1	Global patterns and rates of habitat transitions across the eukaryotic tree of life
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3	Mahwash Jamy <sup>1</sup> , Charlie Biwer <sup>1</sup> , Daniel Vaulot <sup>2,3</sup> , Aleix Obiol <sup>4</sup> , Homgmei Jing <sup>5</sup> , Sari
4	Peura <sup>6,7</sup> , Ramon Massana <sup>4</sup> , Fabien Burki <sup>1,7*</sup>
5	
6	<sup>1</sup> Department of Organismal Biology (Systematic Biology), Uppsala University, Uppsala, Sweden
7	<sup>2</sup> Sorbonne Université, CNRS, UMR7144, Team ECOMAP, Station Biologique, Roscoff, France
8	<sup>3</sup> Asian School of the Environment, Nanyang Technological University, Singapore
9	<sup>4</sup> Department of Marine Biology and Oceanography, Institut de Ciències del Mar (ICM-CSIC), Barcelona, Spain
10	<sup>5</sup> CAS Key Lab for Experimental Study Under Deep-sea Extreme Conditions, Institute of Deep-sea Science and
11	Engineering, Chinese Academy of Sciences, Sanya, China
12	<sup>6</sup> Department of Ecology and Genetics (Limnology), Uppsala University, Uppsala, Sweden
13	<sup>7</sup> Science for Life Laboratory, Uppsala University, Sweden
14	
15	*Corresponding author: <u>fabien.burki@ebc.uu.se</u>
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17	Abstract
18	The successful colonisation of new habitats has played a fundamental role during the evolution of life.
19	Salinity is one of the strongest barriers for organisms to cross, which has resulted in the evolution of
20	distinct marine and terrestrial (including both freshwater and soil) communities. Although microbes
21	represent by far the vast majority of eukaryote diversity, the role of the salt barrier in shaping the
22	diversity across the eukaryotic tree is poorly known. Traditional views suggest rare and ancient
23	marine-terrestrial transitions, but this view is being challenged by the discovery of several recently
24	transitioned lineages. Here, we investigate habitat evolution across the tree of eukaryotes using a
25	unique set of taxon-rich environmental phylogenies inferred from a combination of long-read and
26	short-read metabarcoding data spanning the ribosomal DNA operon. Our results show that overall

- 27 marine and terrestrial microbial communities are phylogenetically distinct, but transitions have
- 28 occurred in both directions in almost all major eukaryotic lineages, with at least 350 transition events
- 29 detected. Some groups have experienced relatively high rates of transitions, most notably fungi for
- 30 which crossing the salt barrier has most likely been an important aspect of their successful
- 31 diversification. At the deepest phylogenetic levels, ancestral habitat reconstruction analyses suggest
- 32 that eukaryotes may have first evolved in non-saline habitats, and that the two largest known
- 33 eukaryotic assemblages (TSAR and Amorphea) arose in different habitats. Overall, our findings
- 34 indicate that crossing the salt barrier has played an important role in eukaryotic evolution by providing
- 35 new ecological niches to fill.

### 36 Main text

37 Adapting to new environments with very different physicochemical properties represent large 38 evolutionary steps. When successful, habitat transitions can be important drivers of evolution and trigger radiations<sup>1-4</sup>. The marine-terrestrial boundary (here terrestrial encompassing both freshwater 39 40 and soil<sup>5,6</sup>)—the so-called salt barrier—is considered one of the most difficult barriers to cross, 41 because salinity preference is a complex trait that requires the evolution of multi-gene pathways for physiological adaptations<sup>7-10</sup>. These adaptations have been best studied in macroorganisms, for which 42 the recorded marine-terrestrial transitions are few<sup>11-13</sup>. Microbes (prokaryotic and eukaryotic) are also 43 44 typically regarded as infrequently crossing the salt barrier in spite of much larger population sizes and 45 high dispersal ability<sup>12,14</sup>, but the role of the salt barrier as an evolutionary driver of microbial diversity 46 remains poorly understood. For bacteria, higher habitat transition rates than anticipated have been 47 reported<sup>15</sup>. For microbial eukaryotes, which represent the vast majority of eukaryotic diversity, no data 48 exist to infer the global patterns and rates of habitat transitions at a broad phylogenetic scale. Extant 49 marine and terrestrial eukaryotic communities are distinct in terms of composition and abundance of 50 taxa<sup>6,16</sup>, a pattern that has been attributed to rare and ancient transitions between marine and terrestrial environments<sup>14,17-22</sup>. However, increasing inferences of recent transitions in specific clades such as 51 52 dinoflagellates suggest that the strength of the salt barrier might not be as strong as previously 53 envisioned<sup>23-26</sup>.

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55 In this study, we used a unique hybrid approach combining high-throughput long-read and short-read 56 environmental sequencing to infer habitat evolution across the eukaryotic tree of life. We newly 57 generated over 10 million long environmental reads (ca. 4500 bp of the ribosomal DNA operon) from 58 21 samples spanning marine (including euphotic and aphotic ocean zones), freshwater, and soil 59 habitats. The increased phylogenetic signal of long-reads allowed us to establish, together with a set of 60 phylogenomics constraints, a broad evolutionary framework for the environmental diversity of 61 eukaryotes. We then incorporated existing, massive short-read data (~234 million reads) from a 62 multitude of locations around the world to complement the taxonomic and habitat diversity of our 63 dataset. With this combined dataset, we inferred the frequency, direction, and relative timing of

64	marine-terrestrial transitions during the evolution of eukaryotes; we investigated which eukaryotic
65	lineages are more adept at crossing the salt barrier; and finally, we reconstructed the most likely
66	ancestral habitats throughout eukaryote evolution, from the root of the tree to the origin of all major
67	eukaryotic lineages. Our analyses represent the most comprehensive attempt to leverage
68	environmental sequencing to infer the evolutionary history of habitat transitions across eukaryotes.
69	
70	Results
71	Long-read metabarcoding to obtain a comprehensive environmental phylogeny
72	A range of samples collected globally from marine and terrestrial habitats were deeply sequenced with
73	PacBio (Sequel II) to obtain a comprehensive long-read metabarcoding dataset spanning the broad
74	phylogenetic diversity of eukaryotes. These samples covered all major ecosystems, including the
75	marine euphotic and aphotic zones (surface/deep chlorophyll maximum, and
76	mesopelagic/bathypelagic, respectively), freshwater lakes and ponds as well as tropical and boreal
77	forest soils (see Supplementary Table 1 for details). In total, we obtained 10.7 million Circular
78	Consensus Sequence (CCS) reads spanning ~4500 bp of the ribosomal DNA (rDNA) operon, from the
79	18S to the 28S rDNA genes. After processing, sequences were clustered into Operational Taxonomic
80	Units (OTUs) within each sample at 97% similarity, resulting in 16,821 high-quality OTUs. To assess
81	the potential biases of long-read amplicon sequencing, we performed a direct comparison with
82	Illumina data (for the V4 and V9 hypervariable regions of the rDNA gene, and 18S reads extracted
83	from metagenomic data) previously obtained for the same DNA from three marine samples <sup>27</sup> . This
84	comparison revealed that our long-range PCR assay followed by PacBio sequencing retrieved similar
85	eukaryotic community snapshots, with most groups detected at comparable abundances
86	(Supplementary Figures 1-2). Additionally, the PacBio datasets detected several taxonomic groups
87	that are absent from the V4 and V9 datasets. Importantly, over 80% of the V4 sequences were
88	identical to the PacBio OTUs, indicating that our protocol for CCS processing generates high-fidelity
89	data comparable to classical short-read metabarcoding (Supplementary Figure 1).

