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Spatial Analysis of Bacteria in Brackish Lake Sediment

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• Highlights:

- Macro to micro scale analysis of bacterial diversity
- Spatial heterogeneity of physico-chemical variables linked with bacterial distribution
- Salinity plays a major role in bacterial distribution in sediment from a brackish lake
- Streamlines from fish farms are related to decreases in bacterial diversity and increases in pathogens
- Coupling macro-analytical tools (GIS) and micro tools (16S rRNA gene sequencing) to determine the spatial distribution of bacteria in sediment.

Spatial analysis of bacteria in brackish lake sediment

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1 Abstract

- 2 The spatial distribution of bacterial communities inhabiting sediment is heterogeneous at different spatial scales, but mostly
- 3 unexplored. Here, it is postulated that the heterogeneity of the bacterial community composition varies at the same scale of
- 4 the heterogeneity of sediment chemical properties. The large spatial scale (km) diversity in sediment from a brackish water
- 5 lagoon (Chilika Lake, India) is studied, considering the large scale physical and chemical characteristics of land cover,
- 6 climate, pH, and salinity. Seventy-two samples (24 stations, 3 seasons: winter, rainy, and summer) of sediment from Chilika
- 7 Lake were analyzed by 16S ribosomal ribonucleic acid (rRNA) gene sequencing, and the relations with land cover and other
- physico-chemical parameters are discussed. More samples were collected after a severe cyclonic storm (Phailin) passed near
 the lagoon in 2013, to see the impact of the tropical storm on the spatial and temporal distribution of bacteria in the sediment.
- 10 The results demonstrate clear spatial relations between physico-chemical parameters (e.g., salinity), land surfaces (e.g.,
- 11 drainage area), and the distribution of sediment microbial communities.

12

14 1. Introduction

15 Microorganisms play a major role in global biogeochemical cycles of carbon, nitrogen, and phosphorus. With the growth rate 16 of all microorganisms on earth estimated at over 10^{30} cells/yr, these microorganisms are considered major drivers of 17 geochemical cycling (Withman et al., 1998). Soils and sediment are among the largest microbial reservoirs. In addition to the 18 high number of microorganisms, soils also have relatively high heterogeneity of physical, chemical, and biological 19 characteristics (Beare et al., 1995; Ramette, 2007). This heterogeneity might influence reaction rates through the distances 20 that metabolites must move between different bacteria and archaea in soils, yet, little is known about the spatial distribution 21 of bacteria in soils (and less about sediment) and the drivers of the spatial heterogeneity. Several studies have focused on soil 22 (Dechesne et al., 2003; Grundmann, 2004; Remenant et al., 2009; Vos et al., 2013) and sediment (Córdova-Kreylos et al., 23 2006; Piza et al., 2005) in order to understand spatial distribution of bacteria at the micro-scale and on soils at the macro-24 scale (Fierer & Jackson, 2006; Griffiths et al., 2009).

25 Part of the difficulty is the physical-chemical description of the different samples at the micro-scale. The use of sediment to 26 study spatial-scale effects on microbial community distribution simplifies the micro-heterogeneity of physico-chemical 27 parameters due to the sediment pore water, which increases local nutrient and metabolite diffusion compared to unsaturated 28 soils (Urban et al., 1997). Some studies have evaluated the bacterial composition and spatial distribution in sediment of 29 brackish lakes (Pramanik et al., 2015; Thureborn et al., 2013; Webster et al., 2015; Zaitseva et al., 2014). These studies 30 identified specific functions and microbial communities associated with different types of sediment environments. One 31 marine sediment study identified large-scale shifts in microbial community structure in sediment pore water along a salinity 32 gradient in the Baltic Sea (Herlemann et al., 2011). While salinity clearly has an influence on microbial community 33 composition, several other physical-chemical characteristics also could play important roles. Some of these characteristics 34 could be driven by global land use practices. It is hypothesized that the spatial distribution of the microbial community in 35 large coastal lakes could be modeled in part by sample characteristics and by global land use as determined by satellite data 36 and geographic information systems (GIS). As lake sediment has less physical and chemical heterogeneity than river 37 sediment (Brönmark & Hansson, 2005) and result from the indigenous production and the transport of terrestrial material 38 (organic or inorganic compounds) from the drainage area (Beaulne et al., 2012; Teisserenc et al., 2010), the sub-hypothesis of 39 this paper is that the physico-chemical parameters having a spatial gradient at the scale of the lake (e.g., salinity in this case) 40 will drive bacterial composition and diversity. The combination of marine and fresh water creates a wide range of habitat 41 types that can sustain relatively high microbial diversity in Chilika Lake. This paper examines marine and riverine influences 42 through the use of GIS coupled with standard multivariate analyses.

