Contrasting drivers of abundant phage and prokaryotic communities in tropical, coastal		
ecosystems across the Isthmus of Panama		
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- 33

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- 34
- 35 ABSTRACT

36 Phages, or viruses that infect bacteria and archaea, are ubiguitous and abundant members of 37 Earth's ecosystems that impact the flow of nutrients, evolution of microbes, and food web 38 dynamics by selectively infecting and killing their prokaryotic hosts. Because phages can only 39 replicate through their hosts, they are inherently linked to processes impacting their hosts' 40 distribution and susceptibility to infection. Despite these links, phages can also be affected by 41 environmental parameters independent of their hosts, such as pH or salinity which impact cell 42 adsorption or virion degradation. To understand these complex links, in this study, we leverage 43 the unique ecological context of the Isthmus of Panama, which narrowly disconnects the 44 productive Tropical Eastern Pacific (TEP) and Tropical Western Atlantic (TWA) provinces and 45 compare factors that shape active marine phage and prokarvotic communities. Metagenomic 46 sequencing of seawater from mangroves and reefs of both the TEP and TWA coasts of Panama 47 suggest that pronounced environmental gradients, such as along the TEP mangrove rivers, 48 result in common dispersal and physicochemical parameters shaping both prokaryotic and 49 phage community composition and diversity. Conversely, we find that when environmental 50 conditions are relatively similar across adjacent habitats, such as between the mangroves and 51 reefs in the TWA, prokaryotic communities are more influenced by local abiotic conditions while 52 phage communities are shaped more by dispersal. Collectively, this work provides a framework

53 for addressing the co-variability between viruses and their hosts in marine systems and for

54 identifying the different factors that drive consistent versus disparate trends in community shifts,

55 which is essential to inform models of these interactions in biogeochemical cycling.

56

57 INTRODUCTION

58 Microbes are crucial components of Earth's ecosystems, particularly in the ocean, where they 59 form the foundation of food webs, power biogeochemical cycles, and expand the ecological niches of plants and animals^{1,2}. Outnumbering even microbes, viruses serve as major top-down 60 61 control on microbial communities and modulate microbial ecology and evolution through 62 selective killing via infections, horizontal gene transfer via transduction, and metabolic 63 reprogramming during infections³. Understanding viral impacts on microbes is critical toward 64 modeling the movement of nutrients through ecosystems⁴, the evolution of microbial pathogens⁵, and the dynamics of organismal-associated microbiomes⁶. While rapid advances in 65 66 sequencing and microscopy technologies over the past few decades have begun to unfold the 67 vast diversity, complexity, and breadth of viruses in nature⁷⁻⁹, major questions remain on which 68 factors shape viral communities and how this relates to concomitant shifts in microbial 69 communities.

70 Because viruses are restricted to reproducing through their hosts, viruses are inherently linked 71 to processes related to their hosts' distribution and susceptibility to infection. Despite these tight 72 links, patterns in the composition and diversity of these two groups can differ depending on the 73 parameter or environment. Showing coupled shifts, for instance, viral diversity and microbial 74 diversity in the ocean has been shown to increase with depth^{10,11}, and the pH of soils has been shown to co-vary with viral and prokaryotic (bacteria or archaea) diversity¹². In contrast to this 75 76 coupling, a study examining soil communities and one on communities in freshwater springs 77 showed that viral communities shifted over spatial scales and environmental parameters that did 78 not always match that of microbes^{13,14}. Several possibilities have been suggested to explain

these contrasts, such as a broader host range of viruses lowering the impact of available host composition on viral community structure¹⁵, or metacommunity dynamics¹⁴ such as the importance of high dispersal versus species local adaptation that may differ between microbes and viruses. Taken together, these studies highlight the necessity to untangle the complexity in the link between viral communities and microbial communities, to better characterize roles of microbes and viruses in the environment.

85 In this study, we leverage the unique biogeography of the Isthmus of Panama to uncover factors 86 shaping viral and microbial communities across a diverse array of tropical coastal environments 87 in two oceans. The Isthmus of Panama gradually formed and finally completely disconnected the Tropical Western Atlantic Ocean (TWA) from the Tropical Eastern Pacific Ocean (TEP) 88 89 approximately 2.8 million years ago¹⁶. The TWA became oligotrophic, leading to the proliferation 90 of reef-building corals. The TEP remained eutrophic, with patchy coral reefs dominated by fewer 91 species of scleractinian corals. Expansive mangroves thrive adjacent to coral reefs in both the 92 TEP and the TWA. Nonetheless, mangroves of the TWA are influenced by much smaller tidal 93 oscillations than in the TEP. In addition, the TWA supports thinner fringes of mangroves made 94 of shorter trees than in the productive TEP¹⁷. These contrasting coasts with parallel habitat 95 types of mangroves and coral reefs allow comparisons of viral and microbial communities at two 96 spatial scales, locally between habitat types and globally between oceans¹⁸. Given the intrinsic 97 link of viruses to their hosts, our null hypothesis was that factors shaping viral communities 98 mirror those of microbial communities, and this similarity would be most visible at global scales 99 between the oceans since the spatial separation and chemical differences between oceans are 100 so large. An alternative hypothesis is that factors shaping viral communities would not match 101 those of the corresponding microbial communities, and these differences would be most 102 apparent at smaller scales where subtle differences in environmental parameters can influence contact rates of viruses to hosts, growth rates of hosts, and other physical aspects that may 103 104 decouple viral communities from microbial communities.

105 To address these hypotheses, we examined both prokaryotic and viral community diversity in seawater metagenomes filtered for the 0.22-0.8 µm size fraction. While most known viruses are 106 107 smaller than 0.22 µm, the viruses detected in this size fraction correspond to a subset of the 108 viral community that includes larger viruses (e.g. jumbo bacteriophages), actively replicating 109 (pre-lytic) viruses, lysogenic viruses (those integrated in the genomes of hosts) or viruses that 110 have stuck to particles, putatively representing an active or abundant subset of the viral 111 community¹⁹. Because viruses of bacteria and archaea that belong to the class *Caudoviricetes* 112 are ubiquitous members of ecosystems, we focused our analyses on these viruses that we refer 113 to as phages. For the microbes, we focused on the prokaryotes, as they are the putative host 114 pool of the phages. To directly compare phage and prokaryotic diversity and minimize 115 information loss, we applied a gene-based approach. We selected marker genes for both 116 prokaryotes and phages that we benchmarked against more commonly used contig-based 117 analyses.

