



Differential analysis of prokaryotic communities from pristine mangrove tidal zone sediments reveal distinct structures and functional profiles

Carolina Oliveira de Santana¹; Pieter Spealman²; Vânia Maria Maciel Melo³; David Gresham²; Taíse Bomfim de Jesus⁴,

Fabio Alexandre Chinalia¹

¹Programa de Pós-Graduação em Geoquímica: Petróleo e Meio Ambiente. Instituto de Geociências (IGEO), Universidade Federal da Bahia (UFBA), R. Barão de Jeremoabo, s/n - Ondina, Salvador, BA 40170-290, Brazil

²Center for Genomics and Systems Biology, Department of Biology, New York University, New York, New York, United States of America.

³Laboratório de Ecologia Microbiana e Biotecnologia (LEMBiotech), Departamento de Biologia, Centro de Ciências, Universidade Federal do Ceará, Campus do Pici, Bloco 909, Avenida Mister Hull s/n, 60.455-970 Fortaleza, CE, Brazil

⁴Programa de Pós-Graduação em Modelagem em Ciências da Terra e do Ambiente (PPGM), Universidade Estadual de Feira de Santana (UEFS), Avenida Transnordestina, s/n - Novo Horizonte, Feira de Santana, BA 44036-900, Brazil

Correspondence to: David Gresham (dgresham@nyu.edu)

Abstract

Mangrove forests are intertidal ecosystems that constitute a large portion of the world's coastline, as such, they are composed of, and reliant upon, microhabitats defined by the tides. However, we are only beginning to understand tidal



21 microhabitat biodiversity and their role in nutrient cycling. The majority of metagenomic studies have so far been conducted
22 on anthropogenically impacted areas. As even mild disruption can severely alter ecosystems and lead to decreased biodiversity
23 and local extinctions, this is a critical issue . Here, we characterize prokaryotic populations and their involvement in nutrient
24 cycling across the tidal zones of a pristine mangrove forest within a Brazilian Environmental Protection Area of the Atlantic
25 Forest. We hypothesize that tidal zones in pristine mangroves constitute distinct microhabitats, are composed of different
26 prokaryotic communities and, consequently, distinct functional profiles. Samples were collected in triplicate from zones below,
27 between, and above the tidal waterline. Using 16S rRNA amplicon sequencing, we find significantly different prokaryotic
28 communities with diverse nutrient cycling related functions, as well specific taxa with varying contribution to functional
29 abundances between zones. Our findings contrast those observed in anthropogenically impacted mangroves and suggest that
30 some aspects of mangrove zonation may be compromised by human activity.

31 **Keywords:** functional prokaryotic ecology; mangrove; metagenomics; tidal zones; prokaryote microbiome; pristine
32 mangrove forest

33 1. Introduction

34 Soils are among the greatest sources of microbial diversity on the planet (Tveit *et al.* 2013, Kaur *et al.* 2015, Nesme
35 *et al.* 2016). These microorganisms are fundamental to many processes such as carbon and nitrogen cycling, as they shape
36 and define important characteristics of their habitats through metabolic activities (Wendt-Potthoff *et al.* 2012; De Mandal,
37 Chatterjee and Kumar 2017; Kumar and Sai 2015). Mangrove ecosystems constitute a large portion of the tropical and
38 subtropical coastlines of Earth (Yunus *et al.* 2011; dos Santos *et al.* 2011). Beyond their value as natural barriers that reduce
39 erosion and the impact of storms, they are economically valuable for medicinal, energetic, and eco-tourist uses (Purahong *et*
40 *al.* 2019), as well as being critical ecosystems in climate change mitigation (Howard *et al.* 2017; Carugati *et al.* 2018). Many
41 studies have assessed the association between microbial communities from soils and plant development (Panke-Buisse, Lee



42 and Kao-Kniffin 2017; Wolińska *et al.* 2017; Wagner *et al.* 2014, Zarraonaindia *et al.* 2015, Capdeville *et al.* 2018) with
43 multiple lines of evidence supporting a plant-soil feedback loop of microbiomes affecting plant diversity while also being
44 shaped themselves (Van Der Heijden, Bardgett and Van Straalen 2008; Mariotte *et al.* 2018; Bennett and Klironomos 2019;
45 Miller, Perron and Collins 2019). Thus, considering the dependency of the mangrove forests on the sediment microbiome, it
46 is important to understand the microbial activities in these sediments in greater detail (Yunus *et al.* 2011; Lin *et al.* 2019).

47 However, a mangrove forest does not have only a single type of sediment, as tidal ecosystems, they are characterized
48 by periodic tidal flooding. This dynamic leads to varying environmental conditions across small spatiotemporal scales, with
49 levels of nutrients, oxygen and salinity periodically fluctuating, resulting in frequent anaerobic conditions and a wide range of
50 redox potentials (Andreote *et al.* 2012; Lin *et al.* 2019). Dynamic conditions like these can lead to high microbial diversity,
51 and these microbes play essential roles in the functioning and maintenance of the greater ecosystem (Andreote *et al.* 2012;
52 Imchen *et al.* 2017; Lin *et al.* 2019; Huergo *et al.* 2018). Although previous research has sought to characterize the prokaryotic
53 microbiomes across mangrove tidal zones, these works were conducted in anthropogenically impacted areas (Rocha *et al.*
54 2016; Zhang *et al.* 2018), which, given the sensitivity of the mangrove microbiome, can confound the interpretation of the
55 community structure (Pupin and Nahas 2014; Alongi 2008; Carugati *et al.* 2018; Nogueira *et al.* 2015).

56 The Atlantic Forest in Brazil is one of the most biodiverse ecosystems on the planet, containing numerous varieties
57 of dry and wet broadleaf forests, savannas and mangrove forests, the later of which are primarily composed of genera
58 *Rhizophora*, *Avicennia*, *Laguncularia* and *Conocarpus* (Pupin and Nahas 2014). This biome is threatened by anthropogenic
59 disturbances such as logging and farming, as well as habitat loss and fragmentation due to human encroachment, resulting in
60 a severe decline in its original area (Ditt *et al.* 2013; Ministério do Desenvolvimento Agrário 2010; Pupin and Nahas 2014;
61 Ghizelini, Mendonça-Hagler and Macrae 2012; Nogueira *et al.* 2015). However, in the southern part of Bahia State, Brazil, a



62 significant fragment of the Atlantic Forest remains preserved within the Environmental Protection Area (APA) of Pratigi
63 (Ministério do Meio Ambiente 2004). Recent studies on the environmental conditions of the area show that preservation efforts
64 initiated in 1998 have been generally effective, resulting in high environmental quality relative to most mangroves, both in
65 Brazil and globally (Ditt *et al.* 2013; Lopes 2011; Mascarenhas *et al.* 2019). This preserved area constitutes an important site
66 for the understanding of the ecology of unimpacted mangrove forests.

67 Therefore, in order to improve our understanding of mangrove ecology, in this study we characterize the prokaryotic
68 microbiota present in pristine mangrove sediments of the Serinhaém estuary within the Pratigi APA using 16S rRNA amplicon
69 metagenomics. This approach allows us to identify diverse taxa without the laborious task of culturing them (Kaur *et al.* 2015;
70 Mocali and Benedetti 2010; Bornemann *et al.* 2015; Nesme *et al.* 2016). Furthermore, we assess the community structure and
71 functional aspects of these prokaryotes to achieve a deeper understanding of the terrestrial processes at work in different
72 environments (Mahmoudi *et al.* 2015). Although mangroves have previously been shown to have prokaryotic populations
73 distinct from the regions they border (ie. mountain forest and restinga), (Mendes and Tsai 2018), only recently has there been
74 work to understand the differences between mangrove microhabitats (Rocha *et al.* 2016; Zhang *et al.* 2018). Considering
75 mangrove zonation as driven, primarily, by tide variation, we hypothesized that sediments of different mangrove regions would
76 differ significantly in richness and composition of prokaryotic communities, with the intertidal zone having the highest
77 diversity. We assessed the prokaryotic communities, the influence of environmental variables and the functional profile of
78 these sediments. We also identified the possible taxa driving the different nutrient cycles between zones. Our study provides
79 insight into the role of microbes in the functioning of mangrove forests and establishes a baseline for monitoring the health of
80 this important ecosystem.

