 78:141–154. 690 Froese, Rainer and Pauly D. 2012. Species Fact Sheets: Gadus morhua 38.

691		(Linnaeus, 1758) Food and Agriculture Organization of the United
692		Nations.
693	39.	Link JS, Bogstad B, Sparholt H, Lilly GR. 2009. Trophic role of Atlantic
694		cod in the ecosystem. Fish Fish 10:58–87.
695	40.	Link JS, Garrison LP. 2002. Trophic ecology of Atlantic cod (Gadus
696		morhua) on the Northeast US continental shelf. Mar Ecol Prog Ser
697		227:109–123.
698	41.	Michalsen K, Johannesen E, Bogstad B. 2008. Feeding of mature cod
699		(Gadus morhua) on the spawning grounds in Lofoten. ICES J Mar Sci
700		65:571–580.
701	42.	Godø OR, Michalsen K. 2000. Migratory behaviour of North-east Arctic
702		cod, studied by use of data storage tags. Fish Res 48:127–140.
703	43.	Berg PR, Star B, Pampoulie C, Sodeland M, Barth JMI. 2016. Three
704		chromosomal rearrangements promote genomic divergence between
705		migratory and stationary ecotypes of Atlantic cod. Sci Rep 6:1–12.
706	44.	Berg PR, Star B, Pampoulie C, Bradbury IR, Bentzen P, Hutchings JA,
707		Jentoft S, Jakobsen KS. 2017. Trans-oceanic genomic divergence of
708		Atlantic cod ecotypes is associated with large inversions. Heredity (Edinb)
709		119:418–428.
710	45.	Star B, Boessenkool S, Gondek AT, Nikulina EA, Hufthammer AK,
711		Pampoulie C, Knutsen H, André C, Nistelberger HM, Dierking J, Petereit
712		C, Heinrich D, Jakobsen KS, Jentoft S, Stenseth NC, Barrett JH. 2017.
713		Ancient DNA reveals the Arctic origin of Viking Age cod. Proc Natl Acad
714		Sci U S A 114:9152–9157.
715	46.	Kirubakaran TG, Grove H, Kent MP, Sandve SR, Baranski M, Nome T,
716		Rosa MCDE, Righino B, Johansen T. 2016. Two adjacent inversions
717		maintain genomic differentiation between migratory and stationary
718		ecotypes of Atlantic cod. Mol Ecol 25:2130–2143.
719	47.	Sodeland M, Jorde PE, Lien S, Jentoft S, Berg PR, Grove H, Kent MP,
720	. / .	Arnyasi M, Olsen EM, Knutsen H. 2016. "Islands of Divergence" in the
721		Atlantic Cod Genome Represent Polymorphic Chromosomal
722		Rearrangements. Genome Biol Evol 8:1012–1022.
723	48.	Holt RE, Bogstad B, Durant M, Dolgov A V, Ottersen G. 2019. Barents
724	10.	Sea cod (<i>Gadus morhua</i>) diet composition: long-term interannual,
725		seasonal, and ontogenetic patterns. ICES J Mar Sci May 2019.
726	49.	Norecopa. Norecopa guidelines for animal experiments.
727	50.	Gilbert JA, Meyer F, Jansson J, Gordon J, Pace N, Ley R, Fierer N, Field
728	50.	D, Kyrpides N, Glöckner F. 2010. The Earth Microbiome Project: Meeting
729		report of the "1st EMP meeting on sample selection and acquisition" at
730		Argonne National Laboratory October 6th 2010.
731	51.	Andrews S. 2010. FastQC.
732	51. 52.	Bolger AM, Lohse M, Usadel B. 2014. Trimmomatic: A flexible trimmer
733	54.	for Illumina sequence data. Bioinformatics 30:2114–2120.
155		Tor mamma sequence data. Diomiormatics 30.2117–2120.