90 We then used a phylogeny-aware method to label all OTUs with appropriate taxonomic information<sup>28</sup> 91 (see Materials and Methods for details), and reconstructed a global eukaryotic phylogeny of 92 environmental diversity based on the 18S-28S rDNA genes (Figure 1). In order to allow for transition 93 rates to be estimated within a guiding taxonomic framework (see below), the major eukaryotic groups 94 shown in Figure 1a were constrained to be monophyletic based on established relationships derived 95 from phylogenomic inferences (reviewed in <sup>29</sup>). These major lineages were defined as rank 4 in the 96 taxonomic scheme of an in-house database derived from the protist ribosomal reference (PR2) 97 database<sup>30</sup> called *PR2-transitions*<sup>31</sup>. This phylogeny contains almost all known major eukaryotic 98 lineages (Figure 1); most of the missing groups (e.g. kelp and seaweed) represent large multicellular 99 organisms, or protists found in specific environments not sampled here (e.g. anoxic environments, see 100 Supplementary Table 2). We also uncovered a proportion of novel diversity, i.e. OTUs highly 101 dissimilar to reference sequences that are typically difficult to confidently assign to taxonomic groups. 102 Long-read metabarcoding alleviates the issue of taxonomic assignment of highly diverging sequences, 103 for example we found 863 sequences with <85% similarity to references in PR2 which were attributed 104 a taxonomy based on their position in the tree, mostly belonging to apicomplexan parasites, fungi, and 105 amoebozoans (Figure 1a and Supplementary Figure 3).

## 106 Detection of a salty divide in microbial eukaryotes

107 The global phylogeny in Figure 1 allows to visualize habitat preferences across the eukaryotic tree of 108 life. Overall, we observed a clear phylogenetic distinction between marine and terrestrial lineages, 109 with almost no OTU overlap between these two communities (Figure 1b-c; Unifrac distance = 0.959, 110 p-value < 0.001). Within the marine and terrestrial biomes, soil and freshwater communities were 111 found to be more distinct from each other (Unifrac distance = 0.76, p-value < 0.001) than the marine 112 euphotic and aphotic communities (Unifrac distance = 0.64, p-value < 0.001) (Figure 1b and 113 Supplementary Figure 4). However, we detected several sequences with high identity (>97% similar) 114 present in the marine euphotic and aphotic samples (854 OTUs), and in the soil and freshwater 115 samples (771 OTUs), suggesting that some taxa may be generalists in these sub-habitats (Figure 1c 116 and Supplementary Figure 5).

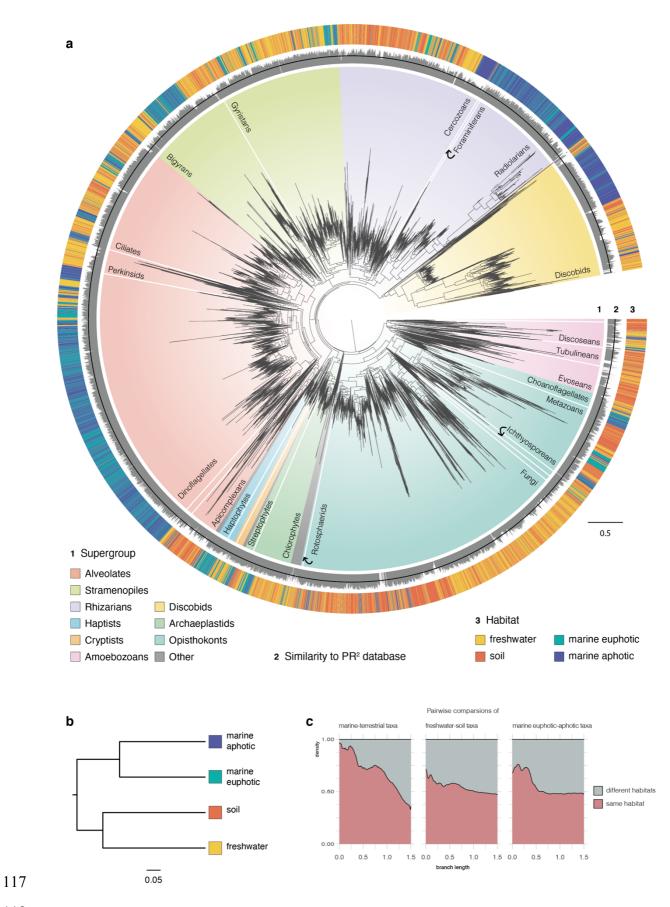


Figure 1. Global eukaryotic 18S-28S phylogeny from environmental samples and the distribution of habitats. (a)
This tree corresponds to the best maximum-likelihood (ML) tree inferred using an alignment with 7,160 sites

120	and the GTRCAT model in RAxML <sup>32</sup> . The tree contains 16,821 OTUs generated from PacBio sequencing of 21
121	environmental samples. The innermost ring around the tree indicates taxonomy, and the major eukaryotic
122	lineages considered in this study are labelled. The second ring depicts percentage similarity with the references
123	in the PR2 database and was set with a minimum of 70 and a maximum of 100, with the black line in the middle
124	indicating 85% similarity. The third ring depicts which habitat each OTU belongs to. (b) Hierarchical clustering
125	of the four habitats based on a phylogenetic distance matrix generated using the unweighted UniFrac method. (c)
126	Stacked density plot of branch lengths between taxa pairs from the same or different habitats. Note that this plot
127	should be interpreted with caution as each taxa-pair does not represent independent data-points due to
128	phylogenetic relatedness.

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130 We next sought to increase the number of samples and covered diversity by taking advantage of the 131 mass of available short-read metabarcoding datasets. We gathered data from 22 studies conducted 132 globally (including marine and terrestrial ecosystems), amounting to 234 million reads in total after 133 processing (Supplementary Figure 6, Supplementary Table 3). We opted to use only the V4 region (ca. 134 260 bp) of the 18S rDNA gene as it was shown to have a greater phylogenetic signal than the V9 135 region<sup>33</sup>. The V4 reads were clustered into OTUs at 97% similarity for the marine euphotic (9977 136 OTUs), marine aphotic (2518 OTUs), freshwater (3788 OTUs), and soil (11935 OTUs) environments 137 (Supplementary Table 4). These short-read OTUs were then phylogenetically placed onto the global long-read-based eukarvotic phylogeny using the Evolutionary Placement Algorithm (EPA)<sup>34</sup> 138 139 (Supplementary Figure 7), for which we compared the placement distributions for each sub-habitat. 140 Interestingly, most placements occurred close to the tips of the reference tree, indicating that our long-141 read dataset adequately represents the diversity recovered by short-read metabarcoding 142 (Supplementary Figure 7). Furthermore, the placement distributions for each habitat is consistent with 143 our results based on the long-reads only, namely that marine and terrestrial communities are distinct, 144 and at a finer level, soil and freshwater communities are more different from each other than 145 communities in the surface and deep ocean (soil-freshwater earth mover's distance = 1.14, marine 146 euphotic-aphotic earth mover's distance = 0.809; Supplementary Figure 8). 147

## 148 Marine-terrestrial transition rates vary across major eukaryotic clades

149 The above results confirm that the salt barrier leads to phylogenetically distinct eukaryotic 150 communities. We next asked how often have transitions between marine and terrestrial habitats 151 occurred during evolution, which eukaryotic lineages have crossed this barrier more frequently, and in 152 which direction? To answer these questions, we calculated habitat transition rates across the global 153 eukaryotic phylogeny by performing Bayesian ancestral state reconstructions using continuous-time 154 markov models<sup>35</sup>. We tested a null model, where transition rates from marine to terrestrial habitats 155 (qMT) and vice versa (qTM) are constant throughout the eukaryotic phylogeny, against a 156 heterogeneous model where qMT and qTM are estimated separately for each major eukaryotic lineage 157 (illustrated in Figure 1). The null model had a posterior density of log-likelihoods with a mean of -158 2008.45 (Supplementary Figure 9). Under this model, transitions from marine to terrestrial habitats are 159 just as likely as the reverse across the tree. However, this general analysis hides important variations 160 in habitat transition rates between groups, and indeed the heterogenous model presented a much better 161 fit (log-likelihood score of -1819.91; Log Bayes Factor = 269.3; Supplementary Figure 9), indicating 162 that habitat transition rates vary strongly across the tree.