118 Our results reveal a variety of contexts when factors shaping phage and prokaryotic 119 communities align and when they diverge. Supporting our null hypothesis, the phage and 120 prokaryotic communities were both distinct between oceans. The importance of habitat type, 121 however, differed between these groups. Distinctions in phage community composition between 122 mangroves and reefs depended on the ocean, with the mangrove communities being highly 123 distinct from the reef communities in the TEP but not in the TWA, likely due to the strong salinity 124 gradient of the mangrove rivers in the TEP. Meanwhile, the prokaryotic communities were 125 equally distinct between the habitat types in both oceans. The lack of separation between the 126 habitat types of the phage communities in the TWA compared to the prokaryotic communities 127 suggests that changes in environmental parameters influence prokaryotic communities nearly 128 equally as dispersal limits or physical separation, while phage communities are more structured 129 by dispersal limitations than local conditions. Most strikingly, we found that phage communities 130 were more diverse in the TWA, while prokaryotic communities were more diverse in the TEP,

131 suggesting phage production or breadth of host range may differ in ecosystems of the more 132 productive island archipelago of the TEP than in the oligotrophic coastal bay of the TWA. 133 Overall, these findings highlight the necessity to examine viruses with their potential host 134 community together to better untangle processes driving their interactions with each other and 135 the environment in natural, mixed communities. The contexts when phage and prokaryotic 136 communities do not couple each other is crucial for modeling phage-host interactions as they 137 relate to microbial mortality, and ultimately biogeochemical cycling in ecosystems.

138

139 RESULTS AND DISCUSSION

140 Benchmarking methods to assess phage and prokaryotic diversity.

141 In total, fifty-seven samples of seawater from mangroves and reefs were collected from the 142 TWA and TEP coasts of Panama for metagenomic sequencing (Figure 1). To directly compare 143 phage and prokaryotic diversity and minimize information loss from the metagenomic data, we 144 benchmarked and employed a novel gene-based approach (see Methods), in which families of 145 the major capsid protein (MCP) and terminase large subunit (TerL) belonging to the class 146 *Caudoviricetes* compiled from the Virus Orthologous Groups database (vogdb.org; 147 Supplementary Dataset 2) were detected within the proteins of the contig assemblies from the 148 metagenomes (Supplementary Dataset 2). Phage contigs were also detected for comparison. 149 Reads from all samples were then mapped on to these sequences for their relative abundances 150 in each sample (Supplementary Dataset 3, see Methods), and ecological statistics held for all 151 three metrics (MCP, TerL, contigs; Supplementary Dataset 4). Results from TerL were reported 152 here as this was the most prevalent gene (Supplementary Dataset 4) and enabled direct 153 comparison with prokaryotic single-genes (versus metagenome assembled genomes). 154 Prokaryotic diversity was detected with proteins families of three genes from the Clusters of 155 Orthologous Groups (COG) database²⁰: RNA polymerase β (COG85), RNA polymerase β ' 156 (COG86), and a ribosome-binding ATPase YchF (COG12) which has been used in a previous

157 study²¹. Reads from all samples were then mapped on to these sequences for their relative 158 abundances in each sample (Supplementary Dataset 3, see Methods), and ecological statistics 159 held for all three metrics (COG85, COG86, COG12; Supplementary Dataset 4). The results of 160 RNA polymerase β (COG85) are reported here as this was the most prevalent gene in the 161 dataset (Supplementary Dataset 3). Details can be found in the Methods to use this approach 162 for other datasets and studies.

163

Proximity and physicochemical variation determine whether factors shaping phage and prokaryotic community composition align.

166 When comparing the oceans, both phage and prokaryotic community composition significantly 167 differed between the TEP and TWA (Figure 2a,b; ANOSIM *p* values < 0.05), and their variation 168 correlated with each other (Mantel test *p* value < 0.05). The phage composition, however, 169 differed to a larger extent than did the prokaryotic composition between the oceans, with only

170 12% of phages found in both oceans (Figure 2g) compared to 24% of prokaryotes detected in

both oceans (Figure 2h). Furthermore, more physicochemical parameters varied strongly with

172 phage composition than with prokaryotic composition (Figure 2a,b). The stronger distinction of

173 phage composition between oceans compared to prokaryotes may result from higher dispersal

174 limitations of most phages in the ocean. Although some phages have been detected globally²²,

this cosmopolitan distribution may be less common for phages than for prokaryotes which could

176 be investigated further in future studies.

Distinctions in community composition between the mangroves and reefs depended on theoceans. In the TEP, both phage and prokaryotic community compositions significantly differed

between the habitat types (Figure 2e,f). In the TWA, only the prokaryotic composition was

180 distinct (Figure 2d), and the phage composition did not differ (Figure 2c). In the TEP, eight of the

- 181 twelve mangrove samples were collected along two rivers, with samples spanning from fully
- 182 saline to fully freshwater. The other four mangrove samples were collected along a fully saline

183 mangrove channel (Supplementary Dataset 1). Although salinity is known as one the greatest factor limiting species ranges^{23,24}, the separation of the prokaryotic and phage communities here 184 185 appear to cluster by river rather than by salinity (Supplementary Figure 1). Nevertheless, the 186 same physicochemical parameters seemed to vary with the both phage and prokaryotic 187 community composition in the TEP (Figure 2e,f). This suggests that the phage and prokaryotic 188 communities in the TEP are likely impacted by dispersal and environmental parameters 189 similarly. Meanwhile in the TWA, physicochemical differences between mangroves and reefs 190 were less pronounced than in the TEP (Supplementary Figure 2), and these habitats were 191 closer in proximity (Figure 1). Despite the lower variation in physicochemical parameters within 192 and between reef and mangrove habitats, prokaryotic community composition partitions 193 between habitats. This suggests that prokaryotic communities may respond to factors that we 194 have not measured in this study, such as the distribution of dissolved and particulate organic 195 matter. The close proximity of the mangroves and reefs, however, may have resulted in high 196 dispersal of phages between the habitat types leading to lower distinctions in the composition, 197 as most phages were found in both habitat types (65%; Figure 2g). The dispersal of phages 198 between habitat types can result in a lag or delay in the shifts of phage community structure to 199 changes in host composition because phages can only replicate upon attaching to and infecting 200 their hosts.

Taken together, these results suggest that phage and prokaryotic community composition align when environmental conditions and spatial scales strongly structure putative host communities such as for the TEP samples (Figure 2a,e,f). Meanwhile, when these parameters are less variable, as in the TWA here, dispersal forces may structure phage communities more so than for putative host communities (Figure 2b,c,d). Another explanation between uncoupled patterns of composition in the TWA could be that physicochemical parameters interact differently on the phage and prokaryotes. For instance, pH can impact the adsorption of phages to their hosts,

208 despite the presence of their hosts²⁵. These parameters, however, would need to be tested
209 directly.

210

211 The most prevalent and influential phages and prokaryotes distinguishing the

212 communities belong to diverse taxa and ecological groups.