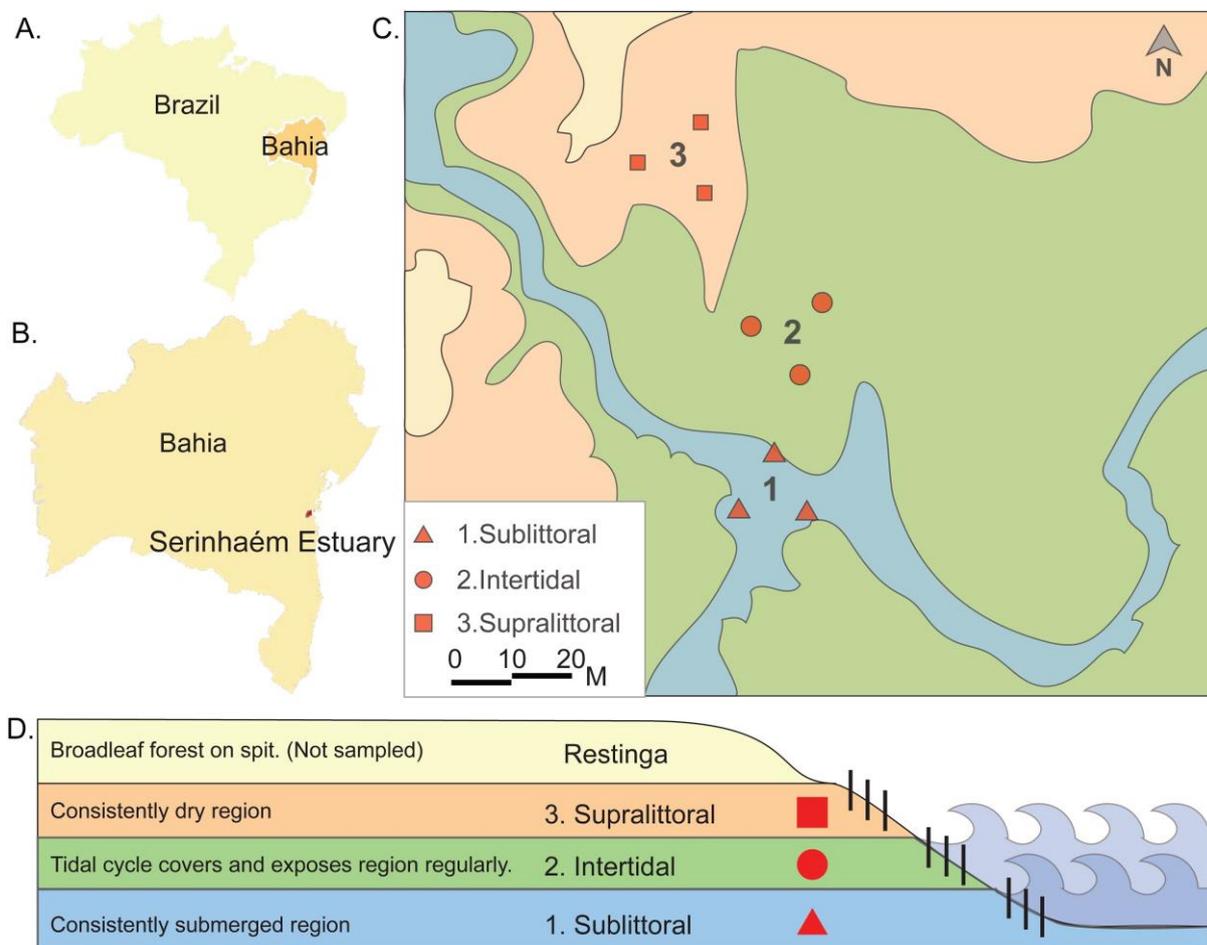


81 Importantly, this work was conducted before a massive oil spill occurred off the coastline of Brazil in August 2019,
82 impacting hundreds of miles of coastline including the Serinhaém estuary where this research was conducted. This work
83 therefore serves as a baseline measure of the prokaryotic communities of the tidal zones of what was a pristine mangrove
84 forest. We hope that this will spur subsequent research into the effects that anthropogenic effects have on mangrove
85 ecosystems.

86 2. Materials and Methods

87 2.1 Study area

88 The Serinhaém Estuary is located in the Low South Region of Bahia State, Brazil (Fig. 1), between the coordinates
89 13°35'S and 14°10'S and 39°40'W and 38°50'W. The estuary is within the Pratigi Environmental Protection Area (APA), one
90 of the few remaining Atlantic forest regions with a total area of 85 686 ha, enclosing a 32 km long portion of the lower Juliana
91 River and emptying directly into Camamu Bay along with several smaller rivers (Corrêa-Gomes *et al.* 2005).



92

93 **Fig 1. Map and schematic of sediment sampling sites.** Here we show the locations of the sampling sites relative to Brazil (A) and Bahia
94 (B). A satellite picture shows the relation of the three sampling sites within each zone (1. sublittoral, 2. intertidal, 3. supralittoral, (C)). A
95 schematic shows the topographic and tidal relation of each sampling site (D).

96 2.2 Sampling and DNA extraction

97 For clarity, we refer to the location of a sample as a ‘site’ and the collection of sites within a tidal zone as a ‘zone’.
98 Samples were collected from 3 tidal zones (centered around 13°42'59.0"S, 39°01'35.9"W) in the Serinhaém estuary in July
99 2018 during the morning low tide period. No sites exhibited signs of anthropogenic disturbance or pollution. The 3 collection
100 zones were chosen based on tidal influence; sublittoral, intertidal, and supralittoral regions (Fig. 1). From each tidal zone, 3



101 samples of superficial sediments (top 10 cm of the surface layer) were collected with a cylindrical sediment core sampler. To
102 ensure that our replicates sampled a broad representation of each zone, sample sites were located a minimum of 15m from
103 each other in a triangle. Plant and other organic material was manually removed from core samples, with precautions taken to
104 avoid the disruption of rhizospheres associated with vegetation.

105 Physical-chemical parameters such as temperature, salinity and dissolved oxygen in the water column were measured
106 using a multiparameter monitoring system (YSI model 85, Columbus). Each zone had different vegetation densities, with the
107 sublittoral zone having the greatest plant density, and the supralittoral the least, with almost no vegetation. Metal concentrations
108 were not collected as previous analysis performed by our lab (Jesus, T.B.) found no significant difference in metal
109 concentrations relative to background within the Serinhaém estuary. After collection, samples were transferred to the
110 laboratory. For each sediment core an aliquot was separated and kept in the -20°C freezer for subsequent DNA extraction
111 while the remainder of the sample was used for measuring organic matter content. The total genomic DNA was extracted from
112 0.25 g of sediment using the PowerSoil DNA Isolation Kit (Qiagen, Carlsbad, CA, USA). All DNA samples were stored at -
113 20° C before library preparation and sequencing .

114 **2.3 Library preparation and sequencing**

115 After DNA extraction, we used PCR to amplify the V4 region of the bacterial 16S rRNA using the primer pair 515F-
116 Y (Parada, Needham and Fuhrman 2016) and 806R-XT (Caporaso *et al.* 2011). PCR was performed with a thermal cycler
117 using the following program: 2.5 µl of each sample were added with 5 µl the forward and reverse primers and 12.5 µl of the
118 2x KAPA HiFi HotStart ReadyMix, making up for total 25µl and subjected to one cycle of 95°C for 3 minutes, 25 cycles of
119 95°C for 30 seconds, 55°C for 30 seconds and 72°C for 30 seconds and one cycle of 72°C for 5 minutes. The samples were
120 amplified in triplicates that were subsequently pooled back as one sample prior to sequencing. After amplification of the V4
121 region, Illumina sequencing adapters and dual-index barcodes were added to the amplicon target using the Nextera XT indices
122 Kit according to manufacturer's directions (Illumina, San Diego, CA, USA). The amplified DNA was then checked for size
123 using a Bioanalyzer. DNA sequencing was performed using Illumina MiSeq platform, V2 kit (300 cycles).



124 2.4 Data analysis

125 2.4.1 Sequence Trimming

126 Trimmomatic (Bolger, Lohse and Usadel 2014) was used to filter and trim demultiplexed sequences
127 (ILLUMINACLIP:TruSeq3-PE.fa:2:30:10 LEADING:3 TRAILING:3 SLIDINGWINDOW:4:15 MINLEN:100). A minimum
128 average read quality score of 15 was required for inclusion while the sliding window cuts any read at the point where the
129 median quality score over a 4 nucleotide window is less than 15.