163 To investigate in more detail the rate of habitat transition within each major eukaryotic group, we 164 inferred taxon-rich clade-specific phylogenies by combining short-read data with the backbone 165 phylogenies obtained from long-read data. Incorporating these short-read data allowed us to detect 166 additional transition events that would have otherwise been missed with the long-read data alone 167 (Supplementary Figure 10). We modelled habitat transition rates along clade-specific phylogenies 168 containing both marine and terrestrial taxa that were sufficiently large (at least 50 tips) to get precise 169 estimates. We also excluded discobid excavates and discosean amoebozoans as preliminary analyses 170 showed ambiguous transition rate estimates owing to large phylogenetic uncertainty. Fungi were 171 found to have by far the highest transition rates for a given amount of evolutionary change; we 172 estimated around 90 expected transition events along a branch length of one substitution/site in the 173 phylogeny. These results indicate that habitat shifts are associated with very little evolutionary change 174 in the ribosomal DNA sequences (Figure 2a). After fungi, cryptophytes and gyristans (ochrophyte

- algae, oomycete parasites and several free-living flagellates) had the highest global rates (around 8.2
- and 3.4 and expected transitions per substitution per site).

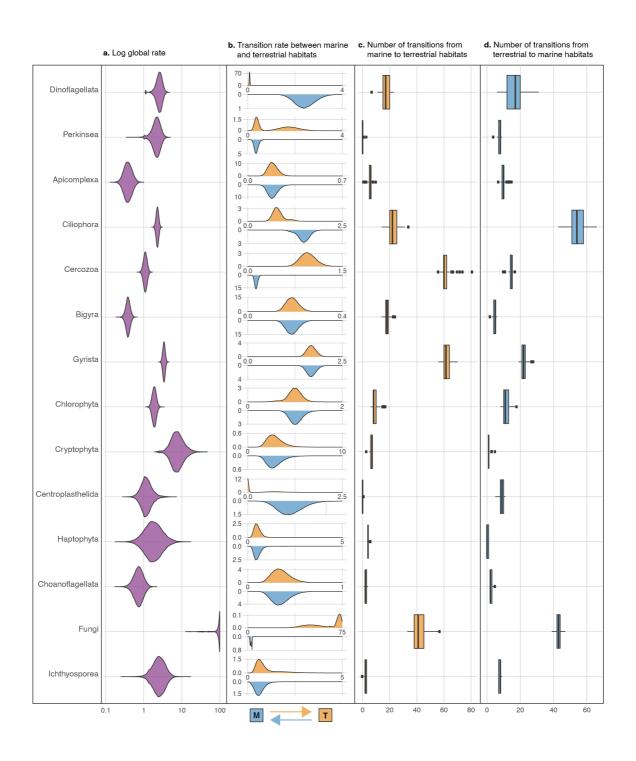


Figure 2. Habitat transition rates and number of transition events estimated for each major eukaryotic lineage.
(a) Posterior probability distributions of the global rate of habitat evolution, which indicate the overall speed at
which transitions between marine and terrestrial habitats have occurred in each clade regardless of direction.
Rates were estimated along clade-specific phylogenies (see Supplementary Figure 10) using Markov Chain

Monte Carlo (MCMC) in BayesTraits using a normalized transition matrix. (b) The posterior probability
distribution of transition rates from marine to terrestrial habitats (top in orange), and from terrestrial to marine
habitats (below in blue). (c) Number of transitions from marine to terrestrial habitats and (d) in the reverse
direction for each clade as estimated by PASTML using Maximum Likelihood (see Materials and Methods for
details).

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188 At a finer phylogenetic resolution, several subclades within stramenopiles, as well as ciliates, seemed 189 particularly adept at crossing the salt barrier, especially chrysophytes, diatoms, and spirotrich ciliates 190 (11.8, 8.7, and 3.8 expected transition events per substitution per site respectively; Supplementary 191 Figure 11-12). At the other extreme, groups such as bigyrans (heterotrophic stramenopiles related to gyristans) and apicomplexans (a group of parasites including the malaria pathogen) displayed the 192 193 lowest habitat transition rates (around 0.4 expected transitions for every substitution per site). These 194 results were further confirmed with sequence similarity network analyses, which showed high 195 assortativity between marine and terrestrial sequences for bigyrans and apicomplexans (meaning that 196 terrestrial and marine sequences formed distinct clusters at varying similarity thresholds), as opposed 197 to gyristans and fungi, which showed low assortativity (Supplementary Figure 13).

198 Within each major eukaryotic group, we next inferred the frequency for each direction of the 199 transitions between marine to terrestrial habitats. We found that all clades investigated had non-null 200 transition rates in both directions, with the exception of centrohelids which had a terrestrial 201 colonization rate that was not significantly different from zero in 99/100 trees used for calculation 202 (Figure 2b). These results indicate that in nearly all major eukaryotic lineages containing terrestrial 203 and marine taxa, transitions have occurred in both directions. Some clades had symmetrical transition 204 rates, indicating that the tendency to colonize marine environments was not significantly different 205 from the tendency to colonize terrestrial environments; this was for example the case of 206 apicomplexans, bigyrans, gyristans, chlorophytes, cryptophytes, haptophytes, and choanoflagellates 207 (Figure 2b). However, some groups showed marked differences in one direction or the other. 208 Dinoflagellates, for example, show a much greater transition rate for colonizing marine habitats (about

209 31 times more likely). Ciliates have also transitioned more frequently towards marine environments, 210 but the difference is smaller (1.8 times more likely). On the other hand, transitions to terrestrial 211 environments were significantly more likely than the reverse direction for fungi and cercozoans (about 212 21.5 and 7.2 times more likely, respectively). Finally, the directionality of habitat transition appears to 213 be heterogeneous also within the major eukaryotic groups (Supplementary Figures 14-17). Indeed, for 214 some selected subclades such as ascomycetes and basidiomycetes within fungi, the transition rates to 215 marine environments were higher as compared to non-Dikarya fungi, although fungi as a whole 216 showed a marked tendency to colonize terrestrial habitats (qTM = 8.47 vs. 1.65 respectively; 217 Supplementary Figure 17).