To determine which groups of phages and prokaryotes were driving the distinctions in the composition of communities, we classified the sequences using multiple approaches. The phages were classified based on the taxonomy of their putative host estimated by the alignment of the terminase large subunit (TerL) sequences to genes of RefSeq 207 and examining the host of the hits. RNA polymerase beta subunit (RNAP β) sequences used to represent prokaryotic diversity here were classified based on the consensus classification of the contig on

219 which the RNAP β was present (See Methods for details).

220 Of the top ten most prevalent genera based on average relative abundance across samples,

only three genera overlapped for prokaryotes and putative phage hosts: *Synechococcus*,

222 *Prochlorococcus*, and *Pelagibacter* (Figure 3). These genera are known as dominant members

of the ocean^{26,27}; furthermore, because the phage sequences may also correspond to integrated

224 phages of the prokaryotic community, this may have resulted in the co-prevalence of these

genera in both phage and prokaryotic communities. Nevertheless, the general lack of overlap in

226 prevalent phage and prokaryotic genera may have resulted from several factors such as

technical limitations in classifying both the phages and prokaryotic sequences or that most viral

228 lysis occurs for rare but highly productive microbes, as has been observed off the coast of

229 British Columbia in Canada²⁸, which would result in dominant viruses that infect rarer hosts.

230 In general, the average genus composition of both the putative phage hosts and of the

231 prokaryotes corroborate the compositional distinctions observed above when using sequence

diversity (Figure 2), with the phage communities being highly similar between mangroves and

reefs in the WA, but very distinct in the TEP (Figure 3b), and the prokaryotic communities being

234 distinct between mangroves and reefs in both oceans (Figure 3d). In both phages and 235 prokaryotes, the enrichment of *Prochlorococcus* in the TEP relative to the TWA highlights the 236 physicochemical features of the ocean, as the TEP sites were more exposed to pelagic waters 237 than the TWA sites and *Prochlorococcus* is known to be more dominant in pelagic waters than 238 coastal waters where Synechococcus is prevalent²⁹. Notably, a fully freshwater sample 239 (EPM 13A, 0 ppt salinity) only contained *Prochlorococcus* of the top genera in the putative host 240 community for the phages (Figure 4a). Prochlorococcus bacteria are rarely found in brackish or 241 freshwater conditions^{30,31}, and instead, a *Prochlorococcus*-like bacteria that is larger in cell size 242 than its marine counterpart has been reported in estuaries³¹. Thus, the presence of this phage 243 terminase with homology to that of a *Prochlorococcus* phage in the fully fresh sample suggests 244 that either (i) this phage infects this Prochlorococcus-like freshwater bacteria, (ii) that it has a 245 broad host range that enables it to infect marine and freshwater bacteria, (iii) or that its 246 homology is a result of the limitation of the reference database. Of the prokaryotic community in 247 this freshwater sample, only an unknown genus in the Proteobacteria phylum was found that 248 was also prevalent in the other samples (Figure 4c), which is unsurprising as diverse 249 Proteobacteria are common in freshwater systems³². The remarkable divergence of the genera 250 in this freshwater sample for both the prokaryotes and putative host community of the phages 251 highlights the crucial role of salinity in shaping microbial communities^{23,24}. 252 We then examined which phages and prokaryotes drove the most variation between the 253 samples, determined by those that significantly varied the most with variation in the 254 communities (envfit test; p values < 0.05; See Methods; Supplementary Dataset 5). When 255 examining all samples of both oceans and habitats, the phage (WA 000000419261 10), whose 256 terminase showed high homology to that of the *Puniceispirillum phage HMO-2011*, drove the 257 most variation followed by ten other equally influential phages that putatively infect a diversity of 258 host genera (Prochlorococcus, Puniceispirillum, Acinetobacter, Mycobacterium, Kiloniella, 259 Lacevella, Escherichia) spanning four phyla (Supplementary Dataset 5). Matching the whole

260 community distinctions between oceans and habitats of Figure 2, all but one of these 11 phages were exclusively detected in one ocean (TEP or TWA), with phages of the TWA mostly present 261 262 in both mangroves and reefs and most of those exclusive to the TEP found only in mangrove 263 samples. In contrast, variation in the prokaryotic communities was primarily driven by 264 Synechococcus bacteria (top 3 most influential; Supplementary Dataset 5), which follows its known distinction between pelagic and coastal conditions²⁹, such as the TEP between TWA 265 266 here. The elevated importance of phages predicted to infect chemoheterotrophic in driving 267 phage community composition compared to the elevated importance of photoautotrophic 268 bacteria in driving prokaryotic community composition further highlights that phage lysis 269 predominantly occurs on the most productive members of the community, which are often 270 heterotrophic bacteria that experience boom-and-bust cycles as nutrients become available²⁸. 271 When examining samples of the TWA and TEP separately, the primary genera or putative host 272 genera driving the variation in the prokaryotic and phage communities respectively aligned in 273 trophic niche for the TWA but contrasted in trophic niche for the TEP. This is surprising because 274 the phage and prokaryotic communities did not align in habitat distinction or physicochemical 275 parameters driving their composition in the TWA (Figure 2c,d) but they did in the TEP (Figure 276 2e,f). In the TWA, the two phages that drove most of the variation showed high homology to the 277 terminase of *Pelagibacter* phage HTVC008M and the *Puniceispirillum* phage HMO-2011, host genera that are both heterotrophic bacteria found throughout the global ocean^{26,33}. For the 278 279 prokaryotes, the top genera also belonged to heterotrophic groups with the top prokaryote 280 belonging to an uncultivated genus WTJO01 in the Puniceispirillales order, and the next most 281 influential belonging to an uncultivated genus UBA974 in the Flavobacteriales order. These heterotrophic bacteria are also found throughout the oceans^{26,34}. The overlap in trophic niche of 282 283 these genera for driving the phage and prokaryotic communities in the TWA, despite the 284 differences in physicochemical and habitat distinctions, highlights the robust conditions that

these groups can inhabit which could explain the lack of alignment in the environmental featuresdriving the overall phage and prokaryotic community compositions.

287 In contrast to the TWA, the putative hosts of phages driving the variation of phage communities 288 within the TEP did not align trophically despite their overlap in significant physicochemical 289 parameters and habitat distinctions (Figure 2e,f). The most influential phages primarily 290 putatively infect bacteria belonging to the photosynthetic Synechococcus genus (seven of the 291 top ten), while the most influential prokaryotes primarily belonged to unknown genera in the 292 Betaproteobacteria class (Supplementary Dataset 5). Although these genera contrast each 293 other in trophic lifestyles, these bacteria are known to be highly influenced by salinity^{31,35,36}, 294 which widely varied in the TEP as the mangrove samples were collected along freshwater 295 rivers. These results suggest that while phage and prokaryotic communities both vary 296 substantially with salinity, the types of bacteria and putative hosts of phages that are most 297 affected by salinity in these sites do not necessarily align.