43 2. Material and Methods

44 2.1. *Study area*

Chilika Lake, which is the second largest brackish lagoon in the world and the largest in Asia, is situated on the east coast of
India (between 19°28' and 19°54' N and 85°05' and 85° 38' E). Since 1981, the site has been designated as an Indian wetland

47 of international importance (Ramsar Site). The size of the lagoon fluctuates significantly during a year from 906 km² 48 (summer) to 1165 km² (monsoon). The hydrological pattern of Chilika Lake is influenced by three subsystems, namely, 49 Mahanadi River distributaries, streams of the western catchment that bring in freshwater to the lake, and the Bay of Bengal 50 that contributes highly saline sea-water. The lagoon is a collection of very shallow marine, brackish, and freshwater 51 ecosystems. The whole area is categorically organized into four sectors. The northern sector, which receives direct discharge 52 of freshwater from the Mahanadi Delta; the central sector intermixing zone, which is brackish; the southern sector, which has 53 higher salinity levels as compared to central sector; and the outer channel, which exchanges water between the lagoon and the 54 sea. Due to a high salinity gradient, the lake hosts a wide range of macro-organism biodiversity: 314 species of fish, 224 55 species of water birds, and 729 angiosperms. That high biodiversity sustains an important population of fishermen estimated 56 at about 140,000 individuals (Kumar & Pattanaik, 2012). After a decrease in fishing productivity in the 1990's, the lake water 57 quality has been monitored and the lake mouth to the outer channel was reopened.

58

59 2.2. Sampling and physico-chemical analyses

60 The monitoring of the sediment and water was done at 24 stations spread over the entire lake (Fig. 1 - monitoring and 61 sampling stations) in summer (March to June 2011), monsoon (July to October 2011), and winter seasons (November 2011 to 62 February 2012). Sediment and water samples were collected from stations from all four ecological sectors; northern, central, 63 southern, and outer channel (Fig. 1). Bottom surface sediment samples were collected with a Van Veen type of grab sampler 64 (KC, Denmark,) from each station during summer, monsoon, and winter seasons. Both sediment and water (water column) 65 samples were immediately transferred to sterile plastic 500 ml bottles and transported to the laboratory on ice and stored at 66 4ºC. Sediment samples were partially dewatered to improve lysis efficiency and homogenized manually with a sterile mortar 67 and pestle inside a laminar flow hood.

68 For each station, pH, conductivity (mS/cm), turbidity (NTU), salinity (parts per thousand; ppt) and temperature of air and 69 water (°C) were measured onboard immediately with a water quality probe analyser (TOA DKK 24, Japan). Depth and 70 transparency/turbidity (cm) of the water column were measured with a measuring tape and Secchi disk, respectively. 71 Biochemical oxygen demand (ppm) and dissolved oxygen (ppm) were measured with an automated portable probe analyzer 72 (VSI 07, VSI Electronics, Mohali, India). Concentration of nitrite, nitrate, ammonia, phosphate, silicate, and total iron were 73 measured on the collected water and total organic carbon on sediment samples (APHA, 1998). In order to determine seasonal 74 effects, each station was sampled 3 times, during summer, winter, and monsoon seasons. They were also sampled after a 75 major storm (as described in section 2.4).

76 Fig. 1

77

78 2.3. GIS analysis (Land surfaces, hydrogeology and spatial analysis)

The watershed boundaries were determined using a Digital Elevation Model (DEM) from Shuttle Radar Topography Mission (SRTM) images (U.S. National Aeronautics and Space Administration, NASA) using Geographical Resources Analysis Support System (GRASS), an open-source GIS software (Beaulne, 2008). Hydrological modeling was also done in order to create stream, slope, and drainage layers (GRASS). The sub-watersheds were mergedinto four large watersheds to correspond to the scale of the sediment and water sampling.

Land cover classes were assigned using four Landsat Thematic Mapper 7 (TM) satellite images (July 2003) with a 27 m resolution. Land cover classifications were then determined using an unsupervised maximum likelihood analysis method using ArcGIS software. The percentage of forested area, cultivated land, urban area, bare soil, and wetland were calculated for each of the created watersheds (Fig. 2). Areas of aquaculture (mainly shrimp and fish landing centers) also were determined from satellite images (Centre National d'Etudes Spatiales (CNES), 2012, distribution *Astrium* service's spot image). These were considered to be important activities in Chilika Lake, but problematic in terms of water quality in other similar environments (Pushparajan & Soundarapandian, 2010).

91 Fig. 2

92 Interpolations were done with GRASS software for the physico-chemical parameters measured for each sampled point and
 93 each season to find geographical and temporal concentration gradients.

