



Endemicity and climatic niche differentiation in three marine ciliated protists

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Abstract:	The biogeographic pattern of single-celled eukaryotes (protists), including ciliates, is poorly understood. Most marine species are believed to have a relatively high dispersal potential, such that both globally-distributed and geographically-isolated taxa exist. Primary occurrence data for three large, easily identified ciliate species, <i>Parafavella gigantea</i> , <i>Schmidingerella serrata</i> and <i>Zoothamnium pelagicum</i> , and environmental data drawn from the National Oceanic and Atmospheric Administration's (NOAA) World Ocean Atlas were used to estimate each species' spatial and environmental distributions using Maxent v3.3.3k. The predictive power of the models was tested with a series of spatial stratification studies, which were evaluated using partial receiver operating characteristic statistics. Differences between niches occupied by each taxon were evaluated using background similarity tests. All predictions showed significant ability to anticipate test points. The null hypotheses of niche similarity were rejected in all background similarity tests comparing the niches among the three species. This paper provides a first quantitative assessment of environmental conditions associated with three species of ciliates and a first estimate of their spatial distributions in the North Atlantic, which can serve as a benchmark against which to document distributional shifts. These species follow consistent, predictable patterns related to climate and environmental biochemistry; the importance of climatic conditions as regards protist distributions is noteworthy considering the effects of global climate change.

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1 **TITLE**

2 Endemicity and climatic niche differentiation in three marine ciliated protists

3

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34 Marine ciliate niche and distribution

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42 **ABSTRACT**

43 The biogeographic pattern of single-celled eukaryotes (protists), including ciliates, is poorly
44 understood. Most marine species are believed to have a relatively high dispersal potential,
45 such that both globally-distributed and geographically-isolated taxa exist. Primary occurrence
46 data for three large, easily identified ciliate species, *Parafavella gigantea*, *Schmidingerella*
47 *serrata* and *Zoothamnium pelagicum*, and environmental data drawn from the National
48 Oceanic and Atmospheric Administration's (NOAA) World Ocean Atlas were used to
49 estimate each species' spatial and environmental distributions using Maxent v3.3.3k. The
50 predictive power of the models was tested with a series of spatial stratification studies, which
51 were evaluated using partial receiver operating characteristic statistics. Differences between
52 niches occupied by each taxon were evaluated using background similarity tests. All
53 predictions showed significant ability to anticipate test points. The null hypotheses of niche
54 similarity were rejected in all background similarity tests comparing the niches among the
55 three species. This paper provides a first quantitative assessment of environmental conditions
56 associated with three species of ciliates and a first estimate of their spatial distributions in the
57 North Atlantic, which can serve as a benchmark against which to document distributional
58 shifts. These species follow consistent, predictable patterns related to climate and
59 environmental biochemistry; the importance of climatic conditions as regards protist
60 distributions is noteworthy considering the effects of global climate change.

61 INTRODUCTION

62 Marine microorganisms are essential to the planet's health, marine biodiversity, and the
63 fishing industry. They carry out nearly 50% of the world's photosynthesis, 70% of which is
64 locked up in long-term carbon storage (Field et al. 1998). Microbes are the base of the marine
65 ecological pyramid, and are vital to the marine food web that transfers energy from primary
66 producers to higher trophic levels (Azam et al. 1983; Fenchel 2008). Understanding the
67 current spatial distributions of marine microorganisms is thus important to improving
68 understanding of their functions, and is fundamental for impact assessment of effects of
69 climate change.

70

71 Microorganisms—used here to refer to all microscopic organisms (including protists,
72 archaea, bacteria and fungi)—are generally considered to be cosmopolitan in distribution
73 (Beijerinck 1913): “everything is everywhere, but the environment selects”. They are thought
74 to be ubiquitous thanks to huge population sizes and a consequent low probability of
75 extirpation; also, they are limitlessly dispersible, have negligible rates of allopatric speciation
76 compared to larger organisms, and can potentially occur at any place that meets their
77 biological requirements (Finlay et al. 1996). These assumptions suggest that processes
78 driving biogeographic patterns of microbe diversity are fundamentally different from
79 macroscopic organisms, with profound implications for understanding mechanisms driving
80 microbial distributions and evolution. A variant on this view states that whereas most free-
81 living protist species are globally distributed, about one-third of them may be
82 biogeographically restricted (Foissner 1999; Foissner 2006).

83

84 Nonetheless, microbial geographic patterns are still poorly understood. Description of the
85 biogeography of protists is hampered by a scarcity of taxonomists, high frequency of

86 misidentification and dramatic under-sampling (Foissner 2006). Traditionally, protists have
87 been identified by morphological features, though this approach risks lumping organisms
88 with discrete biogeography, ecology and genetics into single morphospecies (Gentekaki and
89 Lynn 2010; Katz et al. 2011). The high rates of synonymy and existence of polymorphic life
90 stages (phenotypic variability) in ciliates can exacerbate the problem (Dolan 2016). Protist
91 biogeography has benefited from molecular studies, as environmental sequencing can provide
92 abundant distributional data. Several contributions have described geographic
93 circumscription in 18 S rDNA sequences from diverse aquatic protist groups—diatoms
94 (Evans et al. 2009), Cercozoa (Bass et al. 2007), Amoebozoa (Aguilar et al. 2014),
95 heterotrophic flagellates (Boenigk et al. 2006), and ciliates (Bass et al. 2009). Several recent
96 morphological and molecular studies have found that many protists, particularly the rarer
97 ones, follow discrete biogeographic patterns (Filker et al. 2016; Logares et al. 2015; Segovia
98 et al. 2017). Further studies relating morphology, functionality and molecular information are
99 needed to describe the dispersal ability of these and other protist species.