298

High prokaryotic diversity is rarely coupled with high phage diversity.

300 Through selective killing by phages and resistance mechanisms by prokaryotes, phages and 301 prokaryotes are known to drive each other's evolution and microdiversity^{37,38}, but how these 302 interactions impact macrodiversity remains poorly studied. Here, we examined the alpha 303 diversity of samples to uncover which environments contain high phage and prokaryotic 304 sequence diversity. We used the Shannon's Diversity index to measure alpha diversity, as this 305 metric accounts for both richness and evenness³⁹. Taxa are proxied here as unique marker 306 sequences (see Methods). When comparing diversity between oceans, surprisingly, phage 307 communities were significantly more diverse in the TWA (Figure 4a) while prokaryotes were 308 more diverse in the TEP (Figure 4b). These patterns between the oceans held when comparing 309 mangrove and reef samples separately (Supplementary Figure 3). The contrasting diversity 310 patterns of phage and prokaryotes between oceans may be a result of several abiotic and biotic

311 factors. Although the TWA is generally more oligotrophic than the TEP, the bay where the samples were collected in this study has historically been subject to high levels of runoff, which 312 313 has been found to elevate bacterial production and density but result in decreased bacterial 314 diversity compared to nearby pristine sites⁴⁰. Thus, the reduced prokaryotic diversity in the TWA compared to the TEP may be due to the pollutants, while the elevated phage diversity in 315 316 the TWA compared to the TEP may be due to increased bacterial production and thus phage 317 replication and release. Alternatively, the higher phage diversity in an environment that has 318 lower prokaryotic diversity could be because a variety of phages infect the same hosts. This 319 would mean that a low diversity of phages could infect a high diversity of prokaryotes. The 320 ecological conditions that would enable the coexistence of diverse phages that infect the same host may be related to the contact rates with hosts⁴¹ or flux between environments introducing 321 322 novel phages, which may differ between the TEP and TWA, but these would need to be tested 323 directly.

324 We then examined diversity within each ocean between habitat types. Within the TWA, phage 325 communities were equally diverse between the mangroves and reefs (Figure 4d). Similarly, 326 prokaryotic communities were equally diverse between mangroves and reefs in TWA (Fig 4e), 327 despite significant differences in their composition between these habitat types (Figure 2d). In 328 the TEP, phage diversity was lower in the mangroves than the reefs (Figure 4g), but prokaryotic 329 diversity did not significantly differ between the habitat types, despite having significantly 330 different composition (Fig 2f). This lack of alignment between the habitat types for prokaryotes 331 in compositional variation and Shannon's demonstrates that high diversity within samples (alpha 332 diversity) does not always correspond to high diversity between samples (beta diversity)⁴². For example, a study by Walters and Martiny (2020)⁴² that compared the microbial diversity across 333 334 a range of ecosystems found that soil samples have the highest number of microbial species 335 (alpha diversity), but sediment, biofilms, and inland waters had the greatest variation in 336 communities between samples (beta diversity). The conflicting alpha and beta diversity patterns

337 of the prokaryotic communities here thus suggest that perhaps niche space is similar across 338 habitat types leading to similar alpha diversity but competition, or local adaptation, within each 339 habitat is strong enough to lead to distinct members of each community or strong beta diversity. 340 Regarding the phage diversity patterns, the lower diversity of the phage communities in the 341 mangroves than reefs in the TEP is likely due to salinity differences between these habitat 342 types, with a median 28.93 ppt in the mangroves versus. 30.655 ppt in the reefs. Furthermore, 343 the mangroves were more acidic than the reefs (median pH 7.885 vs 8.09). While these 344 physicochemical differences between TEP mangroves and reef did not manifest in prokaryotic 345 diversity differences, the phage diversity may have been impacted, as pH and salinity are known to impact adsorption rates of phages to their hosts^{25,43}. 346 347 When plotting phage diversity against prokaryotic diversity, we found that phage diversity rarely 348 correlates with prokaryotic diversity, despite the inherent link of phages to their hosts for 349 replication. In fact, in the TWA samples, their diversities appear to negatively correlate when 350 examining all TWA samples together and separately by habitat type, but this was not significant 351 (Fig 2c,f). In the TEP, there is a significant positive correlation when examining the samples 352 together (Fig 2c), but this is likely driven by the mangrove samples (Fig 2i), as there was no 353 significant correlation between phage and prokaryotic diversity the TEP reef samples (Figure 354 2i). The positive correlation of phage and prokaryotic diversity in the TEP is likely due to salinity 355 differences in the TEP mangrove samples, with two considered freshwater (~0 ppt). Upon 356 removing these two fresh samples, the correlation of diversities is no longer significant 357 (Supplementary Figure 4), highlighting the impact of extreme salinity differences on phage-host 358 interactions⁴³.

359 Differences in correlations with physicochemical parameters between phage and 360 prokaryotes help explains their decoupled relationship.

361 Because phage and prokaryotic diversity rarely correlated with each other here, we plotted 362 phage and prokaryotic diversity against the measured physicochemical parameters of each

363 environment to uncover other potential drivers of their diversity separately (Figure 5). Following 364 our hypothesis that phage and prokaryotic diversity patterns will align most when environmental 365 gradients are high, their diversities generally correlated in the same direction with most of the 366 parameters in the TEP, which had great variation in the parameters compared to the TWA 367 samples. For example, salinity varied widely between the mangrove samples of the TEP (0-30 368 ppt), and phage and prokaryotic diversity both significantly positively correlated with salinity. 369 Meanwhile, in the TWA salinity only ranged between 32 and 34 ppt, and phage and prokaryotic 370 diversities tended to correlate with the parameters in the opposite directions of each other. In 371 both mangroves and reefs of the TWA, phage diversity correlated negatively with salinity while 372 prokaryotic diversity correlated positively with salinity. 373 Taken together, the lack of strong relationship between phage and prokaryotic diversity (Figure

374 4), in addition to their inconsistent correlations with the physicochemical parameters measured 375 here (Figure 5), exemplify the nuances in the relationship between the viral and host diversity. 376 For instance, variation in phage host ranges or variation in host resistance to phage infections 377 could weaken the correlation of their diversities. Furthermore, environmental variation may 378 further decouple their diversity relationship. Our results suggest that when this environmental 379 variation or spatial distances are relatively small between sites, such as between habitat types 380 of the TWA, prokaryotic diversity may be impacted more than phage diversity by these local 381 conditions. Future work could include measuring host production, phage production, and phage 382 host ranges in isolation against different physicochemical parameters to test these hypotheses 383 more directly.