94

95 2.4. DNA analysis

96 Total DNA was extracted from 0.5 g of sediment sample in duplicate using FastDNA® Spin Kit for soil DNA extraction (MP 97 Biomedicals, Santa Ana, CA, U.S.) as per the manufacturer's protocol. DNA was extracted in duplicates for each sampling 98 station and pooled to one tube per individual station and season. Quality and quantity of metagenomic DNA were verified 99 using 1.2% of agarose gel (0.5XTAE) and Nanodrop (Epoch, BioTek Instruments, U.S.) analysis (A260/A280). DNA from a 100 total of 72 samples from the 24 station (3 seasons) from 2011 and 2012 and twelve more samples after the "Phailin", a very 101 severe cyclonic storm (VSCS) which made landfall near 19.26⁰ N / 84.82⁰ E (near Gopalpur, Odisha) at 1600 UTC on 12th 102 October, 2013 were extracted to determine bacterial diversity and community composition. Total DNA extracted from the 103 sediment was amplified for 16S rRNA genes by Polymerase Chain Reaction (PCR). The hypervariable (V4-V6) region of the 104 16S rRNA genes was amplified from using the bacterial primer pair 515F and 1061R (16S-0515F 5'-105 TGYCAGCMGCCGCGGTA-3' 16S-1061R 5'-TCACGRCACGAGCTGACG-3' ~560bp V4-V6 region (M774, Promega, 106 Madison, WI, U.S.)). PCR was done with the modified primers containing an adapter and a barcode sequence (Schloss et al., 107 2009). Paired end 454 pyrosequencing was done on the GS-FLX 454 Titanium platform.

108

109 2.5. Analysis of the community composition

The reads were analyzed through the Quantitative Insights Into Microbial Ecology (QIIME) pipeline (http://qiime.org) (default parameters) and taxonomic identity of each read was assigned using the Bayesian rRNA gene classifier at the 80% confidence threshold and the Ribosomal Database Project (RDP) database (Caporaso et al., 2010; Cole et al., 2009; Wang et al., 2007). Sequences that could not be classified to at least a kingdom level were excluded from subsequent analysis. For phylotyping of 16S rRNA gene data, sequences were aligned using the QIIME pyrosequencing pipeline. Operational taxonomic units (OTUs) were determined at phylum (80%), class (90%), order (92%), genus (95%), and species (97%) level.

116 2.6. Statistical and multivariate analysis

Statistical analyses were done using the JMP 11 software (JMP[®], Version 11. SAS Institute Inc., Cary, NC, 1989-2019). In order to find the link between physico-chemical parameters and relative abundance of OTU's, a large set of relations was evaluated at different taxonomic levels. To find similarities of community composition between samples, CANOCO 4.5 (CANOCOtm CanoDrawtm and the Cornell Ecology Programs, Microcomputer Power, Ithaca, U.S.) was used to draw two major components of the principal component analysis (PCA). Canonical correspondence analysis (CCA) also was done to find which variables are driving the separation of the samples in terms of community composition in the environmental dataset.

124

125 **3. Results**

126

127 3.1. Physico-chemical distribution in the lake and spatial distribution of bacteria

128 In order to identify correlations between physico-chemical parameters and the bacterial spatial distribution, the spatial 129 distribution of the physico-chemical parameters was first examinated. The pH varied between 6.9 and 9.5. No spatial pattern 130 was found in the pH distribution. Only one sampled point (KG) had a pH under 7.8 and only during the monsoon season (the 131 pH was 8.6 during the summer at the same point), probably due to the increase of fresh water (lower pH) coming from the 132 watershed during the monsoon season. Turbidity was in a range between 1.1 to 171 NTUs, but again the nonlinear 133 geographical gradient (patchy distribution) mainly depended on the seasonal effect. The highest turbidity was found in almost 134 all post-Phaillin (cyclonic storm) samples because the cyclonic storm stirred up bottom sediment. While salinity and 135 conductivity were correlated, the salinity was focused on because it can have a strong effect on the microbial spatial 136 distribution and diversity. Salinity was measured between 0.1 and 32.3 ppt (g/L). Both a strong temporal variation and a 137 spatial gradient in salinity values were observed.

The interpolated maps of the geographical variability of the salinity measured for each station during the three seasons showed the highest seasonal variation in the mixing zone near the output channel (Fig. 3a, 3b, 3c). Nitrite, nitrate, ammonia, phosphate, silicate, total iron, and total organic carbon varied considerably between samples at different locations and for different seasons from the same location. The post cyclonic storm samples generally had higher nutrient concentrations due, 142 in part, to higher flux coming from the watersheds. The concentrations of nitrite (0.04 to 8.6 µmol/L), nitrate (0.52 to 25.67

143 μ mol/L) and ammonia (7.31 to 206.41 μ mol/L) generally were low.

