

1 Welcoming more participation in open 2 data science for the oceans

3
4 **Alexa L Fredston**, Department of Ocean Sciences, University of California Santa Cruz

5 **Julia S Stewart Lowndes**, National Center for Ecological Analysis and Synthesis, University of
6 California Santa Barbara

7
8 Keywords: marine science, open science, data science, inclusion

9 Abstract

10 Open science is a global movement happening across all research fields. It builds on years of
11 efforts by individual researchers and a broad array of institutions, agencies, and grassroots
12 organizations. Enabled by technology and the open web, the goal is to share knowledge and
13 broaden participation in science, from team formation and early ideation to making intermediate
14 and final research outputs openly accessible to all (“open access”). Because of its emphasis on
15 transparency and collaboration, the open science movement dovetails with efforts to increase
16 diversity, equity, inclusion, and belonging in science and society. The Year of Open Science
17 (2023), as declared by the US Biden-Harris Administration and many other US government
18 agencies, is a great opportunity to boost participation in open science for the oceans. For
19 researchers day-to-day, open science is a critical piece of modern workflows to analyze,
20 collaborate, and communicate increasing amounts of data. Therefore, we focus this piece on
21 open data science – the tooling and people enabling reproducible, transparent, inclusive
22 practices for data-intensive research – and its intersection with the marine sciences. We discuss
23 the state of various technical dimensions of open science – such as open-source programming
24 and academic publishing – and argue that technical advancements in open science have
25 outpaced our field’s culture change to adopt and incorporate them. We believe that increasing
26 inclusivity and technical skill building are interlinked and must be prioritized within the marine
27 science community to find collaborative solutions for mitigating and adapting to climate change
28 and other threats to marine food sources, biodiversity, habitats, and society. As marine
29 scientists whose careers have been profoundly influenced by and continue to benefit from open
30 science, we provide examples of participation in this movement and the social transformation
31 needed for the field of marine science to become truly “open”.

32 1. Welcome to marine open data science

33 Open science has many different definitions, and we can participate in a spectrum of products
34 (e.g. data, publications) and practices (e.g. sharing ideas and work-in-progress). One definition
35 by Ramachandran et al. (2021) is that open science is “a collaborative culture enabled by
36 technology that empowers the open sharing of data, information and knowledge within the
37 scientific community and the wider public to accelerate scientific research and understanding”.

38 The Biden-Harris Administration, along with NASA, NOAA, NSF, and dozens of other
39 government agencies, have declared 2023 as the Year of Open Science (NASA 2022; The
40 White House 2023) and are developing strategic policy to promote equitable participation.
41 Focusing more specifically on data analytical workflows and sharing in the marine sciences, we
42 define open data science as “the tooling and people enabling reproducible, transparent,
43 inclusive practices for data-intensive research” (Lowndes & Robinson 2022). “Open science” is
44 often interpreted to be synonymous with “open data” as a product; our definition of “open data
45 science” goes far beyond data sharing and includes the social and technical processes that
46 researchers encounter when working with data.

47
48 Open data science tools and practices are revolutionizing science by enabling transparent,
49 collaborative, and reproducible data-driven research, with recent examples including real-time
50 decision making amidst the COVID pandemic due to openly shared data (Zastrow 2020) and
51 capturing the first images of black holes using open source software (NumFOCUS). Most known
52 for underpinning robust data analyses and visualization, open data science also streamlines
53 collaboration and expands communications through modern online channels for contributing to,
54 publishing, and distributing research outputs (Bastille et al. 2021; Lowndes et al. 2017;
55 McKiernan et al. 2016). Aiming to make these practices the norm, scientific journals are
56 increasingly requiring that authors submit all data and code used for the analyses they publish
57 (Berberi & Roche 2022). Simultaneously, many researchers who may not identify as data
58 scientists – including marine scientists – are working with larger and more complex datasets
59 than ever before and code is now a requirement to do their science (Geiger 2018; Stoudt et al.
60 2021). Open access to these datasets enabled the data synthesis revolution and now allows
61 scientists to address foundational questions in fields like marine sciences at scales that were
62 previously unimaginable (Halpern et al. 2020). Yet in marine sciences, large gaps remain
63 between best practices in open science and the status quo (Hörstmann et al. 2021; Lowndes et
64 al. 2017), and a lack of data is still a common problem (Blasco et al. 2020). While the open
65 science movement is widespread in research generally, we identify two reasons why it is useful
66 to explore its intersection with marine science specifically.