130 2.4.2 Sequence denoising and OTU clustering using QIIME2

131 QIIME (Caporaso *et al.* 2010) was used to join forward and reverse reads into single reads (join_paired_ends.py, -j 4
132 -p 1). Reads were denoised using DADA2 (Callahan *et al.* 2016) (denoise-single, --p-trim-left 3, --p-trunc-len 0, --p-max-ee
133 2.0, --p-trunc-q 2) in QIIME2 (Bolyen *et al.* 2019), (q2cli, version 2019.4.0). Denoised sequences are clustered into Operational
134 Taxonomic Units (OTUs). Alpha-rarefaction was calculated using QIIME2 (alpha-rarefaction, max depth=17000), (S3 Fig.).
135 We performed a variety of alpha-diversity (S4 Fig., S5 Fig., S6 Fig., S7 Fig.) and beta-diversity (Fig. 3, S8 Fig.) tests using
136 QIIME2 (core-metrics-phylogenetic, p-sampling depth 9340).

137 2.4.3 Taxonomic assignment, community visualization and environmental tests

138 Taxonomic assignment used Vsearch (Rognes *et al.* 2016) in QIIME2 using Open Reference with 97% similarity (--
139 p-perc-identity 0.97) against the reference 16S rRNA sequences in SILVA database (Silva SSU 132), (McDonald *et al.* 2012).
140 QIIME2 visualizations for 1. OTU abundance (S1 Fig., S2 File), 2. proportional representation between sites, are available as
141 a supplementary files (S2 File), and 3. taxonomy (S3 File).

142 Phylogenetic reconstruction was carried out in QIIME2 using the representative sequences for each OTU and a
143 QIIME2 feature classifier trained using the 97% similarity representative set containing only 16S rRNA sequences (e.g.
144 silva_132_97_16S.fna). All groups were required to be present within at least 2 samples with a minimum of 3 reads each.



145 QIIME2 tree files were accessed in R using QIIME2R (version 0.99.12). Tree visualization (Fig. 7) was performed
146 with R (version 3.4.4) using Metacoder (Foster, Sharpton and Grünwald) (version 0.3.2). Posterior analysis was performed
147 using Phyloseq (McMurdie and Holmes 2013), (version 1.22.3). Analyses in R were plotted using ggplot2 (McMurdie and
148 Holmes 2013; Villanueva and Chen 2019).

149 Vegan (Dixon 2003), (version 2.5-6) was used to test correlations between community structure and environmental
150 variables. Distances were calculated using metaMDS, (engine=monoMDS, try=1000, k=3), and then fit the environmental
151 variables using envfit (default settings, permutations=333), (S4 Table).

152 2.4.4 Functional analysis using PICRUS2

153 Functional analysis was performed using PICRUS2 (version 2.3.0-b) ([Douglas *et al.* 2019](#); [Barbera *et al.* 2019](#);
154 [Czech, Barbera and Stamatakis 2020](#); [Louca and Doebeli 2018](#); [Ye and Doak 2009](#)) with default settings. Both the Kegg
155 Orthologs (KOs) and MetaCyc pathways were analyzed for significant (p-value ≤ 0.05) differential abundances after centered
156 log-ratio transformation (aldex.clr) using the general-linear model method (aldex.kw) of the ALDEx2 package (ver 1.18.0). A
157 heatmap of KOs with differential abundance between sample sites was then generated (Fig. 6).

158 2.4.5 Site-specific taxonomic and functional enrichment

159 In order to identify which species were significantly different in abundance in each zone we performed taxa
160 enrichment analysis (Fig. 5), (Spealman *et al.* 2020). First, OTU abundances were normalized by downsampling to match the
161 least abundant zone (Intertidal). Taxa abundances are the sum of all assigned OTU abundances. For each taxa, we required
162 that a significant difference be found between sites using a Chi-squared, 2x3 test, with correction
163 (scipy.stats.chi2_contingency) using the mean normalized abundance. To correct for false positives due to variance between
164 replicates we required the distributions of unnormalized OTU abundances between sites to also be significantly different
165 (Mann-Whitney U test, scipy.stats.mannwhitneyu, p-val ≤ 0.05). Finally, to ensure biological relevance, we required the
166 effect size to represent at least 5% difference in log-fold abundance between sites.



167 To determine which taxa were associated with differences in functional abundance, we also calculated KO enrichment
168 specific to each taxa at a given taxonomic level (Fig. 5, 7) (Spealman *et al.* 2020) using the functional abundance results of
169 PICRUST2. We required that a given taxa must have at least 10% of all KO functional abundance at the given level; that the
170 functional abundance be significantly enriched using a Binomial exact test (Bonferroni corrected p-value ≤ 0.05), and the
171 taxa must have at least three distinct KOs within a single pathway that meet these criteria. All KOs and their metabolic
172 pathways are available in a supplemental file (S6 File).

173 2.4.6 Accessibility

174 The entire computational workflow is available on Github: https://github.com/pspealman/COSantana_2020.

175 Data used in the performance of the analysis and archival versions of the computational workflow are available on Dryad:
176 <https://doi.org/10.5061/dryad.gf1vhhmkz> (Spealman *et al.* 2020). [Temporary link for reviewers:
177 <https://datadryad.org/stash/share/bwmAgXaOhXT2JNHKbfX15wpIJ3dAxAhOXrjdwNwSSHM>]

178 The data has been deposited as PRJNA608697 in the NCBI BioProject database:
179 <https://www.ncbi.nlm.nih.gov/bioproject/PRJNA608697>

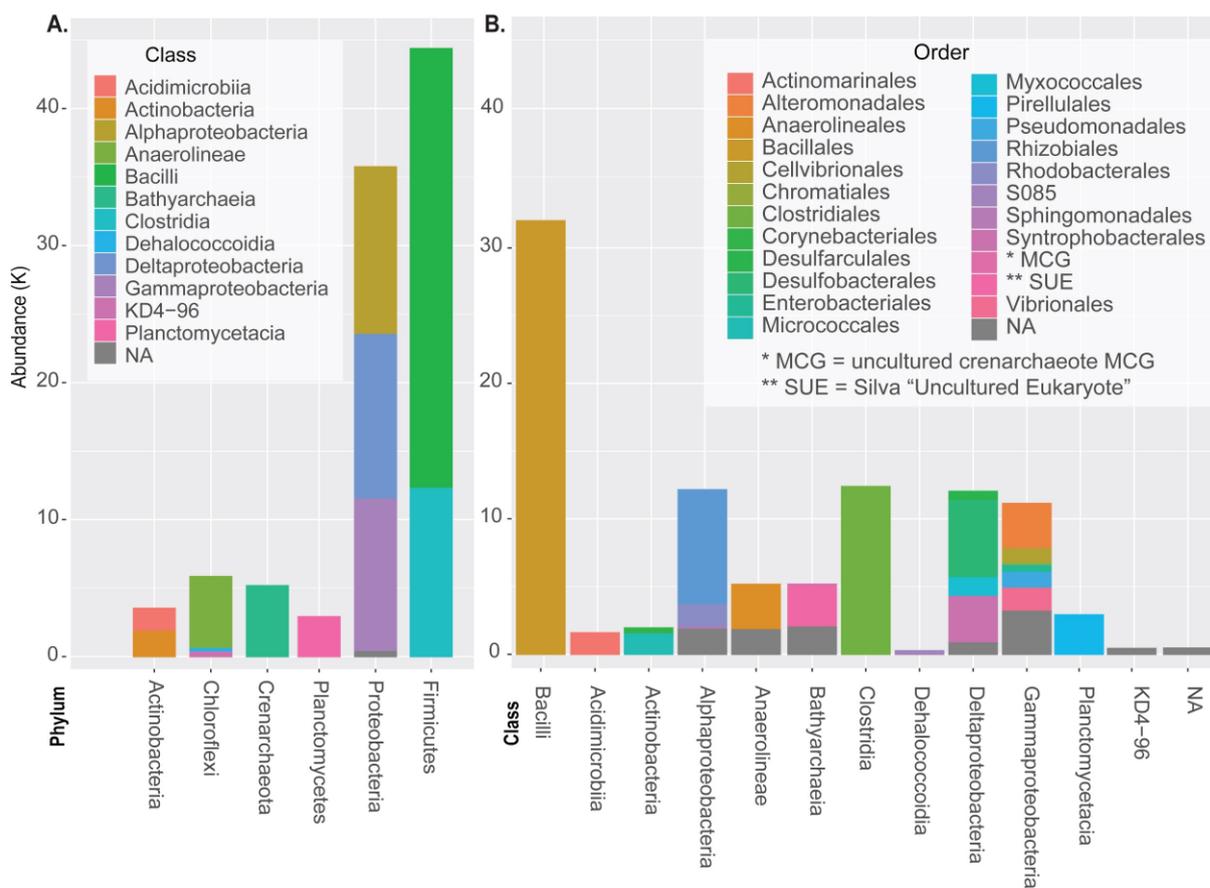
180 3. Results

181 3.1 Taxonomic composition of prokaryotic communities

182 After quality filtering, a total of 204 599 bacterial and archaeal sequences remained for community analysis,
183 corresponding to an average of 22 733.2 sequences per sample. Sequence clustering yielded a total of 1709 OTUs. Of these,
184 1,623 OTUs and 193 143 sequences were assigned to Bacteria (94.4%) and 84 OTUs and 10 707 sequences were assigned to
185 Archaea (5.2%) kingdoms (S1 Fig.). 749 sequences clustered in 2 OTUs (0.4%) that could not be assigned to any prokaryotic
186 kingdom. Additionally, one mis-annotated Archea taxa originally named “uncultured eukaryote” has been manually changed