144 Fig. 3

145 3.2. GIS analysis (land surfaces, hydrogeology and spatial analysis)

Interpolation of physico-chemical parameters was done for each season. Only salinity showed a clear spatial gradient, although the gradient varied from season to season. Interpolation of the bacterial diversity (Shannon index at the genus level) (Morris et al., 2014) was also done using the GRASS software. Figure 3d shows the interpolation of bacterial diversity after merging the samples from the three seasons. A significant "hot spot" for bacterial diversity can be seen. The geographical area where the highest microbial diversity was measured (Figure 3d) also was where the salinity had the highest variability during the year (seasonal effect, Fig 3a, 3b, 3c).

152 3.3- Statistical analysis

Multivariate analysis (PCA, CCA) was done with CANOCO v4.5 to test in part whether one of the main drivers of the microbial community variability was salinity. A separation of the samples was observed following principally the variation of the salinity and conductivity (Fig. 4). The spatial effect also can be responsible for the differences in community structure as all the samples with the higher salinity were located in the same geographical zone, i,e. the outer channel. A PCA also was done to determine the seasonal effect on the bacterial community structure. No clear evidence of a major shift in the community structure as a function of season was observed.

159 Fig. 4

160 In order to test the hypothesis that the rare severe storms smooth out spatial heterogeneity, twelve samples were collected just 161 after a cyclonic storm in October 2013. The samples were collected one week after the storm, 3 samples were taken in each 162 of the four zone of the lake. The samples with high salinity before (dark green points) and after the storm (dark red points) 163 had similar community structures (Fig. 5). The low salinity samples taken after the storm (light red points) had different 164 community structures than those sampled before the storm (light green points).

165 Fig. 5

The same twelve sample points wer compared before and after the storm (Fig. 6). A small shift in the community structure occurred after the storm probably due to the mixing of the sediment (sediment was sampled at a water depth between 1 and 3.2 m) and a major influx of organic matter from the watershed. The sample points located near the ocean were not significantly different after the storm. The major change during the storm is the intrusion of saltwater in to the lake, affecting principally the zone that used to have low salinity.

171 Fig. 6

172 3.4. Analysis of the microbial community structure

173 The number of reads per samples were between 5,387 to 38,265. Taxonomy was done using QIIME pipeline and the RDP 174 database. The most dominant phylum in the current sample was Proteobacteria with a relative abundance up to 0.79 and 175 never under 0.37 (average 0.573). A second more abundant phylum was the Chloroflexi (0.01 to 0.24, average 0.086). In 176 order, the next phyla were Bacteriodetes (0.006 to 0.17, average 0.047), Planctomycetes (0.001 to 0.08, average 0.038), 177 Acidobacteria (0.002 to 0.07, average 0.031), Firmicutes (0.0004 to 0.15, average 0.028), Gemmatimonadetes (0.002 to 178 0.08, average 0.027), Actinobacteria (0.004 to 0.08, average 0.023), Verrucomicrobia (0.0004 to 0.014, average 0.021) and 179 Nitrospirae (0.0005 to 0.06, average 0.017). Looking more specifically at Proteobacteria (the largely dominant phylum in all 180 the samples), a specific distribution was observed at the class level. The most dominant class of Proteobacteria was 181 Gammaproteobacteria with an average relative abundance of 0.252, followed by Deltaproteobacteria (0.178), 182 Betaproteobacteria (0.066), Alphaproteobacteria (0.045), and Epsilonproteobacteria (0.026). The high abundance of 183 Gammaproteobacteria was mainly due to four orders, Chromatiales (up to 0.19), Xanthomonadales (up to 0.12), 184 Pseudomonadales (up to 0.21) and Thiotrichales (up to 0.21). The dominance of Gammaproteobacteria is similar to what 185 have been describe in marine sediment (Fang et al. 2017). Some families and genus were also measured at a high relative 186 abundance in some of the current samples: Piscirickettsiaceae (up to 0.21 relative abundance) and Pseudomonas (up to 0.09). 187 Acidobacteria was observed at a relatively high abundance even if the pH was above 7 in almost every part of the lake. No 188 correlation between the Acidobacteria relative abundance and pH was observed.