218 Finally, we estimated the number of transition events within each clade by generating discrete habitat 219 histories using a maximum likelihood method<sup>36</sup>. We conservatively counted transition events only if 220 they led to a clade with at least two taxa in the new habitat in order to distinguish between biologically 221 active, speciating residents from wind-blown cells, resting spores or extracellular DNA from dead 222 cells<sup>37</sup>. Our analyses revealed at least 350 transition events occurring over eukaryotic history, though 223 the actual number is likely to be higher when considering lineages that have gone extinct. Out of these, 224 72 or more transition events occurred in fungi alone (39-47 transitions to marine environments 225 detected, and 33-57 transitions to terrestrial environments detected) (Figure 2c-d). This was closely 226 followed by gyristans and ciliates, with more than 60 putative switches each between environments 227 (Figure 2c-d).

## 228 Relative timing of habitat transitions during the evolution of the major eukaryotic groups

We next asked when during eukaryote evolution these transitions between marine and terrestrial habitats occurred. To calculate a relative timing for all marine-terrestrial transitions, we converted the clade-specific phylogenies into chronograms with relative dates (as in <sup>38</sup>).

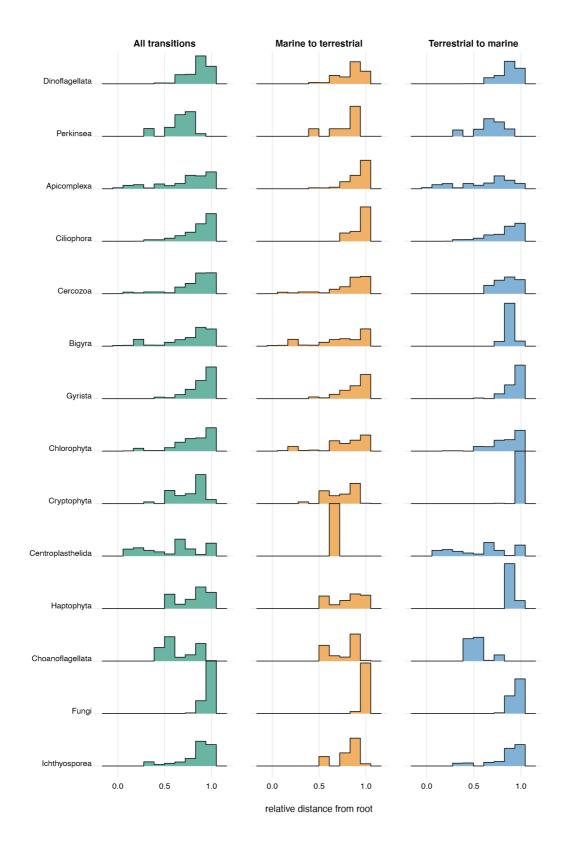


Figure 3. Ridgeline histogram plots displaying the timing of transition events as estimated from relative
 chronograms obtained with Pathd8<sup>38</sup>. The x-axis depicts the relative age for each clade.

235 For each putative transition event, we measured the relative branch length from the inferred transition 236 to the root of the clade. The general trend is that most transitions occurred relatively recently in the 237 history of the groups (Figure 3). For instance, we detected no transition events in fungi older than 25% 238 of the clade's history, with the vast majority of all transitions occurring in the last 10% of the time that 239 this group has been on earth. Assuming that fungi arose around 1 billion years ago<sup>39-41</sup>, this would 240 imply that > 90% of all marine-terrestrial transitions (at least 63 transitions according to our analyses) 241 in fungi occurred in the last 100 million years alone, with older transitions occurring predominantly 242 towards marine environments. The observation that most transitions occurred towards present could be 243 due to the increased challenges of inferring transition events early in the evolution of a group because 244 of poorer resolution of deeper nodes due to little phylogenetic signal, and/or unsuccessful transitions 245 leading to lineage extinctions in the new habitat. However, for a few clades at least (centrohelids, 246 bigyra, apicomplexans, cercozoans, and chlorophytes), we detected a number of early transitions in the 247 evolution of the group (Figure 3). Interestingly, the direction of these early habitat transitions is non-248 overlapping. For centrohelids and apicomplexans, the early transitions were mainly towards marine 249 environments, possibly corresponding to repeated marine colonization events at the onset of the 250 groups' evolution. Early terrestrial colonization events were instead detected in cercozoans, 251 chlorophytes, and bigyrans, together suggesting that early in the evolution of the major eukaryotic 252 groups the pressure to move towards marine or terrestrial habitats was group-specific and directional.

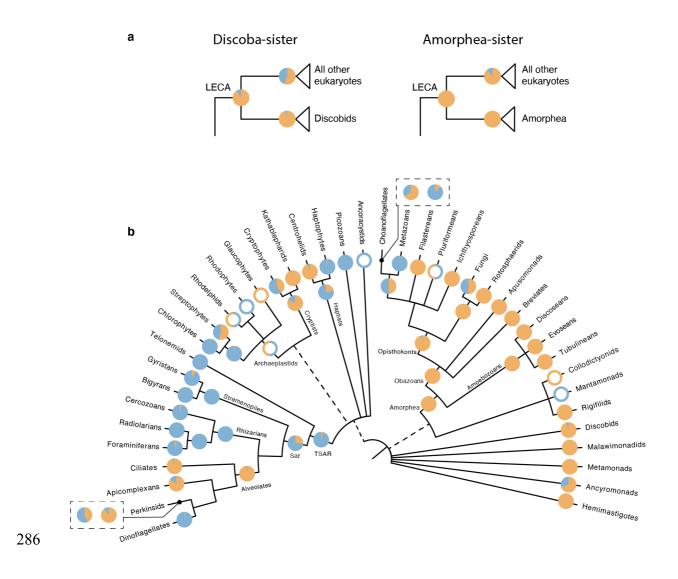
# 253 Ancestral habitat reconstruction of the major eukaryotic clades

254 Our global eukaryotic phylogeny of long-environmental OTUs, combined with the group-specific 255 phylogenies including short-read metabarcoding data, represent a very dense set of environmental 256 information put in a phylogenetic framework. We used this information to reconstruct in a Bayesian 257 analysis the most likely ancestral environments from the root of the eukaryotic tree through the 258 emergence of the major groups. Inferring the ancestral habitat of the last eukaryotic common ancestor 259 (LECA) requires information about the root itself, which remains very contentious $^{29,42}$ . To 260 accommodate uncertainties for the position of the root, we performed ancestral habitat reconstruction 261 analyses using the two most commonly proposed root positions: (1) between the discobid excavates

and all other eukaryotes<sup>43</sup>, and (2) between amorpheans (the group including animals, fungi and
 amoebozoans) and all other eukaryotes<sup>44</sup>. Both root alternatives converged towards the same habitats,
 suggesting with high confidence that LECA evolved in a terrestrial environment (Figure 4a).

265 From this inferred terrestrial root, our analyses suggest that two of the largest mega-assemblages of eukaryotes, likely comprising more than half of all eukaryotic diversity<sup>45</sup>, arose in different 266 267 environments. On one hand, the amorphean group likely originated in a terrestrial habitat (Figure 4b), 268 where it initially diversified into obazoans (which include well-known lineages such as animals and 269 fungi, but also several unicellular related lineages), as well as the amoebozoans. Consistent with previous studies, we inferred a marine origin for metazoans<sup>46,47</sup>, however for two obazoan lineages-270 271 fungi and the group containing metazoans and choanoflagellates-we could not determine a clear 272 preference for their ancestral habitats. On the other hand, our analyses indicate that the expansive 273 TSAR clade (containing the main eukaryotic phyla stramenopiles, alveoates, and rhiziarians, as well as 274 the smaller group telonemids) most likely originated in a marine environment, following the transition 275 of an ancestral population from a terrestrial root (Figure 4b). A marine origin is also likely for the 276 major TSAR members, except for alveolates which were inferred to have a terrestrial origin.