734 53. Schmieder R, Edwards R. 2011. Quality control and preprocessing of 735 metagenomic datasets. Bioinformatics 27:863-864. 736 54. Tørresen OK, Star B, Jentoft S, Reinar WB, Grove H, Miller JR, Walenz 737 BP, Knight J, Ekholm JM, Peluso P, Edvardsen RB, Tooming-Klunderud 738 A, Skage M, Lien S, Jakobsen KS, Nederbragt AJ. 2017. An improved 739 genome assembly uncovers prolific tandem repeats in Atlantic cod. BMC 740 Genomics 18:1–23. 741 Genome Reference Consortium. 2009. Genome Reference Consortium 55. 742 Human Build 37 (GRCh37). 743 56. Li H, Durbin R. 2009. Fast and accurate short read alignment with 744 Burrows – Wheeler transform. Bioinformatics 25:1754–1760. 745 57. Joint Genome Institute. BBMap. 746 58. Li H. 2012. Seqtk. 747 59. Menzel P, Krogh A. 2016. Fast and sensitive taxonomic classification for metagenomics with Kaiju. Nat Commun 7. 748 749 O'Leary NA, Wright MW, Brister JR, Ciufo S, Haddad D, McVeigh R, 60. 750 Rajput B, Robbertse B, Smith-White B, Ako-Adjei D, Astashyn A, 751 Badretdin A, Bao Y, Blinkova O, Brover V, Chetvernin V, Choi J, Cox E, 752 Ermolaeva O, Farrell CM, Goldfarb T, Gupta T, Haft D, Hatcher E, 753 Hlavina W, Joardar VS, Kodali VK, Li W, Maglott D, Masterson P, 754 McGarvey KM, Murphy MR, O'Neill K, Pujar S, Rangwala SH, Rausch 755 D, Riddick LD, Schoch C, Shkeda A, Storz SS, Sun H, Thibaud-Nissen F, 756 Tolstoy I, Tully RE, Vatsan AR, Wallin C, Webb D, Wu W, Landrum MJ, 757 Kimchi A, Tatusova T, DiCuccio M, Kitts P, Murphy TD, Pruitt KD. 758 2016. Reference sequence (RefSeq) database at NCBI: Current status, 759 taxonomic expansion, and functional annotation. Nucleic Acids Res 760 44:D733-D745. 761 61. Racine JS. 2010. Rstudio: A platform-independent ide for R and 762 SWEAVE. Financ Dev 47:36–37. 763 R Core Team. 2017. R: A language and environment for statistical 62. 764 computing. R Found Stat Comput Vienna, Austria. 765 Gentleman R, Carey V, Huber W, Hahne F. 2019. Package 'genefilter.' 63. 766 Wickham H. 2009. Ggplot2Applied Spatial Data Analysis with R. 64. 767 65. Machado H, Gram L. 2017. Comparative genomics reveals high genomic 768 diversity in the genus *Photobacterium*. Front Microbiol 8:1–14. 769 Köster J, Rahmann S. 2012. Snakemake - a scalable bioinformatics 66. 770 workflow engine. Bioinformatics 28:2520–2522. 771 Eren AM, Esen C, Quince C, Vineis JH, Morrison HG, Sogin ML, 67. 772 Delmont TO. 2015. Anvi'o: An advanced analysis and visualization 773 platform for 'omics data. PeerJ 3:1–29. 774 Eren AM, Esen C, Quince C, Vineis JH, Morrison HG, Sogin ML, 68. 775 Delmont TO. 2015. Visualizing SNV profiles using R. merenlab.org. 776 http://merenlab.org/tutorials/infant-gut.