A negative effect of the salinity on the diversity (except for the site located in the mixing zone) (Fig. 7a) also can be seen. The decrease of diversity related to salinity is similar of what have been observed in other study comparing sediment from fresh to salt water (Wang et al., 2012). A positive relation between the salinity and the relative abundance of *Proteobacteria* was found (Fig. 7b). Since the relative abundance of *Proteobacteria* was generally high but variable, a strong negative relation also was found between the abundance of *Proteobacteria* and the Shannon diversity (Fig. 7c)

194 Fig. 7

Links between physico-chemical variables and bacterial community structure were observed with the distribution of the most relatively abundant bacterial phylum and class found in the Chilika Lake sediment (i,e. between *Proteobacteria* and *Gammaproteobacteria* and salinity). With a R² of 0.2829 (p < 0.0001) (Fig. 7d), there is a trend for increasing salinity to relate to the distribution of the *Gammaproteobacteria* in the sediment of Chilika Lake, but the data noise is quite high. At the phylum level, the trend (R² = 0.1896, p < 0.0002) (Fig. 7b) was less obvious.

200 4. Discussion of the drivers of the community structure

Recent research has shown that macrobiotic ecological changes in the Chilika Lake system and its fisheries were due to the effect of water exchange between the lake and the sea that is responsible for control of the salinity, siltation, macrophyte infestation, and increase of marine forms (Ghosh et al., 2006; Sahu et al., 2014; Satyanarayana et al., 1999). It is hypothesized that the microbial community would be similarly affected. The community structure and distribution in Chlika Lake observed in the current study was similar to that observed in 16S rRNA gene studies on brackish water, in the Baltic Sea (Herlemann et al., 2011), in sediment of Sundarbans Mangrove Wetland (Basak et al., 2014), in two samples in Chilika Lake (Pramanik et al., 2015), and in a Brazilian mangrove (Andreote et al., 2012). Salinity was the only variable that was identified as having a spatial gradient of distribution at the scale of the current sampling in the lagoon. Salinity has been described as the major factor shaping microbial communities (Lozupone & Knight, 2007).

210 The distributions of other variables measured in the current study had either a much higher or a much lower spatial scale of 211 variation. Water temperature and depth that could, respectively, correspond to temporal and spatial distribution were not 212 significantly correlated to community structure, probably due to the relative homogeneity of those parameters in Chilika 213 Lake. pH has been described to have strong effect on regulation of life processes and nutrient availability in aquatic systems 214 (Tank et al., 2009). In the case of Chilika Lake (and at the scale of the current study), pH did not appear to be strongly 215 correlated to the spatial distribution of bacteria and diversity index. Although the relative abundance of Acidobacteria has 216 been described to correlate with soil pH (Fierer & Jackson, 2006; Griffith et al., 2011), no correlation between the pH and the 217 relative abundance of the phylum Acidobacteria nor with the genera within this phylum was found.

The higher scale of variation would affect all the samples, making it difficult to measure their effect on the spatial distribution of the bacteria in the sediment (climatic conditions, watershed land cover). On the other hand, the distribution and variability of nitrite, nitrate, ammonia, phosphate, silicate, total iron, and total organic carbon occurred at a much smaller scale than the scale of the current samples. Hotspots for those variables are visible at the micro-scale (Parkin, 1987; Schramm et al., 1999). The size of samples (0.5 g), the nesting of the sampling effort and the fact that the exact same sample was not used to do chemical analysis and 16S rRNA gene analysis hindered finding a correlation between those micro-scale variables and potential drivers of the microbial community structure at the scale of the current study (Fig. 8).

225 Fig. 8

226 Another limitation in the current analysis of the spatial distribution of bacteria and the identification of the physico-chemical 227 drivers is the importance of the transport of material in the lake (watershed - lake, stream). All the samples with high relative 228 abundances (more than 0.08 relative abundance) of a specific family (Chromatiales, Xanthomonadales, Pseudomonadales, 229 and Thiotrichales) or specific genus (Piscirickettsiaceae and Pseudomonas, more than 0.05 relative abundance) were 230 identified. Coupling a map of the fishfarming and a stream map (produced with Hydrological Modeling, GRASS), it was 231 observed that almost all the points with high abundance of a specific Gammaproteobacteria family or genus were located on 232 a streamline from the fishfarming areas (Fig. 9). Those four taxonomic families and two genera are known to contain some 233 pathogen species; some of them have been associated with fish pathogens in salmon fishfarming (Austin & Austin, 2007; 234 Birkbeck et al., 2011). The spatial analysis also highlighted hotspots of bacterial diversity where the seasonal variation of the 235 salinity was high. The more drastic change in salinity was mainly due to the freshwater coming into the lagoon during the 236 monsoon season. Some changes of salinity could be due to the tidal effect on a daily basis.