187 to “SUE” for SILVA uncultured eukaryote. All sites combined, we identified 37 unique phyla, 142 classes, 165 families, 142
 188 genera and 97 species. From the total, 18 087 sequences (approximately 9%) weren’t assigned to the phylum level. More than
 189 88% of all the sequences that could be assigned to the phylum level belonged to 6 phyla: *Proteobacteria* (30.3% abundance,
 190 62 135 sequences), *Firmicutes* (29.4% abundance, 60 307 sequences), *Chloroflexi* (6.4% abundance, 13 225 sequences),
 191 *Planctomycetes* (5.3% abundance, 10 888 sequences), *Actinobacteria* (4.6% abundance, 9390 sequences) and *Crenarchaeota*
 192 (3.8% abundance, 7 921 sequences). The total sequence and OTU abundances in all the observed phyla are summarized in S1
 193 Table. Fig. 2 shows all the classes and orders of the 6 dominant phyla in the data set. Families and genus are shown in
 194 Supplemental (S2 Fig.).



195



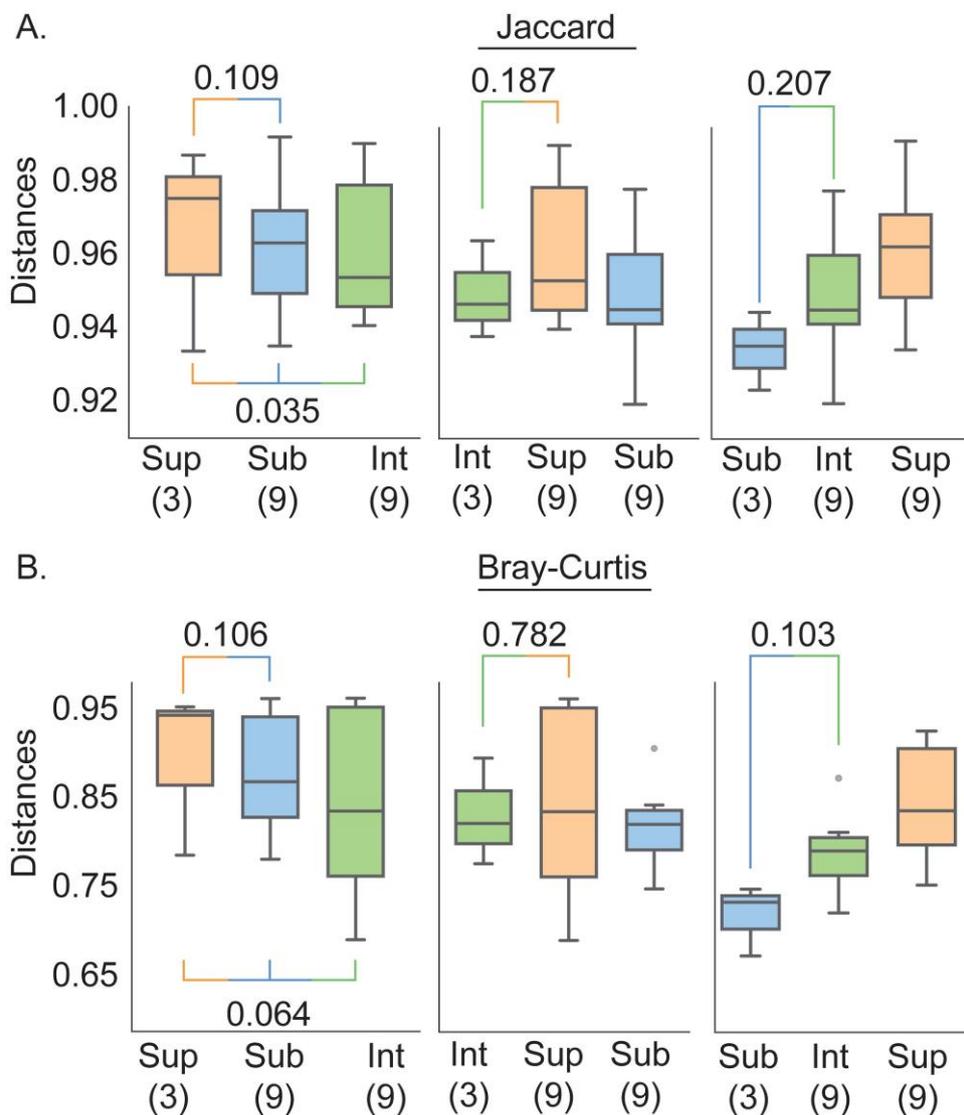
196 **Fig 2. Taxonomic abundances from all sample sites.** Taxa identified within the samples are shown as stacked bar plots, the horizontal axis
197 is the higher taxonomic level while the stacked bars are the lower level. Phylum is the horizontal axis with Class being the stack (**A**), Class
198 is the horizontal axis with Order being the stack (**B**).

199

200 **3.2 Microbial diversity of mangrove tidal zones**

201 The alpha diversity indices for each zone were calculated using QIIME2 (S4 Fig.). Overall, the sediments from the
202 sublittoral zone had higher richness and diversity indices, while the intertidal zone exhibited the lowest alpha diversity indices
203 (S5, S6, S7 Fig.).

204 We used the beta-diversity package of QIIME2 to assess differences in prokaryotic populations between zones (Fig.
205 3), (S8 Fig.) and significant differences using the distance metrics in S2 Table. Using the Bray-Curtis and Jaccard distance
206 metrics (Fig. 3) we found significant differences in population structures between zones (p -value < 0.05 in Jaccard, p -value $<$
207 0.1 in Bray-Curtis, $\alpha = 0.1$, 90% confidence). The pseudo-F test results for pairwise PERMANOVA failed to indicate statistical
208 differences site to site, despite the apparent dissimilarity between the groups in the plots, potentially because this test would
209 require a larger number of samples (see Supplemental Permanova Section).



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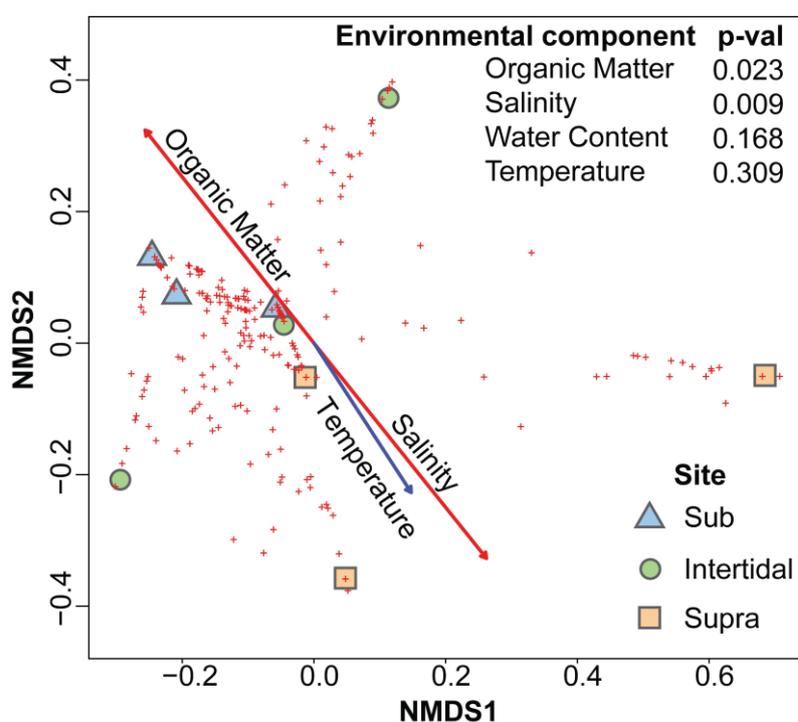
211 **Fig 3. Beta-diversity between sampling sites.** All sites are compared pairwise using distance metrics. The differences between sites are
212 significant for both the Bray-Curtis and Jaccard distance metrics ($\alpha=0.1$).