777	60	Potterson N. Price AL Deich D. 2006 Deputation Structure and
777 778	69.	Patterson N, Price AL, Reich D. 2006. Population Structure and Eigenanalysis. PLoS Genet 2.
779	70.	Oksanen J, Blanchet FG, Friendly M, Kindt R, Legendre P, McGlinn D,
780	70.	Minchin PR, O'Hara RB, Simpson GL, Solymos P, Stevens MHH, Scoecs
781		E, Wagner H. 2017. <i>vegan</i> : Community Ecology Package R package
782		version 2.4–3. 2.4.3.
783	71.	Martinez Arbizu P. 2019. pairwiseAdonis: Pairwise multilevel comparison
784	/1.	using adonis. R package version 0.1.
785	72.	Holm S. 1979. A Simple Sequentially Rejective Multiple Test Procedure.
786	/	Scand J Stat 6:65–70.
787	73.	McMurdie PJ, Holmes S. 2014. Waste Not, Want Not: Why Rarefying
788	,	Microbiome Data Is Inadmissible. PLoS Comput Biol 10.
789	74.	Nordeide JT, Johansen SD, Jørgensen TE, Karlsen BO, Moum T. 2011.
790		Population connectivity among migratory and stationary cod <i>Gadus</i>
791		<i>morhua</i> in the Northeast Atlantic — A review of 80 years of study. Mar
792		Ecol Prog Ser 435:269–283.
793	75.	Llewellyn MS, McGinnity P, Dionne M, Letourneau J, Thonier F,
794		Carvalho GR, Creer S, Derome N. 2016. The biogeography of the Atlantic
795		salmon (Salmo salar) gut microbiome. ISME J 10:1280–1284.
796	76.	Furevik T. 2001. Annual and interannual variability of Atlantic Water
797		temperatures in the Norwegian and Barents Seas: 1980 - 1996. Deep Res
798		48:383–404.
799	77.	Eilertsen HC, Skarðhamar J. 2006. Temperatures of north Norwegian
800		fjords and coastal waters: Variability, significance of local processes and
801		air-sea heat exchange. Estuarine, Coast Shelf Sci 67 67:530–538.
802	78.	Givens CE. 2012. A fish tale: Comparison of the gut microbiome of 15
803		fish species and the influence of diet and temperature on its composition.
804	79.	Martin-Antonio B, Manchado M, Infante C, Zerolo R, Labella A, Alonso
805		C, Borrego JJ. 2007. Intestinal microbiota variation in Senegalese sole
806		(Solea senegalensis) under different feeding regimes. Aquac Res 38:1213-
807		1222.
808	80.	Hovda MB, Fontanillas R, Mcgurk C, Obach A, Rosnes JT. 2012.
809		Seasonal variations in the intestinal microbiota of farmed Atlantic salmon
810		(Salmo salar L.). Aquac Res 43:154–159.
811	81.	Karlsen BO, Klingan K, Emblem Å, Jørgensen TE, Jueterbock A,
812		Furmanek T, Hoarau G, Johansen SD, Nordeide JT, Moum T. 2013.
813		Genomic divergence between the migratory and stationary ecotypes of
814	00	Atlantic cod. Mol Ecol 22:5098–5111.
815	82.	Stensholt BK. 2001. Cod migration patterns in relation to temperature:
816	02	analysis of storage tag data. ICES J Mar Sci 58:770–793.
817	83.	WWF. 2004. The Barents Sea Cod.
818 810	Q /	www.wwf.no/core/pdf/wwf_codreport_2004.pdf.
819	84.	Merrifield DL, Olsen RE, Myklebust R, Ringø E. 2011. Dietary Effect of

820 Soybean (*Glycine max*) Products on Gut Histology and Microbiota of Fish. 821 Soybean Nutr 231–250. 822 85. Ringø E, Zhou Z, Olsen RE, Song SK. 2012. Use of chitin and krill in 823 aquaculture - the effect on gut microbiota and the immune system: A 824 review. Aquac Nutr 18:117-131. 825 Sakshaug E, Loeng H. 1994. Structure, biomass distribution, and 86. 826 energetics of the pelagic ecosystem in the Barents Sea: A synopsis. Polar 827 Biol 14:405-411. 828 Defosse DL, Johnson RC, Paster BJ, Dewhirst FE, Fraser GJ, Haven W. 87. 829 1995. Spirochete Isolated from the Short-Tailed Shrew (Blarina 830 brevicauda) and the White-Footed Mouse (Peromyscus leucopus). Int J 831 Syst Bacteriol 45:78-84. Gajardo K, Rodiles A, Kortner TM, Krogdahl Å, Bakke AM, Merrifield 832 88. 833 DL, Sørum H. 2016. A high-resolution map of the gut microbiota in 834 Atlantic salmon (Salmo salar): A basis for comparative gut microbial 835 research. Sci Rep 6:30893. 836 Loy A, Pfann C, Steinberger M, Hanson B, Herp S, Brugiroux S, Gomes 89. 837 C, Rattei T, Stecher B. 2017. Lifestyle and Horizontal Gene Transfer-838 Mediated Evolution of Mucispirillum schaedleri, a Core Member of the 839 Murine Gut Microbiota 2:1–15. 840 90. Brinchmann MF, Rajan B, Fernandes JMO, Caipang CMA, Rombout 841 JHWM, Kiron V. 2013. Atlantic cod (Gadus morhua) skin mucus proteins 842 - Focus on lectins. Fish Shellfish Immunol 34:1641. 843 FAO_Fisheries. 1982. Synopsis of the biological data on Dolphins-Fishes, 91. 844 Coryphaena hippurus Linnaeus and Coryphaena equiselis Linnaeus. 845 92. Hampson DJ, Ahmed N. 2009. Spirochaetes as intestinal pathogens: 846 Lessons from a *Brachyspira* genome. Gut Pathog 3:1–3. 847 93. Westerman LJ, Boer RF De, Roelfsema JH, Friesema IHM, Kortbeek LM, 848 Wagenaar JA, Bonten MJM. 2013. Brachyspira Species and Gastroenteritis in Humans. J Clin Microbiol 51:2411-2413. 849 850 Oxberry SL, Trott DJ, Hampson DJ. 1998. Serpulina pilosicoli, waterbirds 94. 851 and water: potential sources of infection for humans and other animals. 852 Epidemiol Infect 121:219–225. 853 Thompson FL, Hoste B, Thompson CC, Goris J, Gomez-Gil B, Huys L, 95. 854 Vos P De, Swings J. 2002. Enterovibrio norvegicus gen. nov., sp. nov., 855 isolated from the gut of turbot (Scophthalmus maximus) larvae: a new 856 member of the family Vibrionaceae. Int J Syst Evol Microbiol 52:2015-857 2022. 858 Thompson FL, Thompson CC, Naser S, Hoste B, Vandemeulebroecke K, 96. Munn C, Bourne D, Swings J. 2005. Photobacterium rosenbergii sp. nov. 859 860 and Enterovibrio coralii sp. nov., vibrios associated with coral bleaching. 861 Int J Syst Evol Microbiol 55:913–917. 862 Pujalte MJ, Sitja-Bobadilla A, Alvarez-Pellitero P, Garay E. 2003. 97.