384

385 CONCLUSION

In this study, we leveraged the unique biogeography of the Isthmus of Panama to compare
drivers of phage and prokaryotic diversity at both global scales between oceans and local
scales between habitat types within each ocean by examining mangrove and reef habitats of the

389 TEP and TWA coasts (Figure 1). We found that drivers of phage and prokaryotic communities 390 align most when physicochemical and spatial scales are sharp, such as between the oceans 391 and between the TEP mangroves and reefs. Meanwhile, these factors diverge when there are 392 subtle physicochemical differences and minimal physical separation in environments, like 393 between the mangroves and reefs of the TWA. In these cases, prokaryotic communities may 394 locally adapt to the minor environmental differences, as we observed distinction between 395 prokaryotic communities of the mangroves and reefs. The phage communities, however, may 396 be influenced more by high dispersal between the environments, overwhelming environmental 397 or habitat distinctions, as we observed no significant difference between mangroves and reefs 398 of the TWA. A similar pattern has been observed in a freshwater spring system of southern 399 Florida, where the prokaryotic communities were distinct between the river, head, and mixed 400 zones, but the phages communities were not distinct between the head and mixed zone, which 401 the authors attributed potentially to high dispersal of phages between these two zones¹⁴. 402 Despite cases when drivers of phage and prokaryotic community composition align, our results 403 show that putative host genera of phages that drive phage communities differ from prokaryotes 404 in all spatial and physicochemical scales. Very few of the most dominant phage members infect 405 genera of the most dominant prokaryotes. This may be because most phages are infecting the 406 most productive prokaryotes which exhibit boom and bust reproductive cycles, rather than the 407 most stably abundant²⁸. This infection pattern would support the popular phage-host interaction 408 model, the Kill-the-Winner model, in which phages rise in abundance to kill the most dominant prokaryotes⁴⁴. However, deviations from the Kill-the-Winner model have been observed in a 409 410 freshwater lake where the abundance of some phages have been found to peak before, during, 411 or after their host's peak in abundance⁴⁵, and we would thus need time series data to resolve 412 this possibility.

Counterintuitively, we found that high phage diversity is rarely coupled with high host diversity.
The only context when their diversity correlated was in the TEP mangrove samples when fully

415 freshwater samples were included, which points to the strong role of salinity in shaping both 416 prokaryotic and phage communities^{23,43}. Because phage diversity was higher in the TWA than 417 the TEP, we suspect that host production rates may drive phage diversity such that even if there 418 are blooms of a single bacteria, a variety of phages may surface to infect that host. Conversely, 419 a lower phage diversity amidst high prokaryotic diversity may result if phages have broad host 420 ranges, which may be related to host contact rates⁴¹. We also saw that phage and prokaryotic 421 diversity can be driven by physicochemical parameters differently and align the least when 422 environmental variation and physical separation is subtle, such as in the TWA sites, compared 423 to when they are more pronounced, such as in the TEP sites. Together, these trends indicate 424 that the strength of the link of phage communities between potential host communities depends 425 on the level of variation in the environment. Future work that includes additional measurements 426 such as phage production, prokaryotic growth, organic matter concentrations, and more could 427 reveal more precise dials in the constraints on the link in patterns of phage communities to 428 those of prokaryotic communities.

429 All in all, this study provides a framework and demonstrates an application for comparing phage 430 and prokaryotic community composition and diversity in a variety of marine environments. We 431 uncover conditions when the tight links of phages and prokaryotes result in similar factors 432 driving their diversity and composition, such as between oceans, and when these tight links are 433 weakened, such as between adjacent but distinct habitat types. By understanding when these 434 phage-host links are strengthened or weakened, we can better predict the outcome of 435 interactions between phages and prokaryote populations of different environments to inform 436 models of nutrient cycling mediated by microbes and the release of organic matter through viral 437 lysis of microbes.

438

439 METHODS

440 Sample and environmental data collection.

441 Seawater samples were collected ~1m above the seafloor on coral reefs and mangroves (1-4m 442 depth) in the TEP and TWA coasts of Panama in 2017 (see Supplementary Dataset 1 for 443 coordinates and collection dates). Seawater samples were collected in sterile Whirl-Pak Bags 444 and kept on ice and in the dark until filtration at either the Smithsonian Tropical Research 445 Institute (STRI) Coiba (TEP) or Bocas del Toro research stations (TWA), where they were then 446 vacuum filtered through 0.22 µm nitrocellulose membranes (Millipore). Filters were frozen and 447 transported to STRI's molecular facility at Isla Naos Laboratory in Panama City in liquid nitrogen 448 and stored at -80 °C until DNA extractions. DNA was extracted from each filter using a Qiagen 449 Powersoil extraction kit following the manufacturer's protocol with minor modifications to 450 increase the yield46. Metagenomic shotgun libraries were prepared with the Illumina DNA 451 Nextera Flex kit following the manufacturer's protocol. Shotgun metagenomics reads were 452 sequenced on an Illumina Nextseq platform. Dissolved oxygen, temperature, salinity, and pH 453 were measured with a pre-calibrated Professional Plus handheld YSI (Yellow Springs, USA). 454 Metagenome preparation, sequencing, and assembly.

We used Trimmomatic (v0.39)⁴⁷ for adapter clipping and initial guality trimming of raw 455 456 metagenomic data (N = 57). We used anvi'o $(v7.1)^{48}$ to build a Snakemake $(v.5.10.0)^{49}$ workflow 457 for co-assembly analysis. In the workflow, we used iu_filter_quality_minoche from the Illumina Utils package (v2.12)⁵⁰ for additional guality filtering and MEGAHIT (v1.2.9)⁵¹ for co-assembly (-458 459 min-contig-len: 1000, -presets: meta-sensitive). We performed three separate co-assemblies 460 using MEGAHIT based on initial assessment of the metagenomic data. All TWA samples (reef 461 and mangrove) were co-assembled (n = 29); from the TEP, we performed one co-assembly for 462 reef samples (n = 16) and another for mangrove samples (n = 12). Next, we used anvi-gen-463 contigs-database to generate a database of contigs. Within the Snakemake workflow, KrakenUnig (v0.5.8)⁵² was used for taxonomic classification of short reads against a user-464 465 constructed database of archaea, bacteria, viral, fungi, and protozoa reads from RefSeg and the

466 NCBI nt database. Taxonomic classification of contigs was performed using Centrifuge

467 (v1.0.4_beta)⁵³, against the bacterial, archaeal, human, and viral genomes database.