100

101 Ciliophora is a diverse phylum of heterotrophic or mixotrophic protists—the ciliates—with
102 ~4500 taxonomically valid, free-living species (Foissner et al. 2008), characterized by the
103 presence of hair-like organelles called cilia, used for locomotion, attachment, feeding,
104 sensation, etc., and nuclear dimorphism (Gao et al. 2016). They have been sampled from
105 marine environments; indeed, the Ocean Biogeographic Information System (OBIS) has
106 locality data for nearly 200,000 Ciliophora observations, albeit with a clear bias towards
107 sampling along major trade routes. Some microbial species are more conspicuous and easier
108 to identify with confidence than others, leading some authors to propose these “flagship”
109 species as the ultimate proof of endemism (Foissner 2006). Such species, the “elephants” of
110 the microbial world, cannot be missed if they are present, because of their distinctive

111 morphological features and/or significantly large size, with the caveat that traditional
112 morphological-based identification of microbial species risks lumping diverse ecological and
113 genetic species into a single morphospecies complex and underestimating cryptic diversity
114 (Gentekaki and Lynn 2010).

115

116 Ecological niche models (ENM) infer suitable abiotic habitat conditions for non-model
117 organisms by generating a correlational model that unites occurrence information and
118 environmental data for the taxon of interest to determine the geographic distribution of
119 habitat conditions correlated with species occurrences (Soberón and Peterson 2005). They
120 have been used to explore diverse topics in distributional ecology, including species'
121 geographic distributions, niche conservatism, spread of invasive macrospecies and diseases,
122 and effects of climate change on species distributions; see Peterson et al. (2011) for a
123 summary. ENMs have long been used to infer distributions of species in the marine
124 environment (Wiley et al. 2003). To date, marine ENMs have been used mainly to create
125 models for macrofaunal distributions (Bentlage et al. 2013; Saupe et al. 2014), although some
126 authors have applied ENM approaches to characterize phytoplankton (Brun et al. 2015) and
127 Foraminifera (Langer et al. 2013). Relatively few attempts have been made to describe the
128 ecological dimensions of microbial distributions for free-living terrestrial microbes (Aguilar
129 and Lado 2012). One noteworthy recent contribution explored the phylogeography of 18S
130 rDNA variants of the myxomycete *Badhamia melanospora* (phylum Amoebozoa) (Aguilar et
131 al. 2014).

132

133 The aim of this contribution is to show, for the first time in ciliates, the utility of ENMs in
134 putting ciliate morphospecies on the map, that is, in analysing and understanding their
135 biogeography. We focus on three flagship species from the phylum Ciliophora: *Parafavella*

136 *gigantea* (Brandt, 1896), *Schmidingerella* (= *Favella*) *serrata* (Möbius, 1887), and
137 *Zoothamnium pelagicum* Du Plessis, 1891. These species are particularly obvious thanks to
138 their large body size. *Parafavella gigantea* (up to 750 μm in length) and *Schmidingerella*
139 *serrata* (up to 350 μm long) are relatively obvious within plankton samples as they are much
140 larger than many other protists. *Zoothamnium pelagicum*, though somewhat smaller (up to
141 120 μm), is colonial, and often forms discoidal clusters 2-3 mm thick that are clearly visible
142 to the human eye (Laval 1968). These were the only species for which sample size and
143 taxonomic clarity were sufficient to permit developing ecological niche models (ENMs).

144

145 **METHODS**

146 **Data on ciliate distributions and ecological dimensions**

147 Primary occurrence data for Ciliophora species were downloaded from the Global
148 Biodiversity Information Facility (GBIF; <http://www.gbif.org/>) and Ocean Biogeographic
149 Information System (OBIS; <http://www.iobis.org/>) on 10 December 2015. We obtained more
150 than 200,000 observations, but most were not usable for development of ENMs: many were
151 not supported by voucher specimens or sequences (which we consider to be necessary
152 documentation), and taxonomic resolution for many taxa was poor. Observations were often
153 duplicated from the same locality; and sample sizes were low for most taxa.

154

155 We found only three ciliate species with sufficient sample size, taxonomic clarity, and
156 geographic spread to develop ENMs: *P. gigantea* ($N = 368$), *S. serrata* ($N = 135$), and *Z.*
157 *pelagicum* ($N = 279$). Records for these taxa were collected between 1903 and 2008 (see Fig.
158 1); duplicate records falling within the same 1° grid square were eliminated. The area of
159 coverage was limited to the North Atlantic (20°E-80°W; 0°-90°N) centred on the best-
160 sampled ocean regions worldwide. Occurrence data were visualized in ESRI ARCGIS 9.3.1

161 (ESRI, Redlands, CA, USA), and showed broadly consistent distributional patterns. The few
162 data (<2% for each taxon) that fell outside this region were nonetheless included in
163 development of models to avoid prejudging limits of unknown distributional patterns (see
164 Fig. 1).