863		Carriage of potentially fish-pathogenic bacteria in Sparus aurata cultured
864		in Mediterranean fish farms. Dis Aquat Organ 54:119–126.
865	98.	Company R, Sitja-Bobadilla A, Pujalte MJ, Garay E, Alvarez-Pellitero P,
866		Perez-Sanchez J. 1999. Bacterial and parasitic pathogens in cultured
867		common dentex, <i>Dentex dentex</i> L. J Fish Dis 22:299–309.
868	99.	Pascual J, Macia MC, Arahal DR, Garay E, Pujalte MJ. 2009. Description
869		of <i>Enterovibrio nigricans</i> sp. nov., reclassification of <i>Vibrio calviensis</i> as
870		<i>Enterovibrio calviensis</i> comb . nov . and emended description of the genus
871		<i>Enterovibrio</i> Thompson et al . 2002. Int J Syst Evol Microbiol 59:698–
872		704.
873	100	Org E, Parks BW, Joo JWJ, Emert B, Schwartzman W, Kang EY,
874	100.	Mehrabian M, Pan C, Knight R, Gunsalus R, Drake TA, Eskin E, Lusis
875		AJ. 2015. Genetic and environmental control of host-gut microbiota
875		interactions. Genome Res 25:1558–1569.
	101	
877	101.	Goodrich JK, Davenport ER, Waters JL, Clark AG, Ley RE. 2016. Cross-
878		species comparisons of host genetic associations with the microbiome.
879		Science (80-) 352:532–535.
880	102.	McFall-Ngai M. 2007. Care for the community. Nature 445:2007.
881		
882		
883	Tabl	00
	Ian	

885 **Table 1: Species collected and sample location.**

The table lists the species, their Latin name, ecotype (for Atlantic cod), sampling location, number of specimens (*n*), and the abbreviation used in the current study.

889

Species	Latin name	Ecotype	Sampling location	n	Abbreviation
Atlantic cod	Gadus morhua	Northeast Arctic cod	Lofoten	10	NEAC
Atlantic cod	Gadus morhua	Norwegian coastal cod	Lofoten	10	NCC
Atlantic cod	Gadus morhua	Norwegian coastal cod	Oslo fjord	2	NCC_Oslo
Poor cod	Trisopterus minutus		Oslo fjord	5	PC
Norway pout	Trisopterus esmarkiii		Oslo fjord	4	NP
Northern silvery pout	Gadiculus thori		Oslo fjord	3	NSP

890

891

892 Table 2: Overview of individual metagenomic sequence data from gadoid

893 intestines. The table shows per sample the number of original reads, the

percentage of reads remaining after trimming and filtering, percentage of host DNA, percentage of bacterial DNA and the final number of reads used in the microbiome analysis. PhiX- and human-derived DNA sequences represent a negligible proportion, and are excluded from the table. The bottom rows show total or mean values per column. On average, 42.2% of the quality filtered reads per sample are used for microbiome analysis. For details, see table S7.