468 **Phage marker gene and contig curation.**

469 For the marker gene detection, open reading frames (ORFs) were predicted with prodigal⁵⁴ (-p 470 meta -a -d) on contigs of all sizes (753,612 EP; 574,304 WA contigs | 2,168,906 EP; 1,756,476 471 WA ORFs; 3,925,382 total ORFs). Amino acid sequences of the ORFs were then searched 472 against all MCP and TerL HMM profiles available in Virus Orthologous Group database 473 (voadb.org) version 208 (Supplementary Dataset 2) using hmmsearch (hmmer.org: E value < 474 0.00001, bitscores > 41 and > 33, respectively, minimum length of open reading frame >=826 and >= 885 nucleotides, respectively). The threshold bitscores were determined by searching 475 476 proteins predicted with prodigal (default per genome) from all Caudovirales genomes from Viral 477 Genomes Portal downloaded on July 26, 2021 against the MCP and TerL profiles, taking the top 478 hit from each genome and identifying the minimum bitscore required to include at least 98% of 479 hits. After filtering for bitscore, the minimum length of a hit was decided based on containing at 480 least 98% of those reference hits. This resulted in 3,749 MCP genes and 5,369 TerL genes. 481 These were then de-replicated at 100% identity across the entire length of one sequence using BLASTn⁵⁵, which resulted in 3,722 representative MCP and 5,350 TerL (See Data Availability). 482 483 For the detection of phage contigs, contigs over 10 kilobases (7,619 EP; 10,839 WA) were run through VirSorter2⁵⁶ and CheckV⁵⁷ as follows. First, contigs over 10 kilobases (EP: 7,619, WA: 484 485 10,839;) were run through VirSorter2 (virsorter run --min-score 0.5 all) and retained if they 486 scored over 0.5 for dsDNAphage as their max_group (EP: 1,513, WA: 3,272). These contigs 487 were then run through CheckV (checkv end_to_end) to trim potential host genomes flanking the 488 contigs. Trimmed provirus and virus sequences were combined and filtered for at least 10kb 489 (EP: 1,482, WA: 3,203). The trimmed sequences were then run through VirSorter again and 490 retained if they scored over 0.95 or scored at least 0.5 and encoded at least 2 phage hallmark

491 genes. This resulted in 3,885 contigs. Virus detection summary for each contig is in

- 492 Supplementary Dataset 2.
- 493 **Prokaryote marker gene curation.**

494 The same ORF and amino acid sequences used for the phage marker gene detection were

- 495 searched against HMM profiles corresponding to genes to the Clusters of Orthologous Groups
- 496 (COG) protein families of COG0012 (COG12, ribosome-binding ATP-ase), COG0085 (COG85,
- 497 RNA polymerase β subunit), and COG0086 (COG86, RNA polymerase β' subunit)²⁰ jointly using
- 498 hmmsearch (E value < 0.00001, bitscores cutoffs of 210, 200, and 200 respectively⁵⁸. See
- 499 Supplementary Dataset 4 for the number of hits of each gene.

500 **Distribution detection.**

501 Reads from all samples were subset to an even depth to the number of reads in the sample with 502 the fewest reads (2,992,107 reads) with seqkit⁵⁹ sample (-s 1000, -2). Reads were then mapped 503 to an index of the phage marker genes, phage contigs, and prokaryote marker genes made with minimap2⁶⁰ -x sr. CoverM⁶¹ (https://github.com/wwood/CoverM) was then used for the mapping 504 505 (coverm contig --min-read-percent-identity 95 -m covered fraction rpkm count variance length --506 minimap2-reference-is-index --min-covered-fraction 0 --coupled) and retained with 50% gene covered or 20% of contig covered⁶² (Supplementary Dataset 3). See Supplementary Dataset 4 507 508 for the number of each sequence type detected in at least one sample.

509 Visualizations, statistical analyses and sequence benchmarking.

All plots aside from the maps were created in R (version 3.5.1)⁶³ with Rstudio (version 1.1.456)⁶⁴ using vegan⁶⁵, ggpubr⁶⁶, and ggplot2 (3.1.1)⁶⁷. Maps were created with QGIS (3.24) using the Voyager plug-in for the base and overlaid with sample data. Because statistics and trends held regardless of protein examined per bacteria or phage (Supplementary Dataset 4), we focused on the TerL results to represent phage diversity and COG85 results to represent bacterial diversity, as these genes were the most prevalent in the dataset. Influential sequences and physicochemical parameters were identified by those varying the most with variation in the 517 communities of all samples based on significant vector length (vegan package function envfit, 518 perm=999, na.rm=TRUE; calculated with |NMDS1-NMDS2|; p values < 0.01). Community 519 composition of samples were compared and visualized in non-metric dimensional scaling 520 (NMDS) plots using Bray-Curtis distances of relative abundances calculated with reads per kilobase per million (RPKM) using vegan (metaMDS(distance = "bray")). Two outlier samples 521 522 were excluded in the community compositional analyses as these were highly divergent 523 (WAM TWN and EPM 13A1) and skewed the results (Supplementary Figure 4, 5). WAM TWN 524 was sampled in a highly polluted site, and EPM 13A1 was sampled from a completely 525 freshwater sample, which likely resulted in their aberrant community compositions at the genus-526 level (Figure 3c,d). Significant distinctions between oceans and habitat types were determined 527 with ANOSIM test (vegan package) based on Bray-Curtis dissimilarity matrices using the RPKM 528 data (anosim(distance="bray",permutations=9999)). 529 Gene taxonomy.

530 Prokaryotic sequences corresponding to COG85 were classified via Centrifuge⁵³. For the

531 phages, amino acid sequences of TerL genes were aligned to RefSeq 207 with LAST⁶⁸ (lastal -

532 m 10 -f BlastTab; E value cutoff 10⁻⁵), and the taxonomy of the hit's host was reported (i.e. a hit

to a *Prochlorococcus* phage meant the taxonomy of *Prochlorococcus* was reported). The top hit

534 was detected based on percent identity. The top 10 genera based on average relative

535 abundance across samples was reported.

536

537 DATA AVAILABILITY

538 Reads from metagenomes will be deposited on the European Nucleotide Archive upon

539 publication. Sequences of marker genes and phage contigs can be found on the FigShare

repository upon publication, along with the VOG and COG HMM profiles used for marker gene

541 detection.

543 CODE AVAILABILITY

- 544 Custom scripts used for this study are found in the GitHub repository
- 545 (https://github.com/scubalaina/panama_phages).

546

547 SUPPLEMENTARY FIGURES

- 548 Supplementary Figures can be found at the link here:
- 549 https://github.com/scubalaina/panama_phages/blob/main/Supplementary_Figures.pdf.
- 550
- 551

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- 562

563 AUTHOR CONTRIBUTIONS

564 JJS and ML collected the samples for sequencing and the associated metadata; they also 565 processed the samples and sent them for sequencing. JJS performed initial sequence data 566 processing, quality control, and assembly. ARW and FOA designed the data analysis. ARW 567 performed the data analysis and developed the manuscript. All authors contributed to the 568 interpretation and writing of the manuscript.