67
68 First, the marine sciences are awash in far more data than ever before. For example, the Tara
69 Oceans Consortium has published molecular and environmental data from over 35,000 at-sea
70 samples (Pesant et al. 2015) and the Argo Program has shared temperature and salinity data
71 from over two million vertical profiles (Wong et al. 2020). Yet what each of us hears from our
72 peers and networks is that researchers do not know what data to use, where to find it, and how
73 to analyze it properly. We argue that marine science is not lacking in technical tools, but cannot
74 fully realize the potential of new data streams and computational methods without systemic
75 changes in our teams, communities, and institutions that underpin modern analytical skills and
76 collaborative culture.

77
78 Second, geosciences remain one of the least diverse STEM fields, partially due to hostile
79 workplace climates (Burton et al. 2023; Marín-Spiotta et al. 2020); similar issues persist in
80 ecology (Primack et al. 2023). We recognize that these inequities, and potential solutions, are
81 highly varied and multidimensional. We believe – based in part on our lived experiences – that

82 open data science can promote a sense of community and belonging, thereby helping to build a
83 more inclusive marine science (Fenwick et al. 2023; Gaynor et al. 2022; Lowndes 2019). In
84 other words, filling the gaps in social infrastructure for open science – via skill-building, culture
85 change, and work-life balance – can also move us toward more diverse, equitable, and inclusive
86 marine sciences.

87
88 Open data science is a continuum and not all-or-nothing; anywhere we're starting from is the
89 right place, and we can participate anywhere along this continuum from open hypotheses to
90 open data and code to open publishing. Open data science breaks the narrative of "I work
91 alone" and instead develops a mindset of reuse and of collaboration over competition, where we
92 learn with, from, and for others (Lowndes 2019). This is a mindset we can develop and role-
93 model for others, in any part of our careers. One way to participate incrementally is to work
94 openly with ourselves and our teams; sharing ideas, slides, and other work-in-progress, as well
95 as noticing who is participating, asking questions, listening, and creating inclusive spaces for
96 others to voice ideas and share as well (Robinson & Lowndes 2022). Here we frame our vision
97 for the future of marine science around open data, open code, and open publishing, and share
98 our personal narratives of what it looks like to be a modern marine researcher practicing open
99 science.

100

101 2. Introducing ourselves: Building careers with open science

102

103 As a marine scientist starting a PhD in 2014, Fredston experienced and was influenced by open
104 science from the very beginning, although she did not realize that at the time. Her early
105 research questions focused on understanding climate-related shifts in marine species'
106 geographical ranges, questions she knew were possible to answer by combining data from
107 bottom trawl surveys and hindcast oceanographic data products. While the concepts were clear,
108 the skills and tools needed to analyze the data were far from obvious. Fredston learned with
109 Lowndes and the Ocean Health Index team at the National Center for Ecological Analysis and
110 Synthesis (NCEAS) about coding collaboratively in R, shared version control with GitHub, and
111 open educational resources like online tutorials (Lowndes et al. 2017). "Eco-Data-Science", a
112 local community of practice, gave Fredston a supportive and encouraging space in which to
113 learn – and soon teach – crucial data science techniques (<http://eco-data-science.github.io>).
114 This PhD experience was somewhat unique in that Fredston was trained in open-source
115 technical skills as well as a culture of reuse, continued learning, sharing, and inclusive
116 collaboration from the start. Since then, she has published manuscripts as preprints, shared
117 data and code openly throughout her research workflow, participated in an international
118 collaboration aimed at data harmonization (Maureaud et al. 2023), and joined the Board of the
119 Society for Open, Reliable, and Transparent Ecology and Evolutionary biology (SORTEE;
120 <https://sortee.org>).

121

122 Meanwhile, open science was also changing Lowndes' career path. Open data science enabled
123 the Ocean Health Index team at NCEAS to work more reproducibly and efficiently, building from

124 established team culture of trust and horizontal leadership (Lowndes et al. 2017). Their work
125 was enabled through interacting more with open source software developers and community
126 builders through groups like rOpenSci, RStudio, RLadies, Carpentries, and Mozilla, groups
127 intentionally building a culture for kinder science (Lowndes 2019).