900

Sample	Raw reads	After quality trimming/filtering (%)	Host DNA (%)	Bacterial DNA (%)	Final reads
NCC_01	10 883 740	85.9	87.3	12.7	1 187 649
NCC_02	11 140 950	87.9	62.2	37.8	3 699 538
NCC_03	9 891 322	90.2	41.2	58.8	5 249 515
NCC_04	10 587 865	86.9	85.2	14.8	1 364 663
NCC_05	8 423 091	89.1	57.7	42.3	3 171 737
NCC_06	10 879 319	89.6	30.5	69.5	6 772 948
NCC_07	10 082 237	91.8	31.3	68.7	6 361 506
NCC_08	9 114 703	87.3	80.5	19.5	1 549 210
NCC_09	11 105 189	89.1	62.2	37.8	3 733 846
NCC_10	11 140 743	84.7	86.0	14.0	1 320 875
NEAC_01	13 120 072	89.7	53.6	46.4	5 463 098
NEAC_02	12 119 926	89.6	56.8	43.2	4 687 565
NEAC_03	11 981 093	89.2	54.4	45.6	4 869 722
NEAC_04	12 618 529	91.1	33.5	66.5	7 646 256
NEAC_05	12 154 047	87.6	74.2	25.8	2 747 042
NEAC_06	13 883 762	88.4	59.5	40.5	4 971 507
NEAC_07	12 149 049	89.1	56.9	43.1	4 666 533
NEAC_08	11 861 852	88.7	64.0	36.0	3 787 155
NEAC_09	11 131 413	85.7	75.1	24.9	2 378 591
NEAC_10	15 483 018	83.7	82.9	17.1	2 221 214
OO_cod_01	8 047 125	83.7	85.9	14.1	949 039
IO_cod_01	9 716 392	90.2	43.7	56.3	4 937 260
NP_01	15 621 734	84.7	51.6	48.4	6 400 120
NP_02	16 548 297	90.2	18.1	81.9	12 224 903
NP_03	16 608 312	78.3	72.6	27.4	3 568 206
NP_04	13 459 929	83.0	61.5	38.5	4 300 178
PC_01	10 743 586	87.7	30.5	69.5	6 550 868
PC_02	18 982 339	81.0	29.6	70.4	10 833 201

PC_03 9 420 298 84.1 54.9 45.1 PC_04 9 623 591 87.7 30.0 70.0 PC_05 19 630 680 77.7 55.0 45.0 NSP_01 14 283 994 73.2 67.5 32.5 NSP_02 14 527 770 76.3 63.8 36.2 NSP_03 15 261 446 80.9 66.6 33.4	Mean:	12 418 453	86.0	57.8	42.2	
PC_049 623 59187.730.070.0PC_0519 630 68077.755.045.0NSP_0114 283 99473.267.532.5NSP_0214 527 77076.363.836.2	Total:	422 227 413				155 481 582
PC_049 623 59187.730.070.0PC_0519 630 68077.755.045.0NSP_0114 283 99473.267.532.5	NSP_03	15 261 446	80.9	66.6	33.4	4 123 131
PC_049 623 59187.730.070.0PC_0519 630 68077.755.045.0	NSP_02	14 527 770	76.3	63.8	36.2	4 008 926
PC_04 9 623 591 87.7 30.0 70.0	NSP_01	14 283 994	73.2	67.5	32.5	3 396 962
	PC_05	19 630 680	77.7	55.0	45.0	6 861 474
PC_03 9 420 298 84.1 54.9 45.1	PC_04	9 623 591	87.7	30.0	70.0	5 908 283
	PC_03	9 420 298	84.1	54.9	45.1	3 568 861

⁹⁰¹

902

903 **Table 3: The 10 most abundant bacterial orders in the intestinal** 904 **microbiomes of gadoid species and ecotypes.** The table shows the mean 905 relative abundance (%) of the ten most abundant orders in each of the species or 906 ecotypes. *Vibrionales* is indicated in bold. The asterix indicates that these 907 ecotypes belong to the same species (*Gadus morhua*)