165

166 Environmental data layers were drawn from National Oceanic and Atmospheric
167 Administration's (NOAA) World Ocean Atlas (NOAA 1999), at a native spatial resolution of
168 1°: ~110 km at the Equator. These data layers represent long-term annual means from
169 oceanographic datasets covering 1900-1997. We developed two models: Model 1, based on
170 six annual surface temperature and salinity variables (mean, maximum and minimum, for
171 each), and Model 2, based on the same six variables for surface temperature and salinity, as
172 well as six annual average surface biochemical variables (Supplementary Table 15). The data
173 for each environmental coverage layer were converted into standard normal variates in
174 ArcGIS 9.3.1 prior to analysis.

175

176 **Ecological Niche Modelling**

177 ENMs were calibrated for each species using the maximum entropy algorithm Maxent
178 v3.3.3k (Phillips et al. 2006). Maxent is a correlational algorithm that uses presence-only
179 species occurrence data to estimate ecological niche parameters (Phillips et al. 2006). Maxent
180 fits a suitability surface for the species of interest to the set of pixels across the study region,
181 maximising the entropy of the probability distribution but constrained to return higher per-
182 pixel suitability scores for pixels with environmental variable combinations most like those at
183 which the species has been detected. That is, the resulting suitability surface is a raster map in
184 which each pixel is scored regarding its similarity to climatic and biochemical conditions at
185 sites where the species has been observed.

186

187 From Maxent suitability surfaces, distributions of suitable areas for each species were defined
188 using a minimum training occurrence threshold (Pearson et al. 2007). We defined two
189 thresholds for suitable and unsuitable areas for each species: the least presence threshold
190 (LPT)—equal to the lowest probability at any occurrence location—and a more conservative
191 10th percentile training presence threshold, which is the highest suitability value that includes
192 90% of the calibration data. LPT is appropriate when there is no error in occurrence data that
193 may be introduced during geo-referencing or identification, whereas the 10th percentile
194 approach anticipates up to 10% error among occurrences. More complex methods of
195 thresholding have demonstrated better skill in classifying suitable and unsuitable
196 environments (Jimenez-Valverde and Lobo 2007; Liu et al. 2005), but our dataset lacks true
197 absence data, so a simple omission-based technique was preferable (Bean et al. 2012). In
198 addition, under this approach, the map of suitable conditions for a species is defined to be
199 inclusive of all habitable conditions, and as such should include all known populations
200 (Peterson 2014).

201

202 **Testing predictive power of ENMs**

203 This study examines whether the occurrences of three ciliate taxa follow a consistent and
204 predictable environmental pattern in the Atlantic Ocean. As such, we tested whether models
205 could predict independent subsets of occurrence data reliably. These tests covered the entire
206 study area, using replicate random subsets of the occurrence data, and spatially stratified
207 subsets of the occurrence data, 50% of localities for model calibration and 50% for testing for
208 both models under default parameters, and choosing logistic output format with suitability
209 values from 0 (unsuitable) to 1 (suitable). To avoid extrapolation in model features, no

210 clamping or extrapolation was permitted (Owens et al. 2013); to avoid overly complex
211 models, no hinge or threshold features were permitted.

212

213 For prediction across the study area, 10 bootstrap replicate runs with a maximum of 10,000
214 iterations were conducted using a random seed with 70% of occurrence points. For spatial
215 stratification challenges, occurrence data were split spatially into quadrants above and below
216 the median longitude and latitude of the occurrence data. From this spatial stratification, we
217 developed the following three pairs of quadrants: west versus east of the median longitude,
218 north versus south of the median latitude, and on-diagonal (upper left-hand and lower right-
219 hand quadrants) versus off-diagonal (lower left-hand and upper right-hand quadrants). In
220 each case, we developed both reciprocal predictions, testing the ability of ENMs to anticipate
221 the spatial distribution of occurrence data in areas for which no sampling is available. Models
222 were evaluated by applying partial receiver operating characteristic (ROC) statistics to the
223 50% subset of occurrences withheld from model development for testing. Area under the
224 curve (AUC) ratios were calculated using the Partial ROC function available in CONABIO's
225 NicheToolBox (<http://shiny.conabio.gob.mx:3838/nichetoolb2/>). Final models were
226 developed using 50% of available data.

227

228 One concern in comparisons of models based on different sets of environmental data is that
229 increased model complexity may lead to overfitting (Peterson et al. 2007; Radosavljevic and
230 Anderson 2014). A recent tendency, as a consequence, has been to use the Akaike
231 information criterion (AIC) as a means of comparing model likelihood values while
232 penalizing complex (and ostensibly overfit) models (Warren and Seifert 2011). Specifically,
233 we calculated the sample-size-corrected AICc statistic using ENMTools version 1.3.3
234 (Warren et al. 2010), and chose as the "best" model the one that had the lowest AICc values.

235 Results based on AIC (without the sample-size correction) and the alternative Bayesian
236 information criterion (BIC) metric were similar, and so are not presented.