213 3.3 The influence of environmental variables

214 To determine if differences in population structures between zones correlated with abiotic environmental variables
215 we measured salinity, water content, organic matter and temperature from each tidal zone (S3 Table) and associated the



216 taxonomic abundances from each sample from each zone using Vegan (Dixon 2003) (see Methods). The results revealed a
217 significant correlation between the prokaryotic populations within each zone and salinity and organic matter (Fig. 4, S4 Table).
218 Specifically, increased organic matter was positively correlated with sublittoral population abundances, while increased
219 salinity was positively correlated with supralittoral population abundances. Neither water content or temperature measures
220 reflect a significant difference in community structure between zones.



221
222 **Fig. 4. Correlation between environmental variables and prokaryotic communities.** Dots represent taxonomic abundances per site as
223 plotted by Vegan metaMDS using nonmetric multidimensional scaling (NMDS). Arrow length is a representative of the predictor strength
224 of environmental variable vectors, with red arrows having statistical significance as calculated by envfit (p-value <= 0.05).
225

226 3.4 Taxa enrichment by tidal zones of the mangrove

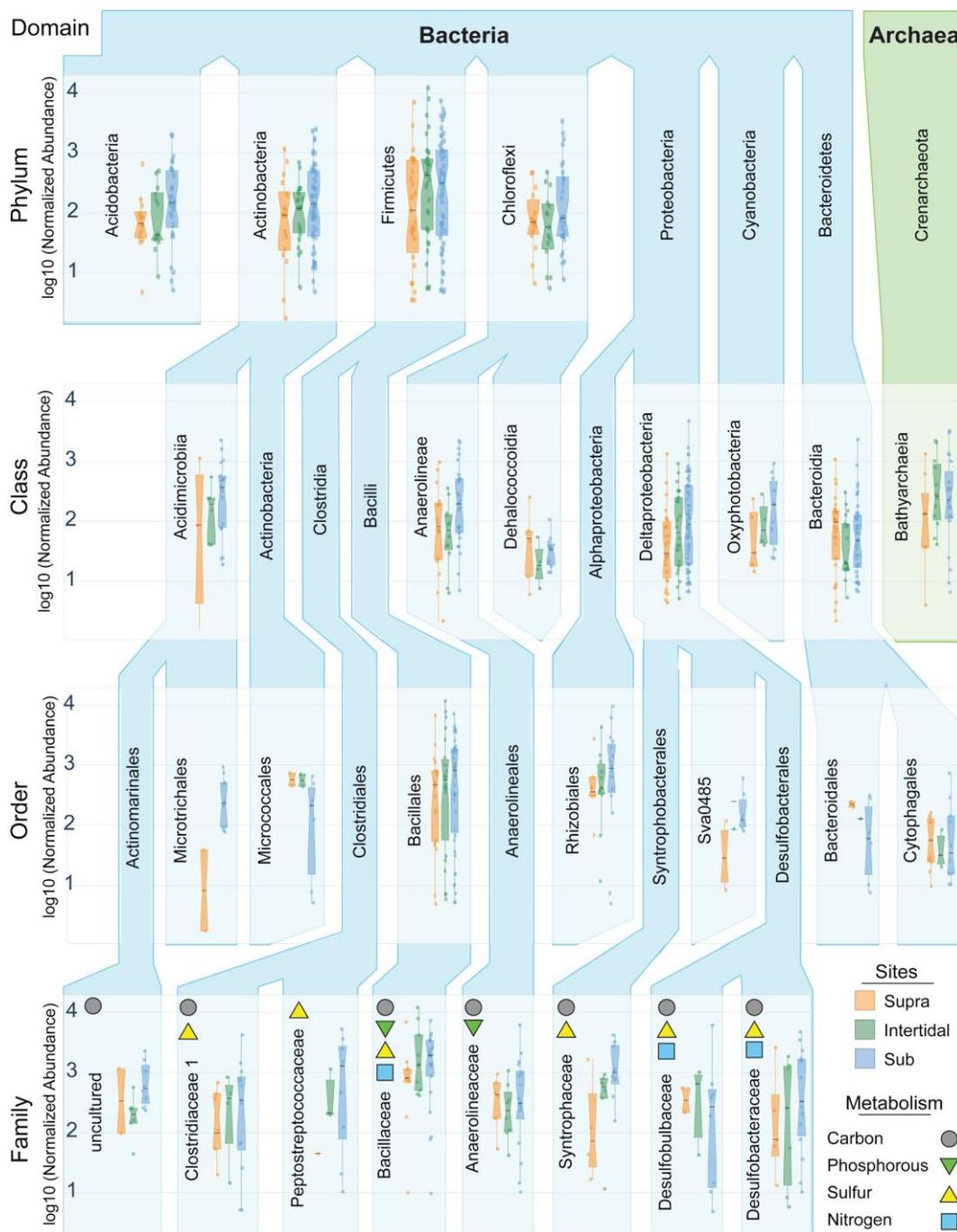
227 To determine what specific taxa were significantly different in abundance between zones, we performed a zone-
228 specific enrichment test for each taxa, (Fig. 5). Briefly, this method tests the normalized abundances of each taxa to identify



229 taxa with statistically significant and large effect size differences in functional abundances between zones (see Methods
230 section). Nearly every taxa (96%, 25/26) identified by this method has its greatest abundance in the sublittoral zone, with 38%
231 (10/26) having an inverse relationship between elevation and abundance, such that the abundance increases from the
232 supralittoral to intertidal to sublittoral zone.

233 We used PICRUST2 (Douglas *et al.* 2019) to correlate identified taxa with Kegg Orthologs (KOs) and then calculate
234 the functional abundance of KOs, (S4 File), (see Methods). Families with enrichment of metabolism associated KOs,
235 (accounting for more than 10% of the total of a given KO) for at least 3 KOs of a single metabolic pathway were then labelled
236 with an icon for that pathway (Fig 5.). (see Methods).

237 We found that 7 of the 8 enriched families make substantial (>10%) contributions to carbon metabolism associated
238 KOs (specifically methane metabolism), with the exception of *Anaerolineaceae*. Similarly, 6 families contribute for numerous
239 sulfur metabolism associated KOs such as assimilatory sulfate reduction contributed by *Bacillaceae*, and dissimilatory sulfate
240 metabolism associated KOs contributed by *Syntrophaceae*, *Desulfobulbaceae*, and *Desulfobacteraceae*. The families
241 *Bacillaceae*, *Desulfobulbaceae*, and *Desulfobacteraceae*, produce substantial amounts of nitrogen metabolism associated KOs,
242 with *Bacillaceae* contributing to dissimilatory nitrate reduction to ammonium (DNRA) and *Desulfobulbaceae* contributing to
243 nitrogen fixation associated KOs. Only *Anaerolineaceae* and *Bacillaceae* contribute to phosphorus metabolism associated
244 KOs. Taken together, this suggests that zone-specific taxa enrichment may also contribute to differential metabolic activities
245 at these zones. We describe the potential metabolic roles of each of these and other families below.