277 Overall, the predicted ancestral habitats of most major eukaryotic clades match their current preferred 278 habitat: this is for example the case for all amoebozoan lineages, radiolarians, dinoflagellates and 279 foraminiferans. An exception is cercozoans for which a marine origin was inferred, but which now dominate terrestrial environments, particularly soils<sup>6,48</sup>. Interestingly, the results derived from the 280 281 global eukaryotic phylogeny and the clade-specific phylogenies (which include short-read OTUs) 282 were largely consistent except in two cases: the phylogeny of perkinsids changed the origin from 283 terrestrial to marine for these parasites of animals, while the phylogeny of choanoflagellates switched 284 from a marine to a terrestrial origin (Figure 4b).



287 Figure 4. Ancestral states of major eukarvotic clades as estimated by BayesTraits on a set of 100 global PacBio 288 phylogenies. Pie charts at each node indicate the posterior probabilities of likelihoods for the character states as 289 follows: blue=marine, orange=terrestrial. Nodes with empty circles indicate wherever there was insufficient 290 taxon sampling to infer ancestral habitats, but a reasonable estimate was made from existing literature (See 291 Supplementary Note 1). (a) Ancestral habitat of the last eukaryotic common ancestor (LECA) as inferred using 292 two different roots. (b) Ancestral states of major eukaryotic lineages. For the two cases where the incorporation 293 of Illumina data inferred a different likely ancestral state, the results are shown in boxes. The pie chart on the 294 right was obtained using the global eukaryotic phylogeny, while the pie chart on the left was obtained from 295 clade-specific phylogenies. The tree is adapted from Burki et al<sup>49</sup>.

## 296 Discussion

297 In this study, we use a unique combination long- and short-read data to obtain an evolutionary 298 framework of environmental diversity and infer habitat preference evolution across the eukaryotic 299 tree. High-throughput long-amplicon sequencing followed by careful processing of the data provide 300 high-quality sequences containing improved phylogenetic signal for the vast environmental diversity<sup>28,50-52</sup>. We generated over 10 million long-read metabarcoding data spanning the eukaryotic 301 302 rDNA operon, which assembled into nearly 17,000 OTUs, for marine and terrestrial ecosystems. We 303 then added two additional layers of phylogenetic information: (i) a much larger mass of available 304 short-read metabarcoding data to more deeply cover the molecular diversity of environmental 305 microbes, and (ii) a set of well-accepted constraints derived from published phylogenomic analyses to 306 fix the backbone of our eukaryotic tree. By combining all this information, we show that we can infer 307 evolutionary patterns at global scales across the tree.

308 We confirm that the salt barrier has been a major factor in shaping eukaryotic diversity<sup>6,14,16</sup>, and that 309 marine-terrestrial transitions are infrequent in comparison to transitions across other habitats such as 310 between freshwater and soil (Figure 1). Our analyses detected at least 350 transition events (Figure 2), 311 although this number is likely to be higher when considering more ancient transitions that were not 312 detected, and more recent transitions that will only be revealed by sequencing more locations (for e.g. 313 in  $^{24}$ ). These difficult-to-achieve environmental crossings have likely played important evolutionary 314 roles by allowing colonizers to reach vacant ecological niches. For example, crossing the salt barrier 315 may have led to the establishment of some major eukaryotic assemblages such as TSAR, or highly 316 diverse lineages such as the oomycetes and vampyrellids (Supplementary Figure 12). Marine-317 terrestrial transitions have also allowed lineages such as diatoms, golden algae and spirotrich ciliates 318 to expand their range to both habitats, contributing to the diversification of the vast eukaryotic 319 diversity we see today. We unexpectedly found that 56% of the detected transitions occurred recently, 320 in the last 10% of the evolutionary history of the respective groups (Figure 3), which is in contrast to a common idea that most marine-terrestrial transitions are ancient<sup>14</sup>. It is however unclear why 321 322 colonization across the salt barrier would be more frequent in recent geological time, so this 323 observation could instead be due to recent colonizing lineages having had less time to go extinct, and

324 thus more likely to be represented in our generated phylogenies<sup>53</sup>. At its deepest phylogenetic level, 325 our analyses suggest that the earliest eukaryotes inhabited terrestrial habitats (Figure 4) and not marine 326 habitats as often assumed (e.g.<sup>54-56</sup>). While the fossil record for early eukaryotes is sparse and difficult 327 to distinguish from prokaryotes, there is evidence for early eukaryotes in non-marine or low salinity environments from at least 1 Gyr ago<sup>55</sup>. Furthermore, other key early eukaryotic innovations, such as 328 329 the origin of the plastid organelles, have been inferred to have occurred around 2 Gyr ago in lowsalinity habitats<sup>57,58</sup>. Terrestrial environments are known to be more heterogenous<sup>59</sup>, and may thus 330 331 have provided a wide range of ecological niches for early eukaryotes to occupy.

332 Our detailed investigation across the main groups of eukaryotes showed marked differences in the 333 rates of crossing the salt barrier (200-fold globally). While some groups have low global transition 334 rates, others show a higher tendency to cross this physiological barrier. Most notably, we inferred 335 based on both the highest transition rates in our analysis and relatively high number of transition 336 events (Figure 2), that fungi are the strongest eukaryotic colonizers between marine and terrestrial 337 environments. This is consistent with previous studies documenting a multitude of close evolutionary 338 associations between marine and terrestrial fungal lineages <sup>60–62</sup>, which in turn suggests that many fungal species may be generalists that can tolerate a wide range of salinities $^{63,64}$ . Interestingly, fungi 339 340 showed a much greater trend (21-fold) for colonizing terrestrial environments, where they are 341 dominant, than the reverse. Whether this reflects a strong preference for terrestrial environments, or 342 instead unequal diversification rates in the two habitats<sup>65,66</sup>, or both, is unclear and should be further 343 investigated.

The differences in habitat transition rates across eukaryotes are likely the result of varying salinity tolerance that has prevented successful colonization events during the evolution in some groups. Among algae, comparative genomics showed large differences in gene content between marine and freshwater species, notably for ion transporters and other membrane proteins that likely play important roles in osmoregulation<sup>67</sup>. These different gene contents may be due, at least in part, to laterally acquired genes (LGT) that could facilitate successful crossing of the salt barrier, as proposed for other environmental adaptations<sup>68–72</sup>. In bacteria, it was hypothesized that particle-associated species can

351 more easily cross the salt-barrier due to increased chances of acquiring osmoregulation-related genes 352 through LGT<sup>73</sup>. Altogether, these observations raise the question of how protists in general acquire 353 these genes, for example groups like diatoms (Supplementary Figures 11-12) which showed multiple 354 transitions in both directions, and whether it is through LGT (as has been shown for some halophilic 355 protists<sup>68</sup>), gene duplication, or through re-wiring of existing metabolic pathways (as shown for the 356 SAR11 bacteria<sup>74</sup>). Other ecological factors also likely play a role as a colonizing organism does not 357 only need to adapt to a different salinity, but also has to adapt to the different nutrient and ion 358 availabilities, and avoid being out-competed or preyed on by the resident community<sup>73</sup>.