237

238 **Niche difference and range restriction**

239 To assess whether niches differed among taxa (which would imply range limits), we
240 evaluated differences in niches occupied by each taxon using background similarity tests
241 (Warren et al. 2008) available in the ENMTools R package version 0.1
242 (<https://github.com/danlwarren/ENMTools>). This test considers similarity between predicted
243 geographic distributions among species, using two statistics, Schoener's D and a modified
244 Hellinger's I metric, to quantify similarity. This test evaluates whether ENMs generated from
245 two species are more different than expected when occurrences are drawn from the same
246 underlying distribution across the region accessible to each species. The test allows
247 specification of an area of analysis ("the background"), which we equate with the area
248 accessible to a species over relevant time periods (Barve et al. 2011; Soberón and Peterson
249 2005). Numbers of points sampled from the background were set at observed sample sizes. In
250 each test, 100 replicate analyses were performed to estimate probabilities associated with null
251 hypotheses of niche similarity. The hypothesis that species were no more like each other than
252 if points had randomly been drawn from the study area was rejected if observed similarity
253 between models fell below the 5th of the null distribution.

254

255 **RESULTS**

256 The area under the curve (AUC) ratios for the independent *Parafavella gigantea* testing data
257 were 1.77 and 1.79 for Model 1 and Model 2, respectively ($P < 0.05$; Table 1). The 10%
258 threshold for both models predicted a broad potential distribution across the cold temperate
259 zone of the North Atlantic north to about 79 / 81° N (Model 1 / Model 2) in the eastern

260 Atlantic, and about 67° N (both models) off the Greenland coast (Fig. 2) and the Labrador
261 Sea. The southerly limit was at about 48° / 47° N in the eastern Atlantic, and 36° / 38° N off
262 the US coast. Both models showed a more northerly prediction in the eastern Atlantic than in
263 the western Atlantic, owing to the ameliorating effect of the Gulf Stream / North Atlantic
264 Drift / Norwegian Current. Locality points with a borderline prediction for both models were
265 principally from the poorly represented southeastern quadrant of the study area, the
266 Greenland Sea in the northeast, and Labrador Sea and Hudson Bay in the northwest. The 10th
267 percentile training presence fractional area predicted was about 22% of the study area for
268 each model, whereas that predicted under least presence threshold (LPT) was about 43 / 44%
269 of the study area.

270

271 The area under the curve ratios for the independent *Schmidingerella serrata* testing data were
272 1.83 and 1.81 for Model 1 and Model 2, respectively (Table 1). Both models predicted a
273 broad distribution in the North Atlantic from about 74° / 71° N in the eastern Atlantic, around
274 65° N off the Greenland coast, and 52° N off the Canadian coast (Fig. 2). The southern
275 predicted limit extended from about 45° / 46° N in the European coast, curving slightly to the
276 north in the mid-Atlantic, and about 36° N off the US coast. Most “borderline” points were
277 either in the Labrador Sea and the northwest Atlantic or the southeast part of the distribution,
278 including the Mediterranean. The least presence threshold prediction was similar between the
279 two models, covering about 30% and 28% of the study area, respectively. The proportion of
280 the study area predicted by the 10th percentile area was 18.3% by Model 1, compared to
281 18.9% in Model 2.

282

283 The area under the curve ratios for the independent *Z. pelagicum* testing data were 1.76 and
284 1.75 for Model 1 and Model 2, respectively (Table 1). Both predicted (at the 10th percentile

285 training presence fractional area) a broad distribution in the North Cold Temperate Atlantic
286 from about 72° / 71° N in the eastern Atlantic, respectively, to around 52° N off the Canadian
287 coast (Fig. 2). The southern predicted limit included much of the North Temperate Atlantic,
288 to about 36° N for both models off the US coast, almost all the Mediterranean, and even some
289 Tropical areas (to 18° / 15° N in the southeast Atlantic). Model 1's least presence threshold
290 prediction was broader than that of Model 2 (66.7% / 61.4% of the study area, respectively).
291 By contrast, the 10th percentile prediction was rather small for each, and, again, more
292 restricted for Model 2: 30.0% / 24.5% of the study area. In each case, it was restricted to a
293 central band (most of the North Temperate Atlantic). All 10 replicates of spatial stratification
294 tests for each taxon and for both models showed significant ability to predict test points ($P <$
295 0.05; see Table 1 and Supplementary Material).

296

297 All tests comparing niches of the three species rejected the null hypothesis of niche similarity
298 between pairs of species when compared to a null distribution generated from the background
299 region ($P < 0.05$; Figure 3, Supplementary Material Table 13). *Parafavella gigantea* was
300 predicted to range in the cold temperate zone of the North Atlantic, extending to the Arctic
301 Ocean, whereas the predicted distribution of *Z. pelagicum* was more southerly, extending to
302 the Caribbean and the Equator; the predicted distribution of *S. serrata* was intermediate. The
303 sample-size corrected Akaike information criterion (AICc) statistics, in all three cases,
304 indicated that the simpler models based only on six annual surface temperature and salinity
305 variables (i.e., Model 1) were preferable to the more complex models that included six
306 biochemical variables as well (Model 2), as differentials in AICc were >90 in all three cases
307 (Supplementary Material Table 14).

308

309 Because all major areas of the predicted-suitable area for each of the three species appeared
310 to be inhabited--at least as far as limited sampling permitted us to conclude (e.g., in parts of
311 the North Sea)--we conclude that these species likely have quite-excellent dispersal abilities.