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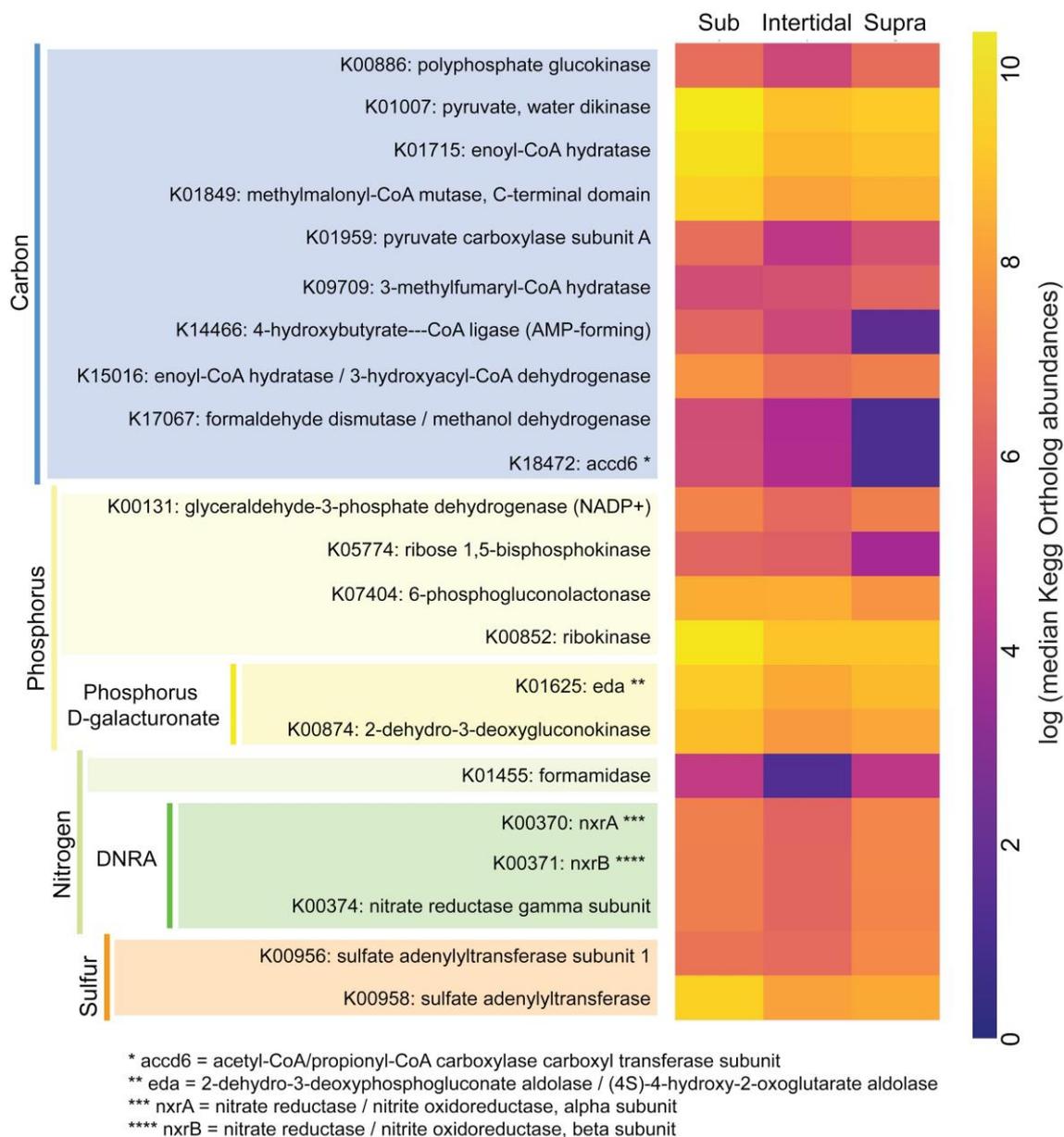
249

Fig. 5. Site specific measures of taxa find significant enrichment in sediments of different tidal zones. Here we show taxa with significantly different abundances between zones (Chi², p-val <= 0.05), statistically different distributions of abundance (MWU, p-val <=0.05), and whose mean effect size exceeded 10% (see Methods).



250 **3.5 Site specific differences in metabolism associated Kegg** 251 **Orthologs**

252 To determine if there were significant differences in metabolic activity between tidal zones we calculated the
253 functional abundance of metabolic KOs for each zone (see Methods). We found 22 metabolism associated KOs with
254 significantly different functional abundance between zones (Fig. 6). 8 KOs show higher abundances in the sublittoral
255 sediments, above both intertidal and supralittoral, including both members of the phosphorus D-galacturonate degradation
256 pathway. Conversely, only one KO is enriched in the supralittoral sample above both sublittoral and intertidal zones, and no
257 KO is enriched in the intertidal zone. 5 KOs exhibit an abundance in negative proportion to elevation, such that sublittoral is
258 greater than intertidal, which is, in turn, greater than supralittoral. Only one KO is observed to have the opposite trend, with
259 highest abundance in the supralittoral sites and lowest in sublittoral. Finally, 7 KOs show a bimodal distribution with near
260 equal abundances between sublittoral and supralittoral, with intertidal being the significantly less, including all four of the
261 nitrogen metabolism associated KOs. Taken together, KO enrichment reinforces the previously observed trend of reduced
262 abundance in the Intertidal site, and greatest abundance at the Sublittoral zone.



263

264 **Fig. 6. Zone specific measures of metabolism associated Kegg Orthologs.** The heatmap shows the 22 metabolism associated KOs that
 265 have significant functional abundance differences between zones for Carbon, Phosphorus, Sulfur and Nitrogen pathways.



266 3.6 Nutrient cycling pathways and taxa identified within 267 communities

268 In order to identify potential functional roles of the prokaryotic communities, we used PICRUSt2, which associates
269 KEGG orthologs (KOs) with 16S rRNA amplicons through gene family associations to reference genomes (Douglas *et al.*
270 2019), (S4 File shows KO assignment and abundance for each taxa). Here, we show the functional profiles for carbon, nitrogen,
271 phosphorus, and sulfur metabolic pathways. Where possible we rely on previous work that has identified microbial groups
272 with specific metabolic activity as potential drivers for processes such as methanogenesis, nitrogen fixation and nitrification
273 (Levipan *et al.* 2016; Fierer 2017). This is complemented with measured functional abundances observed for KOs associated
274 with the carbon, nitrogen, phosphorus, and sulfur metabolic pathways across the three mangrove intertidal zones. We use this
275 information to link taxa found in this study to their most probable and relevant nutrient cycling activities (Fig. 7, a full version
276 with all taxa is available as a supplemental file S5 File).

277 3.6.1 Carbon cycling

278 Carbon cycle pathways such as methane oxidation and methanogenesis showed enrichment in the sublittoral zone.
279 We find that *Syntrophaceae* contributes significantly with different methanogenesis associated KOs including pyruvate
280 ferredoxin oxidoreductase subunits alpha, beta, delta, and gamma (15%, 24%, 43%, and 14%, respectively) and heterodisulfide
281 reductase subunits A2, B2, and C2 (30%, 27% and 36%, respectively), (S4 File). We also find plentiful *Archaeal* families
282 contributing the majority (>50%) of methanogenesis KOs; *Nitrosopumilaceae*, uncultured families of *Lokiarchaeia*, and
283 *Bathyarchaeia* (see S4 File).

284 3.6.2 Sulfur cycling

285 Significantly higher abundances of sulfur transformation KOs were found in the sublittoral zone. The family
286 *Rhodobacteraceae* (Delmont *et al.* 2015) contributes substantial abundances (>10%) of 12 different sulfur metabolism KOs



287 (S4 File). The families *Syntrophaceae*, *Desulfobacteraceae*, *Desulfobulbaceae* contribute to almost 90% of the KOs associated
288 with dissimilatory sulfate reduction, which is in accordance with previous observations (Kuever 2014, Wörner and Pester
289 2019, Wiegel, Tanner and Rainey 2006; Oren and Xu 2014; Meyer *et al.* 2016).

290 Members of the order *Rhizobiales* are observed to be major drivers of the sulfur oxidizing process, as they contribute
291 85% of the sulfur-oxidizing protein SoxY and 65% of SoxZ. The family *Chromatiaceae*, which makes up the majority of
292 purple sulfur bacteria, are known for their role in the sulfur cycle in numerous environments (Wiegel, Tanner and Rainey 2006;
293 Oren and Xu 2014; Meyer *et al.* 2016; Hanada and Pierson 2006; Xia *et al.* 2019) and contribute substantial amounts of sulfur-
294 oxidizing protein SoxA (18%), SoxB (24%), SoxX (18%), and SoxZ (17%).

295 As a side note, *Desulfatiglans*, of the family *Desulfarculaceae* are also reported to act in degrading aromatic
296 hydrocarbons (Sun *et al.* 2010; Jochum *et al.* 2018). We find that they contribute 26% of all the 3-oxoadipate enol-lactonase,
297 and nearly 99% of all benzoyl-CoA reductase subunit BamB, BamC, and various benzoate degradation associated enzymes.

298 3.6.3 Phosphorus cycling

299 Similar to what is observed for the other nutrients, P cycling KOs are, overall, more abundant in the sublittoral
300 sediments. The genus *Pseudomonas* that belongs to the family *Pseudomonadaceae*, contributes a substantial amount (>40%)
301 of three phosphorus metabolism associated KOs and nearly 99% of three others. Along with *Pseudomonas*, the genus *Bacillus*
302 from the family *Bacillaceae*, has been suggested to be active in P cycling in mangrove sediments through phosphate
303 solubilization (Kalayu 2019; Malboobi *et al.* 2009). In our analysis we found that *Bacillus* contributed to phosphorus
304 metabolism associated KOs at the genus level, with 18 KOs greater than 20%, 5 of which are greater than 50%.