237 Fig. 9

238 5. Conclusion

239 This first description of the spatial distribution of bacteria in the Chilika Lake sediment has highlighted the predominance of 240 Gammaproteobacteria. Trends in the spatial and temporal distribution of that subphylum could be identified in the brackish 241 water lagoon. Hotspots of some families and genus from Gammaproteobacteria were linked to the potential transport of 242 material through the lake from fish farming areas. This analysis was made possible by applying GIS-based spatial analysis. 243 The only variable, salinity, that had a spatial pattern of distribution at the same geographic scale as the sampling effort was 244 the only one that could be correlated with the community structure. The relation between salinity and relative abundance of 245 Gamaproteobacteria was an indication that the spatial distribution of bacteria in sediment was possibly driven by this 246 physico-chemical parameter. Using only statistical approaches to identify physicochemical drivers of the community 247 structure can be meaningless if both the spatial and the temporal scale are not considered. Environmental parameters that are 248 generally insignificant can have hotspots of higher importance that explain specific niches for bacteria at the micro-scale. 249 Comparing samples of 0.5 g will not address the importance of parameters that vary at a smaller scale. At a scale larger than 250 the sample set, the variability will be integrated in all the samples, and, therefore, will not differentiate between samples.

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- 252

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260	References	
261		
262 263 264 265 266 267 268 269	1.	Andreote, F. D., Jiménez, D. J., Chaves, D., Dias, A. C. F., Luvizotto, D. M., Dini-Andreote, F., de Melo, I. S. (2012). The microbiome of Brazilian mangrove sediments as revealed by metagenomics. <i>PLoS One</i> , 7(6), e38600.
	2.	APHA, Awwa. WPCF (1998). Standard methods for the examination of water and wastewater, 1998, vol. 20, p. 39-42.
	3.	Austin, B., & Austin, D. A. (2007). Bacterial fish pathogens: Disease of farmed and wild fish. Berlin/Heidelberg, Springer Science & Business Media.
270		
271 272 273 274	4.	Basak, P., Majumder, N. S., Nag, S., Bhattacharyya, A., Roy, D., Chakraborty, A., Bhattacharyya, M. (2015). Spatiotemporal analysis of bacterial diversity in sediments of Sundarbans using parallel 16S rRNA gene tag sequencing. <i>Microbial ecology</i> , 69(3), 500-511.
275 276 277	5.	Beaulne, J. S. (2008). Modélisation de la présence de mercure dans la chair des brochets et des dorés des lacs les plus pêchés de l'Abitibi-Témiscamingue: une approche par les systèmes d'information géographique. M.S. Thesis, Environmental Sciences Department, Université du Québec à Montréal.
278		
279 280 281 282	6.	Beaulne, J. S., Lucotte, M., Paquet, S., & Canuel, R. (2012). Modeling mercury concentrations in northern pikes and walleyes from frequently fished lakes of Abitibi-Témiscamingue (Quebec, Canada): A GIS approach. <i>Boreal Environment Research</i> , <i>17</i> (3/4), 277-290.
283 284 285 286 287 288	7.	Beare M.H., Coleman D.C., Crossley, D.A. Jr., Hendrix P.F., Odum E.P. (1995). A hierarchical approach to evaluating the significance of soil biodiversity to biogeochemical cycling. <i>Plant and Soil</i> , 170; 5-22.
	8.	Birkbeck, T. H., Feist, S. W., & Verner–Jeffreys, D. W. (2011). Francisella infections in fish and shellfish. <i>Journal of Fish Diseases</i> , <i>34</i> (3), 173-187.
289	9.	Brönmark, C., & Hansson, L. A. (2005). The biology of lakes and ponds. Oxford, Oxford University Press.
290 291 292 203	10.	Caporaso, J. G., Kuczynski, J., Stombaugh, J., Bittinger, K., Bushman, F. D., Costello, E. K., Knight, R. (2010). QIIME allows analysis of high-throughput community sequencing data. <i>Nature Methods</i> , 7(5), 335-336.
294 295 296 297	11.	Cole, J. R., Wang, Q., Cardenas, E., Fish, J., Chai, B., Farris, R. J., Tiedje, J. M. (2009). The Ribosomal Database Project: improved alignments and new tools for rRNA analysis. <i>Nucleic Acids Research</i> , <i>37</i> (suppl 1), D141-D145.
298 299 300 301	12.	Córdova-Kreylos, A. L., Cao, Y., Green, P. G., Hwang, H. M., Kuivila, K. M., LaMontagne, M. G., Scow, K. M. (2006). Diversity, composition, and geographical distribution of microbial communities in California salt marsh sediments. <i>Applied and Environmental Microbiology</i> , 72(5), 3357-3366.
302 303 304 305	13.	Dechesne, A., Pallud, C., Debouzie, D., Flandrois, J.P., Vogel, T.M., Gaudet, J.P., & Grundmann, G.L. (2003). A novel method for characterizing the microscale 3D spatial distribution of bacteria in soil, <i>Soil Biology and Biochemistry</i> , 35: 1537-1546
306 307 308	14.	Fang, H., Chen, Y., Huang, L., & He, G. (2017) Analysis of biofilm bacterial communities under different shear stresses using size-fractionated sediment. <i>Scientific Reports</i> , 7, 1299.
309 310 311	15.	Fierer, N., & Jackson, R. B. (2006). The diversity and biogeography of soil bacterial communities. <i>Proceedings of the National Academy of Sciences</i> , <i>103</i> (3), 626-631.
312 313 314	16.	Ghosh, A. K., Pattnaik, A. K., & Ballatore, T. J. (2006). Chilika Lagoon: Restoring ecological balance and livelihoods through re-salinization. <i>Lakes & Reservoirs: Research & Management</i> , 11(4), 239-255.