312

313 **DISCUSSION**

314 Documentation of spatial patterns of microbial species remains undefined and contentious,
315 owing to the complicated nature of detecting them over wide areas. Evidence for
316 cosmopolitan distributions has been demonstrated for several protistan lineages (Cermeño
317 and Falkowski 2009; Richards et al. 2005). However, many microbial species have broad but
318 restricted distributional patterns (Bass et al. 2009; Bass et al. 2007; Foissner 2006); see Bass
319 and Boenigk (2011) for a comprehensive review. This pattern of moderate endemism is also
320 seen in prokaryotic microbes (Noguez et al. 2005; Tamames et al. 2010). Some species are
321 even endemic to discrete geographic areas and ecosystems (Foissner 2006). It is unclear to
322 what extent microbial biogeography is obscured by poor understanding of species limits.

323 Only one study (Aguilar et al. 2014) explored the spatial distribution of a single
324 morphospecies of amoeba, *Badhamia melanospora*, using molecular genotyping and ENMs.
325 The authors detected two geographically-structured groups of ribotypes for *B. melanospora*,
326 each of which showed limited distributions, and concluded that this species is not
327 cosmopolitan. Thus, it may be the case that morphospecies mask microbial diversity and
328 biogeographic patterns (Gentekaki and Lynn 2010; Katz et al. 2011).

329

330 Our ecological niche models for three ciliophoran species clearly detected an environmental
331 signal unique to each, such that each species occupies a distinct fundamental ecological
332 niche. The area identified as suitable for each species varied greatly on broad spatial scales,
333 ranging from about 18% of the study area for *S. serrata* to 25-30% for *Z. pelagicum* (10%

334 threshold). None of the species is likely to be ubiquitous across the entire study area and they
335 showed statistically significant differences in their environmental characteristics. Our models
336 demonstrated that the distribution of each species is constrained by environment, particularly
337 maximum and minimum temperatures.

338

339 Our locality data were drawn from specimens from the North Atlantic Ocean, the region for
340 which sampling is most dense and complete, although we are conscious that these species'
341 ranges may extend more broadly. Our models were consistently able to predict non-North
342 Atlantic locality data, with the following exception: *S. serrata* beyond the study area are
343 found off the coast of Ecuador and Peru, the North Pacific, and the eastern Mediterranean.
344 Although the latter two sets of points were well predicted by our models, the Ecuador-Peru
345 points were in an area with higher temperature and salinity than those used to train our *S.*
346 *serrata* model, and thus fell outside the ecological niche estimated in our models. Still, the *S.*
347 *serrata* models showed a good fit for the North Atlantic locality data, and provide a
348 parsimonious prediction. The non-Atlantic *S. serrata* points may form part of the natural
349 distribution of the species, or represent recent, perhaps human-mediated, invasion. For
350 instance, the points from the Ecuador-Peru coast might be explained by natural dispersal and
351 colonization by the species following the opening of the Panama Canal (similar to the
352 Lesseps immigrant Indo-Pacific species that entered the Mediterranean via the Suez Canal),
353 or direct transport in a ship's ballast water (Foissner 2011). However, we know of no
354 evidence that can clarify which is the case.

355

356 Additional documented populations for the other two ciliates comprised occurrences that
357 were highly consistent with our predictions. *Parafavella gigantea* has been detected in the
358 North Pacific and along the Russian northern coastal areas, consistent with our North Atlantic

359 prediction. The *P. gigantea* models showed a good fit for the North Atlantic locality data, in
360 that they recovered the areas where the species has been detected, without predicting an area
361 too broad to be credible; however, the North Pacific was not included in the model
362 predictions. Additional locality data for *P. gigantea* compiled in a recent study (Dolan et al.
363 2017) comprised 38 additional locality points for the species, and all were anticipated in our
364 least presence threshold model (significant $P < 0.05$). Few locality points for *Z. pelagicum*
365 were available from outside the North Atlantic ($N = 6$), all from the North Pacific. An
366 exhaustive literature search for additional records revealed no additional locality data for *Z.*
367 *pelagicum*. Our model does not provide a prediction for the North Pacific distributions of
368 these three species. However, climatic and biochemical characteristics of the North Pacific
369 are broadly similar to those of the North Atlantic, and we anticipate that these locality points
370 would be predicted by broader analyses.

371

372 The *S. serrata* populations occurring between the Galapagos and the Peruvian coast occupy a
373 tropical niche, quite distinct from the cold temperate distribution of populations detected in
374 the North Atlantic. The true environmental tolerances of this species may indeed be broader
375 than the model we have developed here. If so, our ENM requires locality data from areas with
376 higher temperature and salinity to reflect this ecological tolerance, and our model would thus
377 fail to represent the full fundamental ecological niche. Alternately, the tropical Pacific
378 population may have ecological tolerances distinct from those of the North Atlantic
379 population, and may even represent a distinct species within the morphospecies, suggesting a
380 need for investigation of the ecophysical constraints and genetic distinctiveness of the two
381 populations. Multiple phylogeographic studies on diverse protists, but not ciliate groups
382 (Aguilar et al. 2014; Bass et al. 2007; Evans et al. 2009), have shown distinct geographic
383 ribotypes in morphological species. To the best of our knowledge, no studies have assessed

384 phylogeographic patterns in any of the taxa investigated here. This study demonstrates that
385 three common, morphologically conspicuous and widespread morphospecies occupy distinct
386 geographic distributions and ecological niches in the North Atlantic.