# 359 Conclusions

360 This study represents the first comprehensive analysis of the evolution of saline and non-saline habitat 361 preferences across the global tree of eukaryotes. We inferred that two of the largest assemblages of 362 eukaryotes (TSAR and Amorphea) originated in different environments, and that ancestral eukaryotes 363 likely inhabited non-marine environments. Our results show that marine and non-marine communities 364 are phylogenetically distinct, but the salt barrier has been crossed several hundred times over the 365 course of eukaryotic evolution. Several of these crossings coincided with the birth of diverse lineages, 366 indicating that the availability of new niches has likely played a large role in the vast eukaryotic 367 diversity we see today. We predict that the generation of genomic data from closely related marine and 368 non-marine lineages will shed light on the genetic and cellular adaptations that have allowed crossings 369 over the salt barrier.

### 370 Methods

## 371 Environmental samples for long-read metabarcoding and total DNA extraction

A total of 18 samples were sequenced for this study: five freshwater samples, four soil samples, four marine euphotic samples, and five marine aphotic samples (see Supplementary Table 1 for sample coordinates and details). Additionally, we used reads from three soil samples that were sequenced in a previous study<sup>28</sup> (ENA accession PRJEB25197), resulting in a total of 21 samples that were analysed in this study. The aim here was to get a representative view of the microbial eukaryotic diversity in each environment.

378

# 379 <u>Soil samples (x4 samples)</u>

Peat samples were collected from (1) Skogaryd mire and (2) Kallkäls mire in October-November
2019. 5ml samples with three to four replicates of the top layer of soil were collected at both sites and
visible roots were removed. Samples were kept at 4°C for two days before extracting DNA using the
DNeasy PowerSoil Kit (Qiagen). We also obtained DNA extracted from: (3) rainforest soil samples
(six sites) from Puerto Rico <sup>75</sup>, and (4) boreal forest soil samples (six sites) from Sweden<sup>76</sup>.

# 386 <u>Freshwater samples (x5 samples)</u>

We sampled three freshwater lakes in Sweden in October-November 2019: (1) Lake Erken, (2) Lake 387 388 Ersjön, and (3) Lake Stortjärn. Plantonic samples were collected from the middle of the lakes at 389 multiple depths, and mixed. Up to 3L of water was pre-filtered through a 200 µm mesh net to remove 390 larger organisms before sequentially filtering through 20-25 µm, 3 µm, and 0.25 µm polycarbonate 391 filters (47 mm). Filters were immediately frozen at -20°C and stored at -70°C before further 392 processing. We also collected a (4) freshwater sediment sample (four replicates) from Lake Erken. The 393 upper 0-5 cm of a sediment core was separated and mixed. All samples were kept at 4°C before 394 processing and extracting DNA using the DNeasy PowerSoil Kit. Lastly, we obtained DNA from (5) 395 10 permafrost thaw ponds in Canada<sup>77</sup>.

396

# 397 <u>Marine euphotic samples (x4 samples)</u>

398 One 5L sample was collected from the (1) North Sea at a depth of 5 m. Water was processed, and 399 DNA extracted as described for the freshwater water samples. We used DNA extracts from the nano 400 (3-20  $\mu$ m) and pico (0.2-3  $\mu$ m) fractions of two stations from the Malaspina expedition (Stations 49 401 and 76)<sup>78</sup>. These extracts corresponded to one (2) surface sample at 3 m depth, and (3-4) two DCM 402 layer samples at depths of 70 m and 85 m.

# 404 *Marine aphotic samples (x5 samples)*

We used DNA extracts from the nano and pico fractions of the aphotic marine environment from
Malaspina stations 49 and 76<sup>78</sup>. These corresponded to depths of (1-2) 275 m and 800 m for the
mesopelagic, and (3-4) 1200 m and 2800-3300 m for the bathypelagic samples. Lastly, we obtained
(5) DNA from a Mariana Trench sample from a depth of 5900 m<sup>79</sup>.

409

# 410 PCR amplification and long-read sequencing

We amplified a ~4500 bp fragment of the ribosomal DNA operon, spanning the 18S gene, ITS region, and 28S gene, using the general eukaryotic primers 3NDf <sup>80</sup> and 21R<sup>81</sup>. PCRs were performed with sample-specific tagged-primers using the Takara LA Taq polymerase (Takara) and 5 ng of DNA as input. PCR-cycling conditions included an initial denaturation step at 94°C for 5 min, at least 25 cycles of denaturation at 98°C for 10 sec, primer annealing at 60°C for 30 sec, and elongation at 68°C for 5

416 min, and finishing with a final elongation step at 68°C for 10 min. We limited the number of PCR

417 cycles to 25, where possible, to reduce chimera formation<sup>82</sup>. For samples that did not get amplified, we

418 increased the number of cycles to 30. PCR products were assessed using agarose gels and Qubit 2.0

419 (Life Technologies), and then purified with Ampure XP beads (Beckman Coulter). Amplicons from

420 replicates and different sites from the same sampling location were pooled at this stage. SMRTbell

421 libraries were constructed using the HiFi SMRTbell Express Template Prep Kit 2.0. Long-read

422 sequencing was carried out at SciLifeLab (Uppsala, Sweden) on the Sequel II instrument (Pacific

423 Biosciences) on a SMRT Cell 8M Tray (v3), generating four 30-hour movies.

424

### 425 **Processing reads and OTU clustering**

426 We QC filtered sequences following<sup>28</sup> with some modifications. The CCS filtration pipeline is

427 available at <sup>83</sup>. Briefly, Circular Consensus Sequences (CCS) were generated by SMRT Link

428 v8.0.0.79519 with default options. The CCS reads were demultiplexed with mothur v1.39.5<sup>84</sup>, and then

429 filtered with DADA2 v1.14.1<sup>85</sup>. Reads were retained if they had both primers and if the maximum

- 430 number of expected errors was four (roughly translating to one error for every thousand base pairs).
- 431 We pre-clustered reads at 99% similarity using VSEARCH v2.3.4<sup>86</sup>, and generated consensus

432	sequences for pre-clusters $\geq$ 3 reads to denoise the data. Prokaryotic sequences were detected by
433	BLASTing <sup>87</sup> against the SILVA SSU Ref NR 99 database v132 <sup>88</sup> and removed. We predicted 18S and
434	28S sequences in the reads using Barrnap v0.7 (reject 0.4kingdom euk) <sup>89</sup> , and discarded non-
435	specific and artefactual reads (i.e. those containing multiple 18S/28S, or missing 18S/28S). Chimeras
436	were detected <i>de novo</i> using Uchime <sup>90</sup> as implemented in mothur. Finally, we extracted the 18S and
437	28S sequences from the reads and clustered them using VSEARCH into Operational Taxonomic Units
438	(OTUs) at 97% similarity. After discarding singletons, a second round of denovo chimera detection
439	was performed using VSEARCH, and chimeric OTUs were removed. We calculated sequence
440	similarity of the OTUs against reference sequences in a custom PR <sup>2</sup> database <sup>30</sup> ( <i>PR2-transitions</i> <sup>31</sup> ; see
441	below) using VSEARCH (usearch_global andiddef 1). All references and OTU sequences were
442	trimmed with the primers 3ndf and 1510R <sup>91</sup> to ensure that they spanned the same region.
443	
444	Taxonomic annotation of long-read sequences
445	The modified PR2 reference database
446	Reference sequences were derived from a modified version of the Protist Ribosomal Reference (PR2)
447	database v4.12.0 <sup>30</sup> , called <i>PR2_transitions</i> . This database revised the taxonomy structure of PR2 to 9
448	levels: Domain, Supergroup, Division, Subdivision, Class, Order, Family, Genus, Species. This
449	allowed us to update the taxonomy to accommodate recent changes in eukaryotic classification <sup>92</sup>
450	(changes in taxonomy can be viewed at <sup>83</sup> ). Additionally, we added sequences from nucleomorphs, and
451	several newly discovered or sequenced lineages such as Rholphea, Hemismastigophora, and others.
452	<i>PR2_transitions</i> is available on Figshare <sup>31</sup> . We used the 18S gene alone for taxonomic annotation, as
453	28S databases are much less comprehensive by comparison.
454	
455	Phylogeny-aware taxonomy assignment