569

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- 719 Figure Legends
- 720
- 721 Figure 1. Overview of project design with maps of sample locations. (a) World map with
- 722 Panama denoted as red star. (b) Map of sample sites from the Tropical Western Atlantic (TWA)
- coast of Panama. (c) Map of sample sites from Tropical Eastern Pacific (TEP) coast of Panama.
- (f) Map of TEP mangrove samples zoomed in on those collected along two freshwater rivers

and the nearby reef samples. (e) Graphical abstract of project approaches. Green triangles are
 mangrove samples. Purple circles are reef samples.

727

728 Figure 2. Comparisons of phage and prokaryotic community composition and endemism 729 in marine habitats of the Tropical Western Atlantic (TWA) and Tropical Eastern Pacific 730 (TEP). NMDS plots of samples based on phage or prokaryote community composition (Bray-731 Curtis distance), overlaid with environmental parameters significantly varying with community 732 variation. Solid lines correspond to p-values below 0.01 and dashed below 0.05. Bottom 733 barcharts compare the proportion of phages endemic to an environment or shared between 734 them (in gray). (O - dissolved oxygen, S - salinity, T - temperature). Salinity represents total 735 dissolved solids and specific conductivity, as these variables directly correlated with each other 736 in this dataset.

737

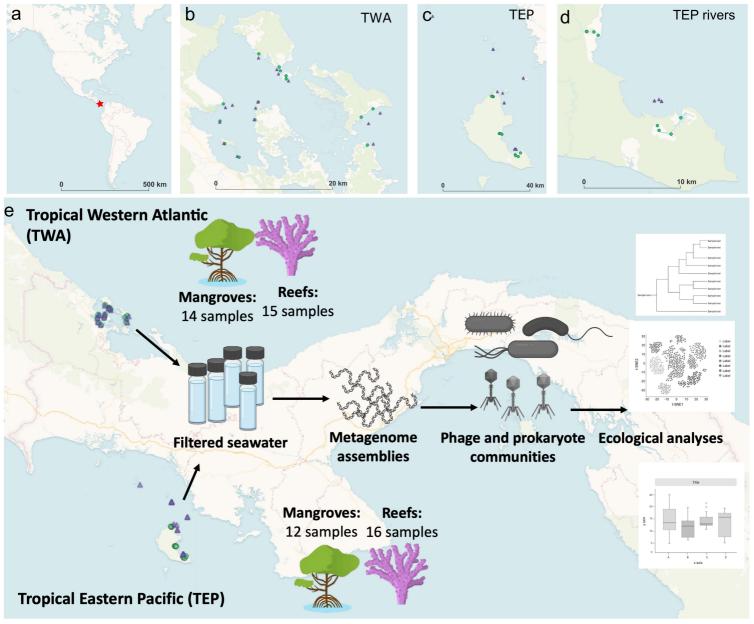
738 Figure 3. Genus composition of phage putative host communities and prokaryote

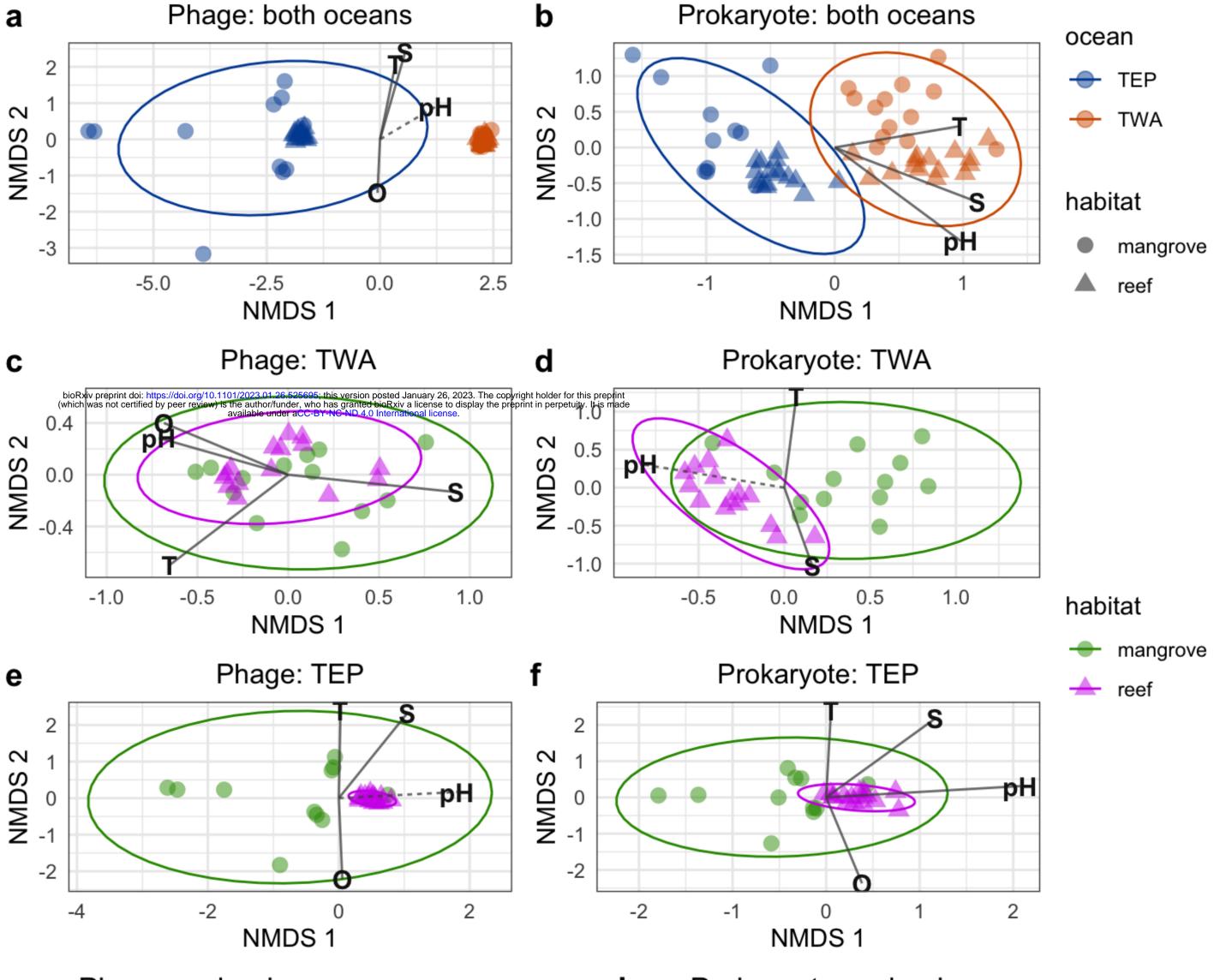
communities. Stacked barplots of the phage putative host communities (a) or prokaryotic
communities (c) colored and sorted by top ten average most abundant genera. Color strips on
bottom indicate ocean and habitat where samples were collected: top row: navy = TEP; orange
= TWA; green = mangroves; pink = reefs. (b,d) the average putative host genera composition
(b) or prokaryotic genera composition (c) of ocean and habitat type combination EPM = TEP
mangrove, ERP = TEP reef, WAM = TWA mangrove, WAR = TWA reef.