387

388 Our occurrence data were collected over a relatively long time period, just over a century
389 (1903-2008). Global climate change and short-term regional climatic phenomena, including
390 the North Atlantic Oscillation, the Atlantic Multidecadal Oscillation, and the El Niño-
391 Southern Oscillation, may have significant effects on distributions of the three ciliate species
392 analyzed. The environmental data layers used as input to our models were averaged over
393 much of the twentieth century, which overlaps well with the temporal provenance of the
394 occurrence data. These considerations may introduce minor biases, nonetheless, owing to a
395 rise in global mean surface temperature of about 0.1°C toward the end of the century (Hansen
396 et al. 2010). More generally, the environmental signature for a given pixel was assigned an
397 average value in our analyses, rather than the values for the year in which the sample was
398 collected. Such models with finer temporal resolution can and should be developed to resolve
399 this issue, considering the appropriate environmental regime over the period of specimen
400 collection. We aim to take this “next step” when we can develop both the relevant data layers
401 and occurrence data that are sufficiently rich and dense to permit such analyses. For now,
402 however, our models provide a first estimate of environmental envelopes for these three
403 ciliate species, with the caveat that predictions may involve a slight northern bias that is
404 unlikely to affect the argument regarding the ubiquity hypothesis versus moderate
405 endemism.

406

407 Predictions across the study area based on different environmental data sets were similar for
408 each species. In all models, minimum and maximum temperature contributed the most to the

409 prediction, although the percent contribution of these two temperature coverages was reduced
410 by the introduction of additional data layers for the Model 2 series. The additional coverages
411 of phosphate and silicate content were most important for *P. gigantea* and *S. serrata*, whereas
412 oxygen saturation and silicate content were most important for *Z. pelagicum*. In contrast, the
413 salinity layers and the apparent oxygen usage layer were relatively unimportant in all models.
414 Model 2 provided finer resolution for each taxon, predicting less of the study area and
415 omitting fewer of the independent evaluation points at the least presence threshold,
416 suggesting a better model. However, the model complexity evaluation indicated clearly that
417 the increased detail of Model 2 for each species did not outweigh the negative effects of
418 increasing model complexity and dimensionality, such that Model 1 was preferable for each
419 species. A clear link exists between the biogeography of each of the species and climatic
420 conditions, so distributions of the species will likely change with climate change. This
421 possibility merits further investigation, considering the current scientific focus on
422 anthropogenic global climate change, and other short-term regional climatic phenomena. This
423 study is a first reference point for the distribution of each species, and can serve as a
424 benchmark against which to compare future distributional patterns.

425

426 **CONCLUSIONS**

427 This study represents a first effort to describe the spatial and environmental distributions,
428 using ecological niche models, of three species of ciliated protists, a group of microorganisms
429 that are essential to marine carbon cycling and trophic chains, marine biodiversity, and even
430 to the fishing industry (Caron and Countway 2009; Lom and Dyková 1992) but for which
431 geographic distributions are poorly documented. This study thus serves as a first reference for
432 the distribution of each species, and as a benchmark against which to compare potential
433 future distributional patterns. Although future work remains to be done to refine these

434 models, particularly to consider climate variability, our findings point clearly to a situation in
435 which each of the studied species has a unique environmental signature and geographic
436 distribution.

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- 570

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583

584 SUPPLEMENTARY MATERIAL

585 This document provides a summary for all data management and spatial stratification
586 procedures followed in developing predictive models for three ciliate species; also provided
587 are heuristic estimate of relative contributions of the environmental variables, maps of all 36
588 spatial stratification predictions, results of the tests to determine range restriction using
589 ENMTools background similarity, and a summary environmental layers used for model
590 development.

591

592 DATA ACCESSIBILITY

593 Primary occurrence data are fully and openly accessible via the Global Biodiversity
594 Information Facility (GBIF; <http://www.gbif.org/>) and Ocean Biogeographic Information
595 System (OBIS; <http://www.iobis.org/>). Environmental coverages were drawn from National

596 Oceanic and Atmospheric Administration's (NOAA) World Ocean Atlas (NOAA 1999). Full
597 results are available in Supplementary Material.
598

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599 **FIGURE LEGENDS**

600 **Figure 1.** Global map showing distribution records of *Parafavella gigantea* (A; $N = 4495$),
601 *Schmidingerella serrata* (B; $N = 1536$), and *Zoothamnium pelagicum* (C; $N = 1738$) based
602 on data drawn from the Ocean Biogeographic Information System (OBIS;
603 <http://www.iobis.org>), and Global Biodiversity Information Facility (GBIF;
604 <http://www.gbif.org/>), both accessed 10.12.2015. The sample sizes that we used for model
605 development were greatly reduced as we included only the locality data from vouchered
606 reference material, and removed all duplicate records of a species at any given site.

607

608 **Figure 2.** Predictions of suitable areas across the study area for each of three species, based
609 on models calibrated using two environmental coverage sets.