17. Griffiths, G., Thomson, R.I., James, B.C., Bell, T., Bailey, M., & Whiteley, A. S. (2011). The bacterial biogeography of British soils. *Environmental microbiology*, 13(6), 1642-1654.

- 18. Grundmann, G. L. (2004). Spatial scales of soil bacterial diversity-The size of a clone. *FEMS Microbiology Ecology*, 48(2), 119-127.
- Herlemann, D. P., Labrenz, M., Jürgens, K., Bertilsson, S., Waniek, J. J., & Andersson, A. F. (2011). Transitions in bacterial communities along the 2000 km salinity gradient of the Baltic Sea. *The ISME Journal*, 5(10), 1571-1579.
- 20. Kumar, R., & Pattnaik, A. K. (2010). Integrated planning management for Chilika Lake. *Chilika Newsletter*, *5*, 1-12.
 - 21. Lozupone, C. A. & Knight, R. (2007) Global patterns in bacterial diversity. *Proceeding of the National Academy of Sciences*. 104, 11436–11440.
 - 22. Morris, E. K., Caruso, T., Buscot, F., Fischer, M., Hancock, C., Maier, T. S., ... Socher, S. A. (2014). Choosing and using diversity indices: insights for ecological applications from the German Biodiversity Exploratories. *Ecology and evolution*, *4*(18), 3514-3524.
 - 23. Parkin, T. B. (1987). Soil microsites as a source of denitrification variability. Soil Science Society of America Journal, 51(5), 1194-1199.
 - 24. Piza, F. F., Prado, P. I., & Manfio, G. P. (2005). Investigation of bacterial diversity in Brazilian tropical estuarine sediments reveals high actinobacterial diversity. *Antonie Van Leeuwenhoek*, 86(4), 317-328.
- Pramanik, A., Basak, P., Banerjee, S., Sengupta, S., Chattopadhyay, D., & Bhattacharyya, M. (2015). Pyrosequencing based profiling of the bacterial community in the Chilika Lake, the largest lagoon of India. *Genomics Data*, 4, 112-114.
- 26. Pushparajan, N., & Soundarapandian, P. (2010). Recent farming of marine black tiger shrimp, Penaeus Monodon (Fabricius) in South India. *African Journal of Basic & Applied Sciences*, 2(1-2), 33-36.
- 27. Ramette, A. (2007). Multivariate analyses in microbial ecology. FEMS Microbiology Ecology, 62(2), 142-160.
 - Remenant, B., Grundmann, G. L., & Jocteur-Monrozier, L. (2009). From the micro-scale to the habitat: Assessment of soil bacterial community structure as shown by soil structure directed sampling. *Soil Biology and Biochemistry*, 41(1), 29-36.
 - 29. Sahu, B. K., Pati, P., & Panigrahy, R. C. (2014). Environmental conditions of Chilika Lake during pre and post hydrological intervention: An overview. *Journal of Coastal Conservation*, *18*(3), 285-297.
 - 30. Satyanarayana, C. H. (1999). *Hydrographic and phytoplankton characteristics of Chilka Lake, a brackishwater lagoon, on the east coast of India* (Doctoral dissertation), Ph. D. thesis, Biology Department, Andhra University, Visakhapatnam, India.
 - Schloss, P. D., Westcott, S. L., Ryabin, T., Hall, J. R., Hartmann, M., Hollister, E. B. & Weber, C. F. (2009). Introducing mothur: Open-source, platform-independent, community-supported software for describing and comparing microbial communities. *Applied and Environmental Microbiology*, 75(23), 7537-7541.
- 32. Schramm, A., de Beer, D., van den Heuvel, J. C., Ottengraf, S., & Amann, R. (1999). Microscale distribution of populations and activities of nitrosospira and nitrospira spp. along a macroscale gradient in a nitrifying bioreactor: Quantification by in situ hybridization and the use of microsensors. *Applied and Environmental Microbiology*, 65(8), 3690-3696
- Tank, S. E., Lesack, L. F. W. & McQueen, D. J. (2009). Elevated pH regulates bacterial carbon cycling in lakes with high photosynthetic activity. Ecology, 90, 1910–1922.