305 3.6.4 Nitrogen cycling



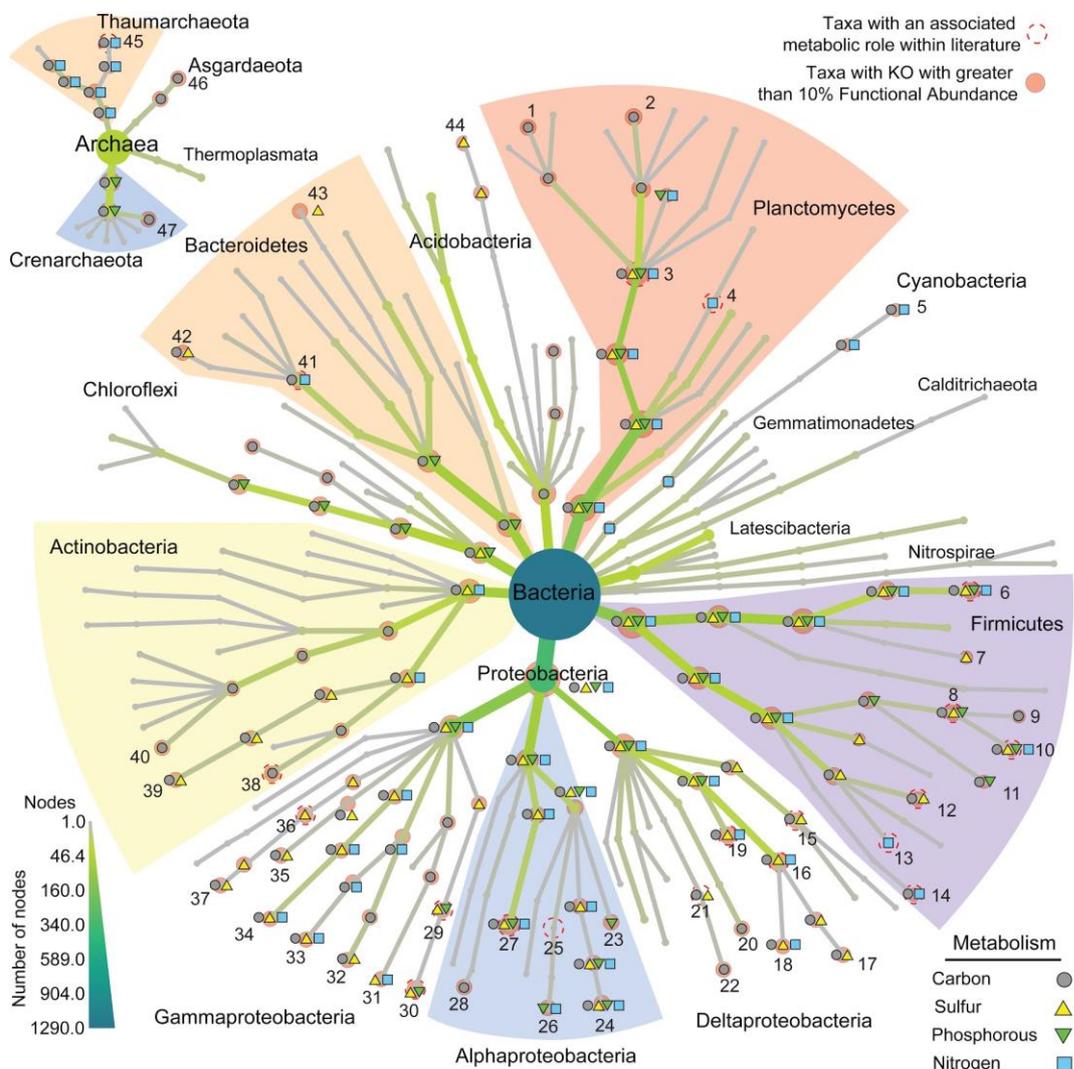
306 We observed significantly lower abundances for nitrogen transformation pathways for the intertidal zone. In the
307 literature, members of the family *Pirellulaceae* are relevant for ammonia oxidation processes (Jiang *et al.* 2015). In our
308 analysis, *Pirellulaceae* contributes only minorly to nitrogen metabolism with only 7 KOs having greater than a 10%
309 contribution. However, the ammonia-oxidizing archaea families represented by *Nitrososphaeraceae* (Kerou *et al.* 2016) and
310 *Nitrosopumilaceae*, make up nearly the entirety of the nitrification associated methane/ammonia monooxygenase KOs subunits
311 A, B, C (72%, 72%, 56% and 28%, 28%, 44%, respectively). Several components of the dissimilatory nitrate reduction to
312 ammonium (DNRA) pathway showed significantly different abundances between zones due to variations in the genus *Bacillus*.
313 *Desulfobulbaceae* members are also major contributors to nitrogen fixation, specifically nitrogenase iron protein associated
314 KOs.

315 Members of the *Clostridiaceae*, one of the most abundant families in the samples, are known for participating in
316 nitrogen-fixing, as well as other nitrogen transformations (Wiegel, Tanner and Rainey 2006; Oren and Xu 2014; Meyer *et al.*
317 2016; Chen, Toth and Kasap 2001). We found the genus *Clostridium* sp. AN-AS6E to significantly contribute to nitrogen
318 metabolism KOs. The genus *Vibrio* contributes greater than 20% to 3 nitrogen metabolism associated KOs, the genus
319 *Marinobacter* with 5 KOs, and the genus *Bacillus* with 10 KOs.

320 Other families that have known capacity for nitrogen fixing in the sediments are *Flavobacteriaceae* (Kämpfer *et al.*
321 2015), represented by 7 genera, *Pseudomonadaceae* (Özen and Ussery 2012), *Spirochaetaceae* (Lilburn *et al.* 2001), and
322 *Rhizobiaceae* (Broughton 2003), although only *Flavobacteriaceae* makes significant contributions to nitrogen metabolism
323 KOs. Some members of *Chromatiaceae* are known to be active in the nitrification process, as the genus *Nitrosococcus*
324 (Campbell *et al.* 2011) as well as the family *Pirellulaceae* (Kellogg, Goldsmith and Gray 2017), although we do not find these
325 making significant contributions in terms of KOs. Finally, organisms capable of performing ammonification are represented



326 by *Micrococcaceae* (Dastager *et al.* 2014) and *Rhodobacteraceae* (Delmont *et al.* 2015), although only *Rhodobacteraceae* has
327 significant contributions; with 4 nitrogen metabolism associated KOs.



- | | | |
|--|--|--|
| 1. Uncultured Pasteuria sp. (Pirellulaceae) | 19. Desulfobulbaceae | 36. Chromatiaceae |
| 2. Uncultured marine microorganism (Pirellulaceae) | 20. Uncultured (Myxococcales) | 37. Uncultured gamma proteobacterium (Thiohalocapsa) |
| 3. Pirellulaceae | 21. Desulfarculaceae | 38. Micrococcaceae |
| 4. Rubinisphaeraceae | 22. Desulfatigians | 39. Mycobacterium |
| 5. Synechococcus sp. CENA143 | 23. Xanthobacteraceae | 40. Uncultured lamia sp. (Actinomarinales) |
| 6. Bacillus | 24. Uncultured bacterium (Stappiaceae) | 41. Flavobacteriaceae |
| 7. Aneurinibacillus | 25. Rhizobiaceae | 42. Robertkochia marina |
| 8. Clostridiisalibacter | 26. Mesorhizobium sp. KYW12 | 43. Uncultured Bacteroidetes bacterium (Prolixibacter) |
| 9. Bacterium YC-ZSS-LKJ129 | 27. Rhodobacteraceae | 44. Uncultured actinobacterium (Bryobacter) |
| 10. Clostridium sp. AN-AS6E | 28. Erythrobacter jejuensis | 45. Nitrososphaeraceae |
| 11. uncultured Firmicutes bacterium (Brassicibacter) | 29. Pseudomonadaceae | 46. uncultured Desulfurococcus sp. (Lokiarchaeia) |
| 12. Clostridium sensu stricto 13 | 30. Pseudomonas | 47. uncultured SUE (Bathyarchaeia) |
| 13. Clostridium sensu stricto 7 | 31. Microbulbifer | |
| 14. Clostridium sp. AN-AS6C | 32. Pseudoalteromonas | |
| 15. Syntrophaceae | 33. Marinobacter | |
| 16. Desulfobacteraceae | 34. Vibrio | |
| 17. Metagenome (Desulfobacteraceae) | 35. Enterobacteriaceae | |
| 18. Desulfatitales | | |



329 **Fig. 7. Phylogenetic tree showing additional metabolic data.** Phylogenetic tree depicting assigned node abundances (Nodes, represented
330 by color and thickness of branches) and whether a given taxa is associated with a metabolic pathway given either literature (dashed red line)
331 or significant enrichment of functional abundance of metabolism associated KOs (see Methods). Names are only shown for nodes with
332 associated literature citations or leaves with significant enrichment. A more complete tree displaying the names of all taxa is available as a
333 supplement (S5 File).