610

611 **Figure 3.** Model 1 background similarity tests comparing the niche occupied by each taxon
612 using the background similarity tests available in the ENMTools R package version 0.1.
613 Schoener's D and a modified Hellinger's I metric, are used to quantify the similarity of two
614 probability distributions. In each case, the niche occupied by each taxon is more different
615 than expected from the study region. Model 2 background similarity tests are shown in
616 Supplementary Material Figure 14.

617

618 **TABLES**
 619 **Table 1. Summary table of ecological niche modelling results for both models of three**
 620 **species of ciliates.** Results of spatial stratification models are shown in Supplementary
 621 Material.

Species	<i>N</i> training / testing points ¹	Testing AUC ratio ²	Fractional predicted area LPT ³	Fractional predicted area 10% PT ⁴
All data predictive models				
<i>P. gigantea</i> 1	184 / 184	1.77	0.430	0.221
<i>P. gigantea</i> 2	184 / 184	1.79	0.442	0.220
<i>S. serrata</i> 1	68 / 67	1.83	0.304	0.183
<i>S. serrata</i> 2	68 / 67	1.81	0.283	0.189
<i>Z. pelagicum</i> 1	140 / 139	1.76	0.667	0.299
<i>Z. pelagicum</i> 2	140 / 139	1.75	0.614	0.245

622

¹ *N*, number of points used for model training / testing.

² Training/ testing AUC data, area under the curve of the receiver operating characteristic calculated using NicheToolBox (<http://shiny.conabio.gob.mx:3838/nichetoolb2/>).

³ LPT, Least Prediction Threshold.

⁴ 10% PT, predictive threshold that excludes the 10% most outlying points.

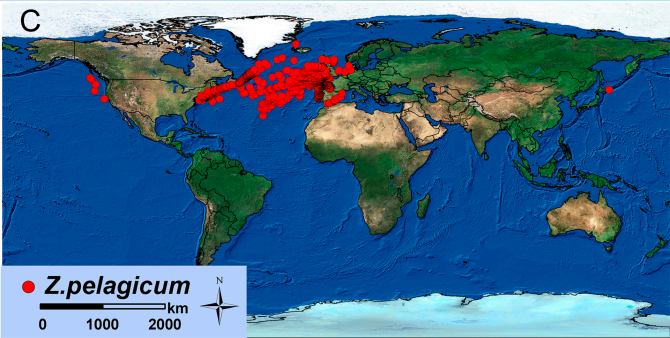
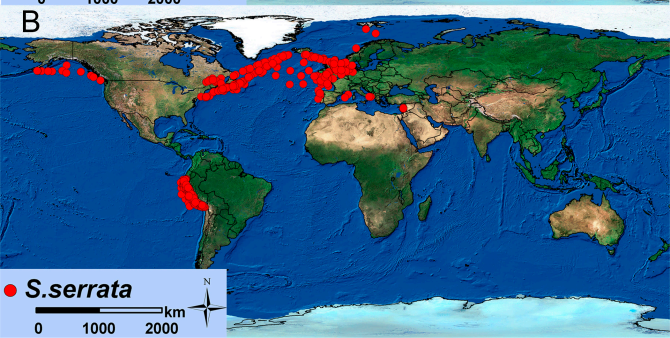
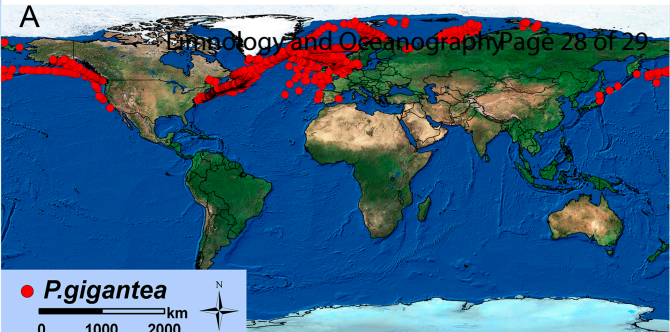
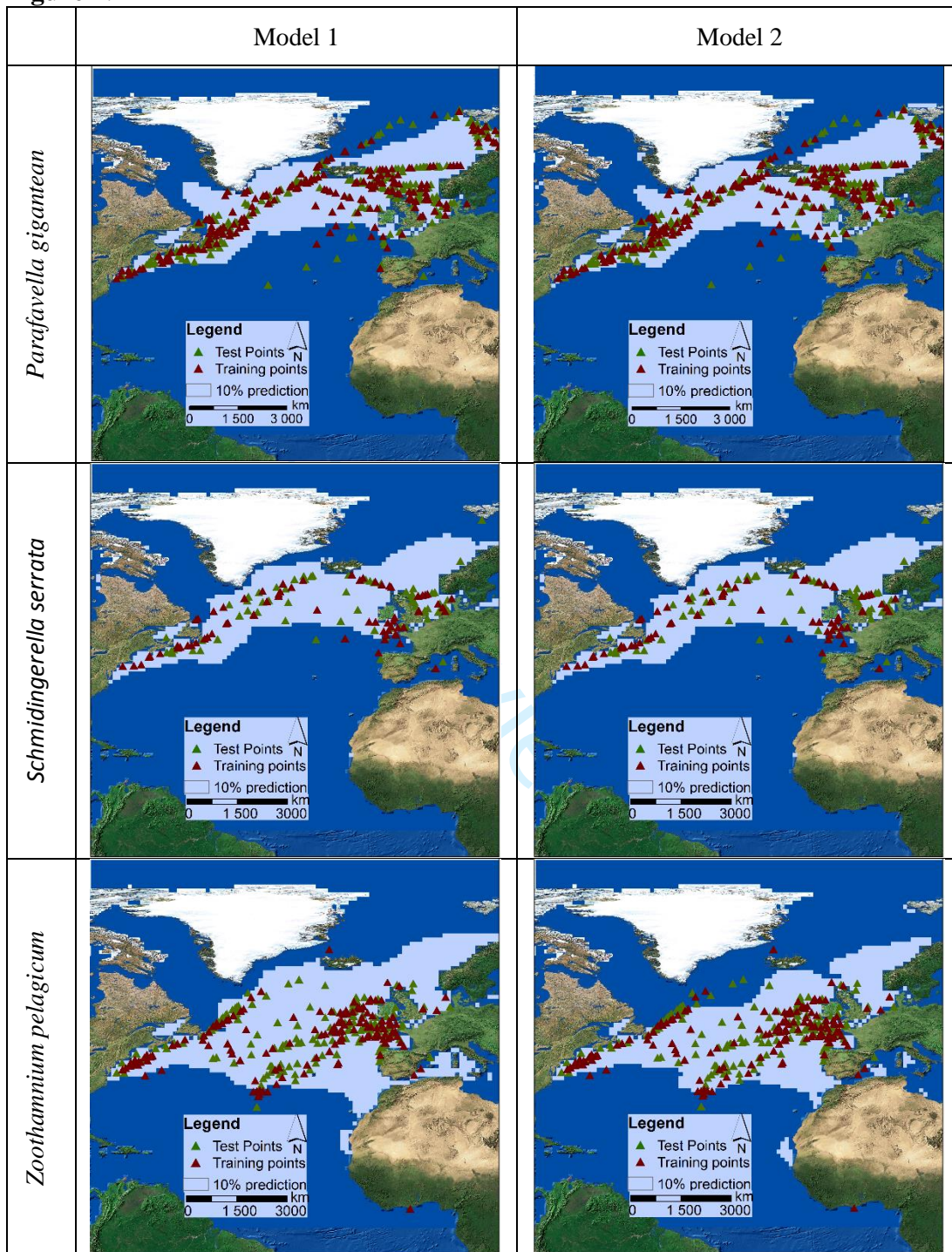