457 approach assigns taxonomy to the appropriate taxonomic rank, such that OTUs branching deep in the 458 eukaryotic tree are labelled to high taxonomic ranks, and vice versa. For each sample, we inferred

459 preliminary maximum likelihood trees along with SH-like support<sup>93</sup> with RAxML v8<sup>32</sup> (using the

GTRCAT approximation as it is better suited for large trees<sup>94</sup>). These trees contained the filtered
OTUs and closely related reference sequences from *PR2\_transitions*. Trees were scanned manually to
identify mis-annotated reference sequences, nucleomorphs, and artefactual OTUs. After removing
these sequences, we inferred trees with RAxML-NG<sup>95</sup> using 20 starting trees.

The final taxonomy was generated by getting the consensus of two strategies. Strategy 1 parses the tree and propagates taxonomy to the OTUs from the nearest reference sequences using the Genesis<sup>96</sup> app *partial-tree-taxassign*<sup>97</sup>. Strategy 2 starts by pruning the OTUs from the phylogeny, leaving behind references only. OTUs are then phylogenetically placed on the tree with EPA-ng v0.3.5 <sup>34</sup>, and taxonomy assigned using the gappa<sup>96</sup> command *assign* under the module *examine*. The resulting taxonomy of the 18S gene of each OTU was transferred to its 28S gene counterpart, as the molecules are physically linked.

472

#### 473 Maximum likelihood analyses of the global eukaryotic dataset

474 18S and 28S sequences were aligned using MAFFT v7.310<sup>98</sup> using the FFT-NS-2 strategy, and subsequently trimmed with trimAl<sup>99</sup> to remove sites with >95% gaps. We inferred preliminary trees 475 from a concatenated alignment with RAxML v8.2.12 under the GTRCAT model<sup>32</sup> which were then 476 477 visually inspected to detect chimeras and sequence artefacts. Taxa were removed if their position in 478 the tree did not match their taxonomy. Four such rounds of visual inspection were performed, two 479 with unconstrained trees, and two with constrained trees (see text below for details on constraints). To avoid long branch attraction, we excluded rapidly evolving taxa using TreeShrink<sup>100</sup> (k=2500). This 480 481 resulted in the removal of Mesodinium, long-branch Microsporidia, several Apicomplexa, several 482 Heterolobosea, and several Colladaria from our dataset.

483

484 After removing chimeras and sequence artefacts, we realigned and trimmed the 18S and 28S

485 sequences as before. After concatenation, the final dataset was composed of 16,821 taxa and 7,160

- 486 alignment sites. Global eukaryotic phylogenies of the taxonomically annotated, 18S-28S
- 487 environmental sequences were inferred using RAxML v8.2.12 under the GTRCAT model<sup>32</sup>, and 100

transfer bootstrap replicates (TBE)<sup>101</sup>. Supergroups, Divisions, and Subdivisions (ranks 2, 3 and 4 in 488 489 PR2 transitions) were constrained to be monophyletic in our tree (i.e. all taxa labelled as a specific 490 subdivision were constrained to be on one side of a split). The one exception was Excavata whose monophyly has not been confidently resolved<sup>29</sup>. One hundred maximum likelihood inferences were 491 492 performed in order to take phylogenetic uncertainty into account for subsequent ancestral state 493 reconstruction analyses. We opted to include only the long-read environmental sequences in our 494 phylogenies because they better represent environmental diversity (compared to reference databases which are more biased towards culturable organisms and marine environments<sup>102</sup>), and because very 495 496 few 18S-28S sequences can otherwise be ascertained to derive from the same organism. The final tree 497 along with metadata was visualised using the anvi'o interface<sup>103</sup> and then modified in Adobe Illustrator<sup>104</sup> to label clades. 498 499

## 500 Short read datasets

### 501 *Datasets collected*

Short-read data corresponding to the V4 hypervariable region were retrieved from 22 publicly available metabarcoding datasets. Data were considered if the following criteria were fulfilled: (i) samples were collected from soils, freshwater, or marine habitats (ii) there was clear association between samples and environment (i.e. no data from estuaries where salinity fluctuates); and (iii) data publicly available or authors willing to share. The search for studies was not meant to be exhaustive and the datasets included in this work were identified and collected by the end of October 2020, unless specified otherwise. A list of these datasets can be found in Supplementary Table 3.

509

# 510 <u>Processing short-read data and clustering into OTUs</u>

511 Raw sequence files and metadata were downloaded from NCBI SRA web site<sup>105</sup> when available or 512 obtained directly from the investigators. Information about the study and the samples (substrate, size 513 fraction etc.) as well as the available metadata (geographic location, depth, date, temperature etc.) 514 were stored in three distinct tables in a custom MySQL database stored on Google Cloud. For each 515 study, raw sequences files were processed independently de novo. Primer sequences were removed

516	using cutadapt <sup>106</sup> (maximum error rate = $10\%$ ). Amplicon processing was performed under the R
517	software <sup>107</sup> using the dada2 package <sup>85</sup> . Read quality was visualized with the function
518	plotQualityProfile. Reads were filtered using the function filterAndTrim, adapting parameters
519	(truncLen, minLen, truncQ, maxEE) as a function of the overall sequence quality. Merging of the
520	forward and reverse reads was done with the mergePairs function using the default parameters
521	(minOverlap = 12, maxMismatch = 0). Chimeras were removed using <i>removeBimeraDenovo</i> with
522	default parameters. Taxonomic assignation of ASVs was performed using the assignTaxonomy
523	function from dada2 against the PR2 database <sup>30</sup> version 4.12 (https://pr2-database.org). ASV
524	assignation and ASV abundance in each sample were stored in two tables in the MySQL database.
525	ASV information was retrieved from the database using an R script.
526	
527	ASVs from each environment (freshwater, soil, marine euphotic, marine aphotic) were clustered into
528	OTUs at 97% similarity using VSEARCH <sup>86</sup> , to make the size of the dataset more manageable for
529	subsequent phylogenetic analyses. To be conservative in what was considered to be present in an
530	environment, we retained only those OTUs that were composed of at least 100 reads, or were present
531	in multiple distinct samples.
532	
533	Phylogenetic placement on global eukaryote phylogeny
534	Short-read OTUs were aligned against the long-read alignment (see Maximum likelihood analyses of
535	the global eukaryotic dataset) using the phylogeny-aware alignment software PaPaRa <sup>108</sup> . Misaligned
536	sequences were systematically checked and removed. OTUs from the four environments were then
537	phylogenetically placed on the global eukaryote tree (the tree with the highest likelihood) using EPA-
538	ng <sup>34</sup> . OTUs with high EDPL (expected distance between placement locations) indicate uncertainty in
539	placement, and were filtered out with the gappa command <i>edpl</i> <sup>96</sup> . The resulting jplace files were