908

NEAC*	NCC*	РС	NP	NSP
Bacteroidales (21.39)	Vibrionales (75.69)	Vibrionales (44.13)	Vibrionales (78.57)	Brachyspirales (15.88)
Vibrionales (16.83)	Alteromonadales (4.34)	Clostridiales (11.16)	Clostridiales (3.36)	Clostridiales (14.14)
Clostridiales (11.66)	Clostridiales (3.47)	Mycoplasmatales (8.92)	Alteromonadales (2.11)	Brevinematales (7.11)
Brevinematales (7.43)	Fusobacteriales (3.46)	Alteromonadales (5.15)	Enterobacterales (2.04)	Deferribacterales (4.81)
Bacillales (2.64)	Oceanospirillales (1.56)	Enterobacterales (2.97)	Bacteroidales (1.02)	Bacillales (4.44)
Alteromonadales (2.61)	Enterobacterales (1.19)	Bacteroidales (2.09)	Mycoplasmatales (0.87)	Fusobacteriales (1.94)
Flavobacteriales (2.17)	Bacteroidales (0.92)	Bacillales (1.81)	Oceanospirillales (0.64)	Desulfovibrionales (1.70)
Fusobacteriales (1.62)	Bacillales (0.60)	Oceanospirillales (1.18)	Burkholderiales (0.47)	Lactobacillales (1.57)
Brachyspirales (1.23)	Pseudomonadales (0.33)	Lactobacillales (0.95)	Bacillales (0.43)	Rhizobiales (1.40)
Deferribacterales (1.23)	Flavobacteriales (0.27)	Burkholderiales (0.84)	Pseudomonadales (0.43)	Spirochaetales (1.28)

909

910

911 Table 4: Effects of covariates on the intestinal microbial diversity (alpha diversity) of gadoid species and ecotypes. Results from the optimal linear regression models used in testing for significant effects of covariates on within-914 sample (alpha) diversity based on non-normalized, order-level sequence counts. 915 Population (species/ecotype) is the only covariate with a significant effect, and

- 916 estimates are given relative to NCC. Significant effects (p < 0.05) are indicated
- 917 in bold.
- 918

	Shannon		Simpson		Inv. Simpson	
	Estimate	p-value	Estimate	p-value	Estimate	p-value
Intercept	1.08	0.0000	0.38	0.0000	1.93	0.0001
NEAC	1.69	0.0000	0.48	0.0000	5.62	0.0000
NP	0.06	0.8367	-0.01	0.8784	-0.25	0.7476
PC	1.18	0.0001	0.37	0.0002	2.44	0.0017
NSP	2.36	0.0000	0.53	0.0000	10.03	0.0000

- 919
- 920

921 Table 5: PERMANOVA analysis of intestinal microbial diversity from gadoid species and ecotypes. The table shows R² values, *p*-values and adjusted 922 923 *p*-values for pairwise comparisons of community composition (beta diversity) between the different species or ecotypes using PERMANOVA. The tests are 924 925 based on Bray-Curtis dissimilarity calculated from order-level, normalized 926 sequence counts. *p*-values are adjusted for multiple testing by the Holm method. 927 Significant differences (p < 0.05) are indicated in bold. Genus- and species-level 928 results are shown in Table S6.

929

Populations	\mathbf{R}^2	p.value	p.adjusted
NEAC vs. NC	C 0.71	0.0001	0.0005
NEAC vs. PC	0.53	0.0002	0.0018
NEAC vs. NP	0.70	0.0010	0.0076
NEAC vs		0.00/1	0.0207
NSP	0.50	0.0041	0.0207
NCC vs. PC	0.42	0.0026	0.0182

NCC vs. NP	0.04	0.7138	0.7138
NCC vs. NSP	0.80	0.0035	0.0207
PC vs. NP	0.53	0.0233	0.0740
PC vs. NSP	0.80	0.0185	0.0740
NP vs. NSP	0.94	0.0286	0.0740

930

- 931
- 932 Figures
- 933

Figure 1: The intestinal microbiomes obtained from a range of gadoid species and ecotypes

936 (A) Map of sampling locations in Norway, Europe. Northeast Arctic cod 937 (NEAC), and Norwegian coastal cod (NCC) were obtained from the Lofoten. 938 NCC (two individuals), poor cod (PC), Norway pout (NP), and northern silvery 939 pout (NSP) were obtained from the Oslo Fjord. (B) Non-metric 940 multidimensional scaling (NMDS) plot of non-normalized, order-level sequence 941 counts from the intestinal microbiomes of all samples. Each point represents an 942 individual sample, and the species or ecotypes are indicated by different shapes 943 and colors. The stress value of the NMDS plot is 0.14.