Figure 2.



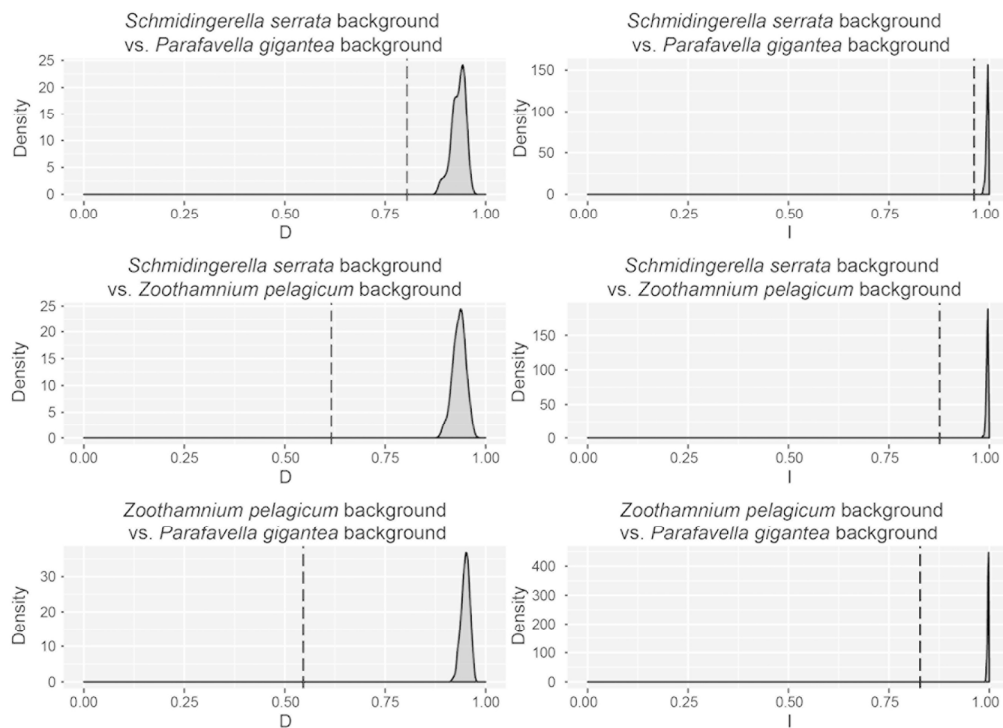


Figure 3. Model 1 background similarity tests comparing the niche occupied by each taxon using the background similarity tests available in the ENMTools R package version 0.1. Schoener's D and a modified Hellinger's I metric, are used to quantify the similarity of two probability distributions. In each case, the niche occupied by each taxon is more different than expected from the study region. Model 2 background similarity tests are shown in Supplementary Material Figure 14.