128
129 Feeling she could contribute to science most impactfully by supporting other researchers rather
130 than continuing research herself, in 2018 Lowndes founded Openscapes through a Mozilla
131 Open Science Fellowship and began formally mentoring marine science teams in academia
132 (Lowndes et al. 2019) and NOAA Fisheries. In 2020, Lowndes partnered with a leader in the
133 NASA Earthdata community and began working with NASA Earthdata teams (Lowndes &
134 Robinson 2021) and communities including the Earth Science Information Partners (ESIP), 2i2c,
135 Pangeo, pyOpenSci, NASA Transform to Open Science (TOPS), Ladies of Landsat, and Black
136 Women in Marine Ecology, Evolution and Marine Biology (BWEEMS). Her collaborative work
137 stems from a “kinder science for future us” mindset to welcome and empower marine,
138 environmental, and Earth scientists with existing open source tooling and practices, helping
139 them develop collaborative skills and join existing communities in the open science movement.

140
141 Stories like ours are important because they show a few of the many ways to participate in open
142 science. These examples are important to share “what open science looks like” in the 2023
143 Year of Open Science. We acknowledge that we are two white women who felt safe sharing our
144 work openly and were supported in our jobs to learn the skills needed to practice open science.
145 Safety and support are two elements that are fundamental to increasing participation in open
146 science in marine ecology, and something for which we continue to fiercely advocate.

147 3. Reusing and contributing data

148 Open data are freely available for reuse online in many data repositories (see Table 1 in Culina
149 et al. 2018). Those repositories are how many of us as marine scientists first engage in open
150 science. For example, we use sea surface temperature (SST) data from NOAA and NASA
151 satellites and find data published with a particular study in data repositories like Dryad and the
152 Open Science Framework (OSF). We are able to reuse and cite these data as we would a
153 research article to give proper attribution. Then, when we share our own data to these
154 repositories, they are assigned a digital object identifier (DOI) that others can use to credit us.
155 Metadata – data that describes and gives information about other data, for example where data
156 was collected and describing the column headers – is critical both for searching for existing and
157 contributing new data. Guidance is available for how and where to share data, including how to
158 format metadata and consider FAIR (findable, accessible, interoperable, and reusable) and/or
159 CARE principles (collective benefit, authority to control, responsibility, and ethics) (Barker et al.
160 2022; Carroll et al. 2020; Roche et al. 2022b).

161
162 However, finding these datasets can be challenging. From genes to populations to spatial data
163 products, our field’s synthesis and publication of data has vastly outpaced our guidance for its
164 interpretation and reuse. Amongst the numerous calls to share and store environmental data so
165 it is not lost (Bledsoe et al. 2022; Poisot et al. 2019; Wolkovich et al. 2012), open datasets have

166 proliferated and so have places to host them. One common question that we, as marine
167 synthesis scientists, often hear, is “Which data source should I use?”. While it is encouraging
168 that it is becoming more straightforward to upload and share data, this leads to difficulty in
169 selecting and interpreting datasets for reuse. For example, despite the relative ease of finding a
170 dataset on ocean temperature, there is almost no guidance to help marine scientists decide
171 which ocean temperature dataset to use and why.

172
173 This challenge is far beyond what we can address in this review, but we advocate for working
174 groups and other interdisciplinary collaborations to not only publish frequently-updated
175 guidelines to available datasets in their fields – pros, cons, and when to use each one – but help
176 teach others to normalize the practice. Important roadmaps for datasets exist for example in
177 ecology and evolution (Culina et al. 2018) and marine eDNA metabarcoding (Shea et al. 2023),
178 emphasizing the need to publish and maintain such roadmaps. Regardless of the topic, each
179 set of guidelines will likely need to address how to decide among seemingly similar datasets as
180 well as questions about spatial and temporal scale, uncertainty, and reconciling contradictory
181 data points.