745

Figure 4. Phage and prokaryotic Shannon's Diversity in marine habitats of the Tropical
Western Atlantic and the Tropical Eastern Pacific. 2a,b,d,e,g,h are violin plots of Shannon's
Diversity of phages and prokaryotes in different samples. 2c,f,i are scatterplots of phage
Shannon's Diversity plotted against prokaryotic Shannon's Diversity in a sample, with linear
regression lines drawn and standard deviations shaded in gray. (Shan. Div. = Shannon's

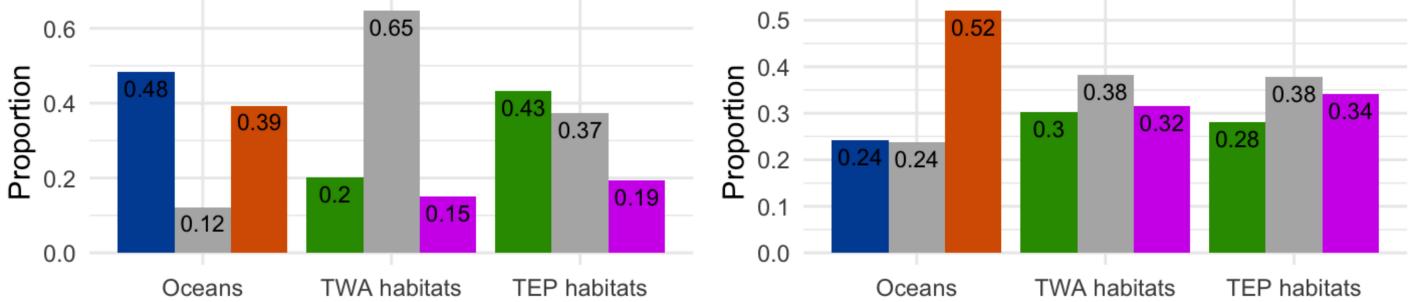
- 751 Diversity).
- 752
- **Figure 5.** Bubble plot of correlations between measured physicochemical parameters (x-axis)
- and phage and prokaryotic diversity (y-axis) of the habitats in each ocean (color strips). Color
- and size of dot correspond to correlation strength. Stars correspond to p value significance (* <
- 756 0.05, ** < 0.01, *** < 0.001). Abbreviations: T temperature, S salinity, O dissolved oxygen,
- 757 Prok prokaryote, R reef, M mangrove. Color strips: blue = TEP, orange = TWA, purple =
- reef, green = mangrove.

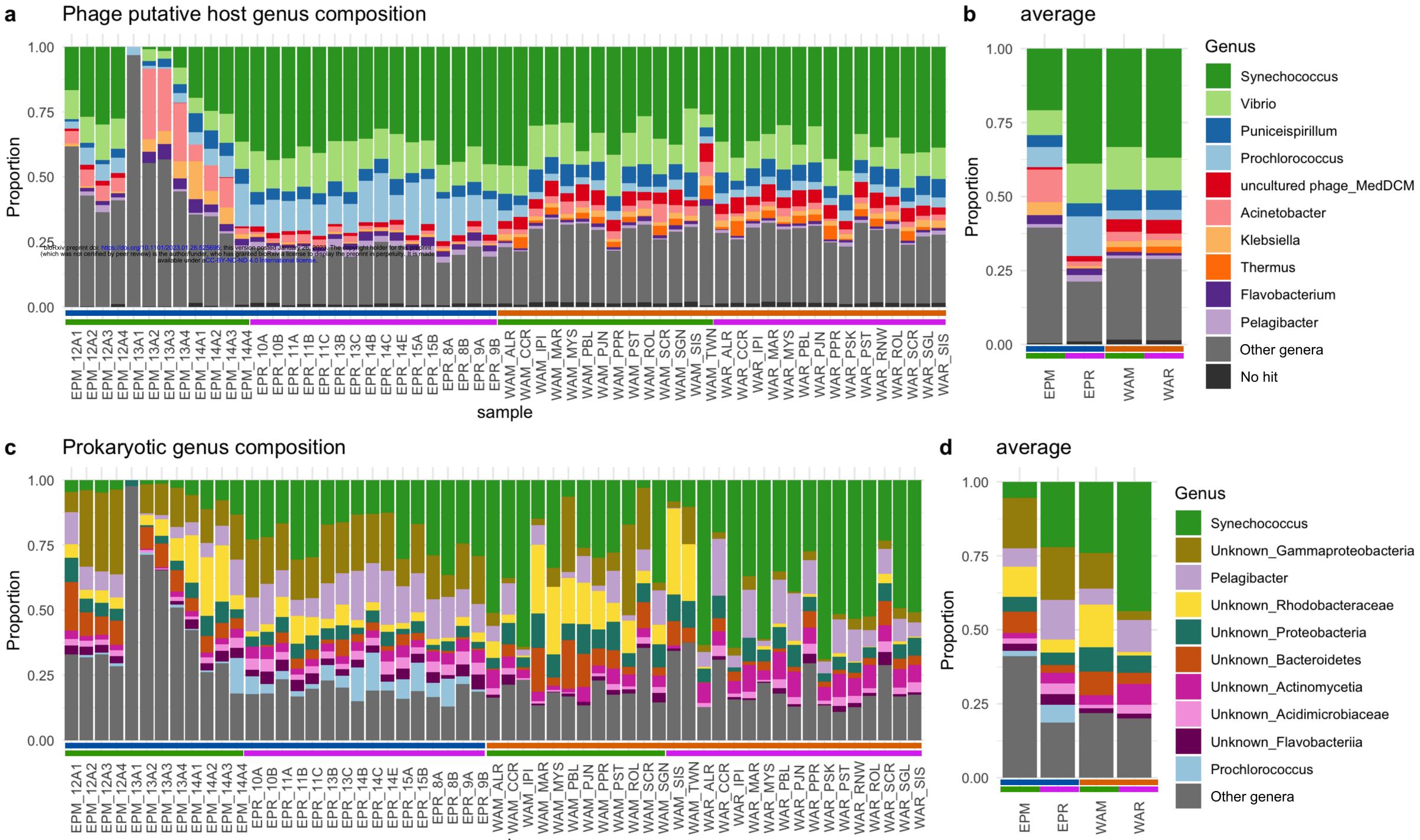




g Phage endemism

h Prokaryote endemism





sample