- 34. Teisserenc, R., Lucotte, M., Houel, S., & Carreau, J. (2010). Integrated transfers of terrigenous organic matter to lakes at their watershed level: A combined biomarker and GIS analysis. *Geochimica et Cosmochimica Acta*,74(22), 6375-6386.
 - 35. Thureborn, P., Lundin, D., Plathan, J., Poole, A. M., Sjöberg, B. M., & Sjöling, S. (2013). a metagenomics transect into the deepest point of the Baltic Sea reveals clear stratification of microbial functional capacities. *PloS One*, 8(9), c 74983.
 - 36. Urban, N. R., Dinkel, C., & Wehrli, B. (1997). Solute transfer across the sediment surface of a eutrophic lake: I. Porewater profiles from dialysis samplers. *Aquatic Sciences*, 59(1), 1-25.
 - 37. Vos, M., Wolf, A. B., Jennings, S. J., & Kowalchuk, G. A. (2013). Micro-scale determinants of bacterial diversity in soil. *FEMS Microbiology Reviews*, *37*(6), 936-954.
 - Wang, Q., Garrity, G. M., Tiedje, J. M., & Cole, J. R. (2007). Naive Bayesian classifier for rapid assignment of rRNA sequences into the new bacterial taxonomy. *Applied and environmental microbiology*, 73(16), 5261-5267.
 - 39. Wang, Y., Sheng, H.F., He, Y., Wu, J.W., Tam, N.F.Y., & Zhou, H.W. (2012) Comparison of the levels of bacterial diversity in freshwater, intertidal wetland, and marine sediments by using millions of illumina tags. *Applied Environmental Microbiology*. 78, 8264–8271.
 - Webster, G., O'Sullivan, L. A., Meng, Y., Williams, A. S., Sass, A. M., Watkins, A. J. & Weightman, A. J. (2015). Archaeal community diversity and abundance changes along a natural salinity gradient in estuarine sediments. *FEMS Microbiology Ecology*, 91(2), 1-18.
 - 41. Whitman, W. B., Coleman, D. C., & Wiebe, W. J. (1998). Prokaryotes: The unseen majority. *Proceedings of the National Academy of Sciences*, 95(12), 6578-6583.
- 401 42. Zaitseva, S. V., Abidueva, E. Y., Namsaraev, B. B., Wang, L., & Wu, L. (2014). Microbial community of the bottom sediments of the brackish Lake Beloe (Transbaikal region). *Microbiology*, *83*(6), 861-868.

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- **Fig. 1.** Map of Chilika Lake showing the locations of 24 sampling stations spanning four different sectors, northern, central, southern, and outer channel.
- 416 Fig. 2. Interpolation of satellite images. Red polygons represent shrimp farming zones and fish landing areas.
- 417 Fig. 3. Interpolation of the salinity measured in monsoon (a), winter (b) and summer (c) seasons, color gradient from red 418 (highest concentration) to blue (lowest concentration). Interpolation of the Shannon Index (d). Color gradient from red 419 (highest index) to blue (lowest index).
- **Fig. 4.** Results of PCA of the microbial community at the genus level (72 samples) with overlays of the environmental variables and the different zones denoted in different colors. (note Env. = Environmental).
- **Fig. 5.** Results of PCA of the microbial community at the genus level (84 samples) with sample color referring to their salinity levels before and after the storm.
- **Fig. 6.** Results of PCA of the microbial community at the genus level (24 samples) with samples taken before (green) and after (blue) the storm.
- 426 Fig. 7. a) Effect of salinity on the diversity (Shannon Index); b) Effect of salinity on *Proteobacteria* relative abundance (coefficient of determination (\mathbb{R}^2) of 0.1896, p < 0.0002); c) Effect of the relative abundance of *Proteobacteria* and diversity (Shannon Index); and d) Effect of salinity on *Gammaproteobactria* relative abundance, \mathbb{R}^2 of 0.2829, p < 0,0001
- 429 Fig. 8. Multiple scale heterogeneity of the spatial distribution of variables driving the composition of bacterial communities.
- **Fig. 9.** Satellite view incorporating GIS stream analysis, fish farms (red polygons) and sampling points with relatively high abundance of single genus or family (yellow dot) and corresponding reduction in diversity.





























Gammaproteobacteria relative abundance



Larger scale heterogeneity (climate, geological substrate...) integrated in all the sample sites

Unmeasurable smaller scale heterogeneity (nutrients, organic compounds...)

Field (lake) scale distribution spatial pattern of physicochemical parameter : Salinity