944

945 Figure 2: Taxonomic composition of the fish intestinal microbiomes

946 (A) Relative abundance of metagenomic shotgun sequences classified at the 947 order level (93%). Colors represent the 28 orders with highest relative 948 abundance, sequences assigned to other orders or viruses, and sequences 949 classified above order level. Numbers along the x-axis indicate the individual 950 samples of the different species/ecotypes. (B) Relative abundance of 951 metagenomic shotgun sequences classified at the species level (66%). The plot 952 includes the most highly abundant species, and other members of their parent 953 bacterial genera ("other" categories) in the different fish species/ecotypes.

954 Numbers along the x-axis indicate the individual samples of the different
955 species/ecotypes. The star denotes the *P. kishitanii* species reclassified from *P. phosphoreum*.

957

958 Figure 3: Within-sample microbial diversity in the gadoid species and

959 ecotypes

960 Boxplots of Shannon (A) and Inverse Simpson (B) diversity in the fish

961 species/ecotypes. Each individual is represented by a point, and the individuals

are grouped and colored by species and ecotype. The middle band represents the

963 median, while the upper and lower band shows the 75th and 25th percentile. The

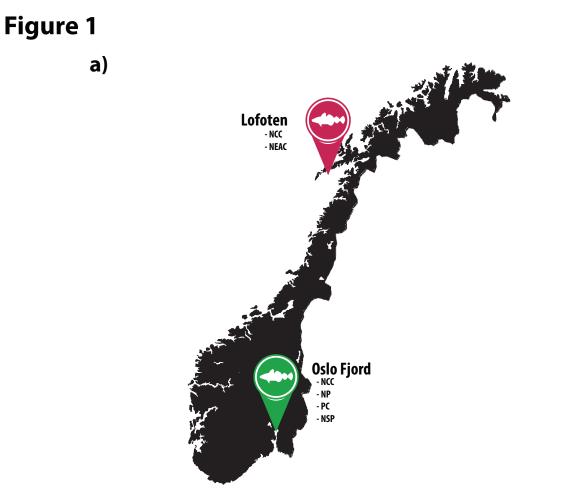
964 boxplots include the minimum and maximum alpha diversity values.

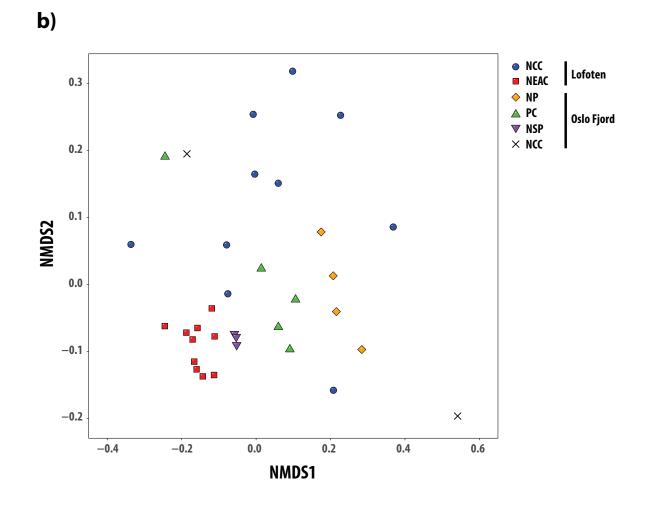
965

Figure 4: SNV variation analysis of the most abundant bacterial genomes in the microbiomes of gadoid species