182
183 Further, there are technical skills needed to access and work with data, which may or may not
184 be familiar to researchers. Datasets that follow FAIR guidelines have lower barriers to reuse
185 because they are designed to be found by a broad audience and reused for various
186 applications. Field-specific standards to enable interoperability exist or are under development
187 in many disciplines, such as the Darwin Core (DwC) standards for biodiversity data
188 (<https://www.tdwg.org/standards/dwc/>) and Minimum Information about any Sequence (MIxS)
189 for genomic data (<https://www.genesc.org/pages/standards-intro.html>). While field-specific
190 standards are a significant step, accurately interpreting and reusing these datasets nonetheless
191 requires domain expertise that marine scientists researching ever-broader topics may not have.
192 Sharing each field’s best practices for common operations, such as downscaling spatial data
193 layers, or extrapolating from traits of closely related species, are critical to provide along with the
194 data, as is support for coding and learning new software tools. These best practices would go
195 beyond metadata to chart a course for researchers to teach themselves how data should be
196 accurately and responsibly reused. Open communities around data specifications, for example
197 spatial operations in the R programming language (<https://r-spatial.org>) and the Zarr array data
198 format (<http://zarr.dev>) are important for marine scientists to engage with.

199

200 4. Coding and learning new software tools

201 Coding (and associated skills such as version control, for example with Git) is increasingly
202 critical for modern marine scientists as the use of climate models, oceanographic data products,
203 advanced statistical methods, and complex and multidimensional datasets in research projects
204 has become routine (Braga et al. 2023; Ram 2013). Open source languages enable broader
205 participation in coding since they are free to access and use. Because they are co-produced by
206 an entire community, open source languages can also more nimbly evolve to match new data,
207 techniques, and ideas. In our experience, R and Python are primary open languages used by

208 marine scientists, and Julia is emerging as another (and additional languages include Matlab,
209 C++, and Fortran.) Learning to code should be done with “good enough practices” (Wilson et al.
210 2017) and community resources rather than ad hoc or alone (Lowndes et al. 2017; McKiernan
211 et al. 2016). This means learning to write collaborative code that we expect others to see,
212 understand, and reuse – most importantly for a researcher’s future self long after the code was
213 originally written, i.e., “Future You” (Wilson et al. 2017) and “Future Us” (Lowndes et al. 2019).
214 Community resources relevant to marine scientists include rOpenSci and pyOpenSci, both of
215 which also focus on code review and package development – which are important practices for
216 creating scientific analyses via code.

217
218 Digital notebooks like Jupyter Notebooks and R Markdown help us iterate between writing,
219 coding, data visualization, and related tasks while we work, since our scripts, outputs (figures
220 and tables), and text are all in the same place (e.g., Czapanskiy et al. 2022; Ovando et al.
221 2021). Further, they have changed how we approach science communication and publishing
222 since we can create not only Word documents and PDFs but also webpages, websites, e-
223 books, slides, and more as a part of our daily workflows. Now, the interoperability between
224 these tools is increasing, with tools like Quarto that enable you to publish websites that are a
225 combination of Jupyter and RMarkdown notebooks, as well as reticulate, which enables you to
226 run python code from R. These tools enable us to leverage tools developed in a variety of open-
227 source languages without needing to translate between them. Open documentation to guide
228 researchers through these tools includes guidelines for best practices in archiving data,
229 metadata, and scripts (Gil et al. 2016; Jenkins et al. 2023; Reichman et al. 2011), code writing
230 (Filazzola & Lortie 2022; Wilson et al. 2017), and collaborative workflows (Barros et al. 2023;
231 Lowndes et al. 2017). Further, the FAIR and CARE standards can be applied to a broad range
232 of research software (Barker et al. 2022).

233
234 Learning to code and use new software takes time, and learning is a continual process, it is
235 never “done”. Despite coding and data skills being a large unmet need (Barone et al. 2017),
236 individual scientists often have to advocate to make learning part of their approved, paid time.
237 Additionally, one element preventing the formal instruction of these skills in universities is that
238 self-taught coders often feel like they are not expert enough to teach others (Williams et al.
239 2019, 2023). This is changing – slowly – as there is more visibility of this skills gap, and more
240 groups like Carpentries (<https://carpentries.org>) and RLadies (<https://rladies.org>) that teach at a
241 community level. It is encouraging to see more research software engineers (RSEs) in
242 academia – following long-standing work by groups like the US Research Software
243 Sustainability Institute (URSSI, <https://urssi.us/>) – and we hope to see increasing numbers of
244 marine science RSEs collaborating in the future.