- 540 visualised with  $iTOL^{109}$ .
- 541

## 542 Inferring clade specific phylogenies with short- and long-read data

543 In order to investigate clade-specific transition rates across the salt barrier, we inferred phylogenies for 544 major eukaryotic groups. We considered only those clades that contained sufficient data to more 545 precisely infer transition rates; i.e. both terrestrial and marine taxa were present, and there were at least 546 50 taxa present. This excluded taxa such as radiolarians (which contains no terrestrial taxa), rigifilids 547 (which contains only terrestrial taxa), and tubulineans (which is predominantly terrestrial with an 548 extremely small proportion of marine taxa). After preliminary analyses, we also excluded the clades 549 discobans and discoseans due to large topological differences in the resulting trees. 550 We extracted all short-read OTUs from the remaining 13 clades using the gappa subcommand *extract*. 551 Short-read OTUs taxonomically annotated as anything other than the respective clade were discarded 552 (for instance we discarded sequences labelled as amoebozoans that were phylogenetically placed in 553 apicomplexa). For each clade, we pruned the corresponding subtree (and an outgroup) from the global 554 phylogeny with the best likelihood score. For each clade, we then inferred 100 ML phylogenies with 555 RAxML (GTRCAT model), using the long-read subtree as a backbone constraint.

556

#### 557 Analyses of habitat evolution

### 558 <u>Unifrac analyses</u>

To estimate whether microbial communities from various habitats were phylogenetically distinct, we calculated unweighted UniFrac distance<sup>110</sup> as implemented in mothur, between (1) marine and terrestrial habitats, (2) marine euphotic, marine aphotic, soil, and freshwater, and (3) each sample sequenced with PacBio. Distances were estimated along the best ML global eukaryotic phylogeny with 1000 randomisations in order to test for statistical significance.

the four habitats (soil, freshwater, marine euphotic, marine aphotic) using the gappa subcommand *krd* 

Similarly, we estimated pairwise Kantorovich-Rubinstein distance (earth mover's distance) between

566 with the short-read placement files (jplace files) as input (See **Phylogenetic placement on global** 

# 567 eukaryotic phylogeny).

568

# 569 <u>Model test on global eukaryotic phylogeny</u>

570 To investigate whether transition rates vary between major eukaryotic clades, we compared a null 571 model (qMT and qTM remain constant throughout the global eukaryotic tree) against a complex 572 model (qMT and qTM estimated separately for each major eukaryotic clade) on the global eukaryotic phylogeny. These models were compared using MCMC analyses in BayesTraits v3.0.2<sup>111,112</sup> in a 573 574 reversible-jump framework in order to avoid over-parameterization<sup>113</sup>. Following the analysis in <sup>114</sup>, 575 we used 50 stones and a chain length of 5,000 to obtain marginal likelihood for each model using 576 stepping stone method<sup>115</sup>, and a Log Bayes Factor (2 \* difference of log marginal likelihoods) of 10 or 577 more was used to favour the complex model over the simple model. 578 Before final analyses in BayesTraits, we tried several prior distributions for transition rates (using a 579 hyperprior approach to reduce uncertainty about prior choice<sup>113</sup>). Specifically we compared gamma 580 hyperpriors with exponential hyperpriors using different values. While the different priors produced 581 qualitatively similar results, we found the exponential hyperprior to be most suitable. All BayesTraits 582 analyses were therefore carried out using an exponential hyperprior with the mean seeded from a 583 uniform distribution between 0 and 2. Additionally, all ancestral state reconstruction analyses were 584 carried out on 100 inferred phylogeneis to take phylogenetic uncertainty into account, and were 585 repeated thrice to check for convergence.

586

### 587 *Clade specific transition rates*

We inferred clade-specific transition rates along the clade-specific phylogenies (long-read + short-read data), on account of these being more complete. The metadata for each taxon was used to label it as either marine or terrestrial. We ran 1 million generations on each tree (100 million generations in total) with 0.5 million generations discarded as burn-in. For each clade, we also inferred the global transition rate, regardless of the direction of transition. This was achieved by normalising the QMatrix<sup>116,117</sup>, with all other parameters unchanged. These analyses also allowed us to infer the ancestral state of each major eukaryotic clade.

595

## 596 *Inferring ancestral states of deep nodes and the last common ancestor of eukaryotes*

597	In order to infer the ancestral habitats at deeper nodes (including the origin of eukaryotes), we			
598	modelled habitat evolution along the global eukaryotic phylogeny using the better suited complex			
599	model. Analyses were run for 500 million generations, forcing BayesTraits to spend 5 million			
600	generations on each tree, and 200 million generations were discarded as burn-in. Analyses were			
601	carried out after rooting the tree at Discoba, and at Amorphea in order to take uncertainty about the			
602	root into account.			
603				
604	Visualising scenarios of habitat evolution			
605	Most ancestral state reconstruction programmes do not explicitly calculate the ancestral state at			
606	internal nodes (but integrate over all possibilities). In order to visualise habitat evolution, we used			
607	PastML, a maximum likelihood ancestral state reconstruction programme which calculates the state at			
608	each internal node, and also generates a concise visual summary of the clade. For each major			
609	eukaryotic clade, we ran PastML on 100 trees. Visualisations for several trees were checked manually			
610	to assess if they displayed similar histories, and one visualisation was chosen randomly for display in			
611	Supplementary Figure 12.			
612				
613	Counting number and relative timing of transitions			
614	We converted all clade-specific phylogenies into relative chronograms (with the age of the root set to			
615	1) using Pathd8 <sup>38</sup> which is suitable for large phylogenies. We ran PastML on these phylogenies (as			
616	before), and used custom scripts <sup>83</sup> to count the number of marine-terrestrial transitions. For each			
617	transition, we calculated the distance to the root to obtain relative timing of transition.			
618				
619	Network analyses			
620	To check that our results about transition rates and timings were not biased by phylogenetic inference			
621	from sequences with poor phylogenetic signal, we constructed sequence similarity networks. These			
622	networks were constructed using representative 18S sequences of the long-read OTUs. Briefly, we			
623	performed all-against-all BLAST searches, and generated networks using a coverage threshold of 75,			

624	and sequence identity thresholds of 80, 85, 90, 95, 97. Networks were visualized on Cytoscape <sup>118</sup> .
625	Assortativities were calculated using scripts available at <sup>119</sup> , and then plotted in R using ggplot <sup>120</sup> .
626	
627	Data availability
628	New sequence data generated for this study were deposited at ENA under the accession number
629	PRJEB45931, while data from Sequel I (generated in <sup>28</sup> ) were deposited under the accession number

630 PRJEB25197. The PR2-transitions database, annotated 18S and 28S OTU sequences, clustered short

631 read metabarcoding sequences used in this study, and all trees have been deposited in an online

- 632 repository<sup>31</sup>. All custom code is available here<sup>83</sup>.
- 633

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- 654

# 655 Author contributions

- 656 F.B. and M.J. conceived the project. M.J., H.J., S.P., and R.M. collected samples and extracted DNA.
- 657 M.J. carried out long-range PCRs and processed the PacBio data. C.B. and D.V. collected and
- 658 processed short-read metabarcoding data. A.O. performed comparisons of long and short-read
- 659 metabarcoding data. M.J, and C.B. performed phylogenetic and ancestral state reconstruction analyses.
- 660 M.J and F.B. wrote the first draft of the manuscript and all authors read and commented on the
- 661 manuscript. F.B. supervised the project.
- 662

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