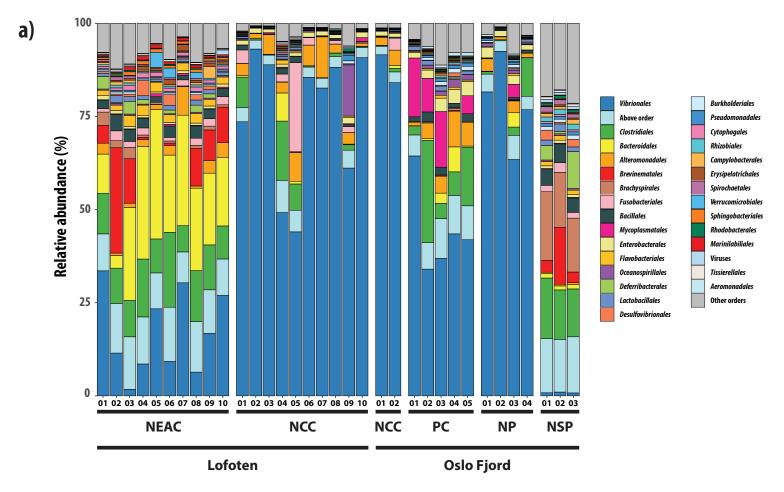
968 For both of the genomes, the figure displays (A) read coverage per single 969 nucleotide variant (SNV) position in each sample from the different 970 species/ecotypes (mean coverage on right side of plot), (B) variation density 971 (number of variable positions per 1000 kbp. reported in each individual sample, 972 independent of coverage in the other samples) per sample (maximum value 973 indicated). The y-axis of the coverage- and variation density plots are scaled 974 across the genomes. (C) Heatmap of a randomly chosen subset of 400 SNVs. In 975 the heatmap, each row represents a unique variable nucleotide position, where 976 the color of each tile represents the two most frequent competing nucleotides in 977 that position. The shade of each tile represents the square root-normalized ratio 978 of the most frequent two bases at that position (i.e., the more variation in a 979 nucleotide position, the less pale the tile is); see legend in the bottom of the 980 figure. (D) Principal component analysis (PCA) plot of the SNV distribution 981 (within-species variation) among the different samples. Each sample is 982 represented by a dot, and colored according to species or ecotype membership.

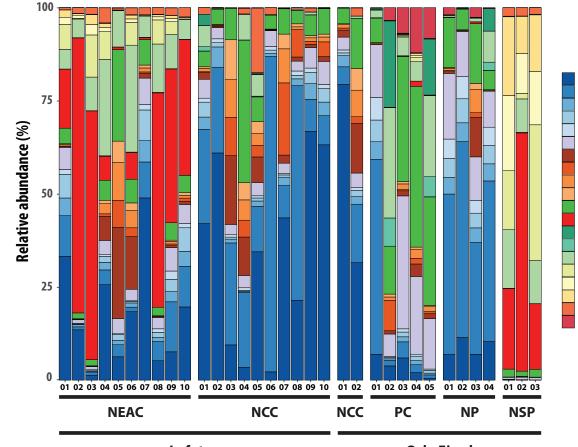
Half-circles to the right of the legend indicate species or ecotypes with
significantly different within-species variation (i.e. different strains). (E)
Relative abundance of the different samples used in variation analysis. The bars
are colored according to the SNV plot in (D).











bacterium iliopiscarium Photobacterium kishitanii* Photobacterium piscicola Photobacterium toruni Photobacterium aquimaris Photobacterium other Aliivibrio wodanis Aliivibrio logei Aliivibrio fischeri *Aliivibrio* other Vibrio spp. Brevinema sp. Clostridium mediterraneense Clostridium sp. ND2 *Clostridium* other Mucispirillum sp. Brachyspira pilosicoli Brachyspira sp. CAG:700 Cetobacterium spp. Enterovibrio norvegicus

Lofoten

b)



a)



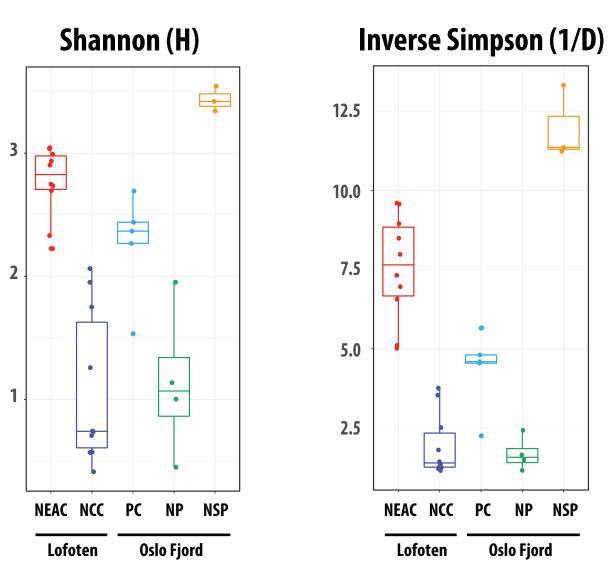


Figure 4