334 4. Discussion

335 Previous work has shown that mangrove forests have variation in community structure, often associated with different
336 biotic and abiotic factors, however, the majority of these have been conducted in anthropogenically impacted areas (Pupin and
337 Nahas 2014; Marcial Gomes *et al.* 2008; Alzubaigy *et al.* 2016; Rocha *et al.* 2016; Ceccon *et al.* 2019; El-Tarabily 2002;
338 Imchen *et al.* 2017; Lin *et al.* 2019; Zhang *et al.* 2018), confounding the makeup of the microbial populations, their abundance,
339 and determination of environmental influences on these population structures. Importantly, the majority of this work does not
340 consider or does not identify the mangroves under study as anthropogenically impacted, despite frequently being only a few
341 km from dense metropolitan and industrial centers (Imchen *et al.* 2017; Lin *et al.* 2019; Zhang *et al.* 2018; Ceccon *et al.* 2019).
342 Notably, studies that sought to identify differences induced by pollution and urbanization on mangroves did find large-scale
343 differences in prokaryotic populations in impacted areas compared to preserved mangrove areas (Pupin and Nahas 2014;
344 Nogueira *et al.* 2015). While pioneering, this research did not study the population differences of distinct microhabitats within
345 mangroves. Here, we extend the study of preserved mangrove areas to characterize prokaryotic populations within tidal zone
346 microhabitats.

347 Previous work that has focused on mangrove microbial diversity has found that composition of bacterial communities
348 in sediments correlates with a broad range of variables, such as; hydrodynamic regimes, granulometry, organic matter content
349 (Colares and Melo 2013), vegetation (Rocha *et al.* 2016) and pollutant distributions, all of which can generate niche variations
350 (Peixoto *et al.* 2011). However, ecosystems can also exhibit a robust community structure, such that even significant
351 differences in variables, such as pH, are mitigated, resulting in less variation between communities than expected (Huerger *et al.*
352 *et al.* 2018). Previous work by (Gong *et al.* 2019) found environmental conditions and historical events play an important role in



353 shaping the bacterial communities as well. In our study, we found both salinity and organic matter to be significantly correlated
354 with community populations in different tidal zones (Fig. 4). While mangrove degradation has long been known to be sensitive
355 to both of these environmental variables (Alongi 2015) these results diverge from others (Rocha *et al.* 2016; Zhang *et al.*
356 2017), making it difficult to infer general trends. While this work adds to our understanding of prokaryotic variation in
357 mangrove forests, it is important to note that more studies need to be performed in mangroves more diverse than the
358 anthropogenically impacted areas of South America and Asia, as the majority of them have been so far (Imchen *et al.* 2017;
359 Lin *et al.* 2019; Huergo *et al.* 2018; Zhang *et al.* 2018; Gong *et al.* 2019; Ghizelini, Mendonça-Hagler and Macrae 2012).

360 The results of the alpha-diversity tests showed a greater number of OTUs and a greater taxonomic diversity in the
361 sublittoral mangrove sediments, while the intertidal zone had the lowest richness and diversity. Differentiation in mangrove
362 sediment communities (as measured by beta-diversity) from zones with distinct biotic and/or abiotic characteristics has
363 previously been reported in the literature (Alzubaidy *et al.* 2016; Rocha *et al.* 2016; Peixoto *et al.* 2011; Jiang *et al.* 2013;
364 Ceccon *et al.* 2019).

365 While differences in environmental variables between sites may partially explain differences in prokaryotic
366 communities between tidal zones (Fig. 5) it is also possible that they are influenced by additional factors, such as fungal,
367 eukaryotic microbe, and plant rhizome contamination (Rocha *et al.* 2016; Zhang *et al.* 2017). The presence of microbes
368 typically associated with plant rhizosphere, has been observed in many previous studies (Alzubaidy *et al.* 2016; Rocha *et al.*
369 2016; Gomes *et al.* 2010; Zhang *et al.* 2018; Ceccon *et al.* 2019). In these studies, the rhizosphere sediments confer enrichment
370 of specific taxa, and have higher alpha-diversity relative to non-rhizosphere associated sediments. While we attempted to avoid
371 the inclusion of plant material in our collection of sediments, the presence of mangrove trees and other vegetation is an
372 unavoidable feature of the tidal zones. Similarly, the higher density of vegetation observed in the sublittoral area may, in part,
373 explain the higher diversity of the prokaryotic populations we identified there (Bennett and Klironomos 2019; Miller, Perron
374 and Collins 2019). Additionally, the microbiome of the mangrove can be heavily influenced by eukaryotic communities
375 (Alzubaidy *et al.* 2015), which would be invisible to our 16S rRNA amplicon sequencing method. We believe our



376 understanding of prokaryotic community structures is only one step in a larger process that should ultimately include rhizome,
377 fungal, and eukaryotic populations information.

378 Functional profiling is a proxy measurement of metabolic activity, and where possible we attempted to supplement
379 the taxonomic and functional profiles generated by our analysis with controlled metabolic studies from the literature. The
380 identification of a rich and divergent set of taxa associated with the diverse nutrient cycles in mangrove sediments was expected
381 due to the previous observations of the microbial diversity in these environments (Rocha *et al.* 2016; Ceccon *et al.* 2019;
382 Cabral *et al.* 2016; Mendes and Tsai 2014; Zhao and Bajic 2015) and was both confirmed and extended in this study as we
383 identified further distinctions between tidal zones. Notably, the majority of metabolism associated KOs (Fig. 5, 6) and
384 pathways (S9 Fig.) had higher abundance in the sublittoral mangrove sediments. The higher abundances of KOs found at the
385 sublittoral zone is likely due to the greater taxonomic diversity that was also observed for this region.

386 Our data suggests that the intertidal regions of mangrove forests have lower prokaryotic diversity than those in the
387 constant environments in the supra- and sublittoral regions. While this is consistent with some microbial models of
388 microhabitat diversity (Alongi 1988) it is in disagreement with more recent studies of microbial communities in tidal zones
389 that suggests higher diversity is maintained by the dynamic tidal environment (Lv *et al.* 2016; Ceccon *et al.* 2019; Imchen *et*
390 *al.* 2017; Lin *et al.* 2019; El-Tarabily 2002; Imchen *et al.* 2017; Lin *et al.* 2019; Zhang *et al.* 2018). One possible explanation
391 of this is that the specialization we observe at the constant environments (sub- and supralittoral) would be lost in mangroves
392 that are degraded, polluted, or otherwise anthropogenically impacted, as these zones would be rapidly colonized by
393 opportunistic species. We hypothesize, that in pristine mangroves, the constant environments offered by sub and supralittoral
394 microhabitats allow for the accretion of specialists, while the frequent and cyclic variation of the tides act as a selective
395 pressures on microbial communities, making the intertidal zone inhospitable for organisms specialized to the supra- and
396 sublittoral environments. Conceptually, this property of a dynamic environment defining a selective niche, similar to a physical
397 barrier, is worthy of further study.

398 As noted previously, the Serinhaém Estuary, where this work was conducted has since been impacted by a large off-
399 shore oil spill, the effects of which are unknown. It remains an open question if the sublittoral zone's greater abundance of
400 taxonomic diversity and enrichment in metabolic function correlate with a resilience to environmental perturbations. One could



401 hypothesize that the combination of diverse communities with organisms possessing redundant metabolic functions may be
402 more stable against perturbations as the larger standing variation of organisms will respond differently to stressors, thus
403 increasing the likelihood of the survival of some taxa (Girvan *et al.* 2005). However, while the diversity of taxa at the sublittoral
404 site may grant it certain advantages, in terms of being a more robust ecosystem, it is also in a more perilous position as the
405 water itself is frequently the carrier of contamination from rivers, as is the case for urban waste (Yunus *et al.* 2011), and from
406 the oceans through the tides, as the case with oil spill contamination (Cabral *et al.* 2016). Thus, it is important to consider that
407 different parts of the mangrove tidal zone would be exposed in different levels of contamination and that this could affect the
408 organisms in a site-specific manner. We hope that our work in characterizing what was once a pristine mangrove forest aids
409 the further exploration of the impact anthropological activities have on the microbial communities of mangrove ecosystems.

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419 7. Conflict of Interest

420 The authors have declared that no competing interests exist.



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