245 5. Rethinking scientific publishing: sharing earlier and 246 rewarding more

247 Journals disseminating scientific results have been a mainstay of the research process for
248 centuries. However, profound issues endemic to the modern academic publishing enterprise –
249 such as the exorbitant costs of accessing articles managed by for-profit publishers and

250 academia’s overemphasis on journal prestige – are also not new (Walter & Mullins 2019).
251 Rather than recapitulate these well-described issues, we want to bring awareness to several
252 new dimensions of scientific publishing that are affecting all fields including marine sciences.

253
254 More and more researchers are sharing their work publicly at the manuscript draft stage as
255 preprints. A preprint is typically hosted on a dedicated server like *ecoevorxiv.org*, *biorxiv.org*,
256 *eartharxiv.org*, or *osf.io/preprints* that generates a DOI for the document. Preprint servers make
257 it easy to update manuscripts with new versions, so researchers often upload an early draft and
258 iteratively improve it with feedback and peer review; these are also increasingly leveraging
259 notebook technology (see previous section). If the article is eventually published in a peer-
260 reviewed journal, authors can link it to the preprint version. Citation counts for the preprint and
261 the published article will be pooled on sites like PubMed and Google Scholar, primary websites
262 for tracking researcher citations. Scientific journals increasingly support preprints, with some
263 even offering an option for researchers to publish their manuscript draft as a preprint upon
264 submission to the journal. Awareness of preprints skyrocketed during the COVID-19 pandemic
265 as scientists raced to share data and research to understand infection rates and spread
266 (Watson 2022). Of course, the key caveat of preprints – that they have not yet been peer-
267 reviewed and deemed sound for publication by experts – remains.

268
269 The core function of a preprint server—to be publicly accessible and free—makes it an
270 appealing option for practitioners of open science. Preprinting a draft manuscript allows
271 researchers to share their ideas with the scientific community much earlier, potentially
272 amplifying the impact of their work and increasing media attention and citation counts (Fu &
273 Hughey 2019). For example, an article Fredston co-authored that was first preprinted on
274 January 25, 2022 was published in a peer-reviewed journal on April 4, 2023; by that time it
275 already had over 1,500 views, almost 500 downloads, and several citations, and had stimulated
276 some discussion on social media. Google Scholar merged those citations and the digital record
277 of the preprint with the journal’s version of the article within a week of its publication. Preprints
278 are also a primary route to “green open access” (in which some version of the final manuscript,
279 such as a preprint, is freely available online; “gold open access” means that the publisher-
280 formatted final version is freely available) which is key to complying with funder mandates for
281 open access (Roche et al. 2022a).

282
283 Several journals have proposed creative models that aim to improve academic publishing. The
284 Public Library of Science (PLOS) operates its open-access-only journals as a nonprofit, and its
285 journal *PLOS ONE* does not evaluate submissions for novelty; the novelty criterion is one
286 possible culprit for the bias toward statistically significant results in published literature (Fanelli
287 2012). Other journals publish “registered reports”, in which authors receive peer review and may
288 have articles provisionally accepted after analyses have been designed but before the study is
289 actually conducted, to minimize publication bias and promote preregistration (O’Dea et al.
290 2021). eLife, also a nonprofit, recently transitioned to a “reviewed preprint” model where all
291 submissions that are sent out for peer review are eventually published open access with the
292 associated peer reviews. Journals published by academic societies, also typically nonprofits that

293 offer open access publishing options, use revenue from publishing for society activities and
294 likely provide a greater social good than strictly for-profit journals (Chytrý et al. 2023).

295

296 Whatever their model, all journals are now confronting a seismic shift toward open access in
297 scientific publishing (Butler et al. 2022), and their policies are evolving rapidly
298 (<https://v2.sherpa.ac.uk/romeo/>). A growing awareness of the difficulty of accessing paywalled
299 research led to “Plan S”, which launched in Europe in 2018 with the requirement that state-
300 funded research be published open access by 2021; agencies from around the globe have
301 since signed on to Plan S (<https://www.coalition-s.org/about/>). The U.S. government declared in
302 2022 that all research done with taxpayer funds must be published open access by 2025 (U.S.
303 Office of Science and Technology Policy 2022).

304

305 Although open access publishing removes one key barrier to equity in science—the paywall—it
306 has been criticized for introducing another one: the article publishing charges (APCs) levied by
307 journals to publish articles “gold open access” typically run in the thousands of dollars and are
308 prohibitively costly for scientists working outside of traditional research institutions in the
309 wealthiest nations. One scholar estimated that the average APC is equivalent to half a year’s
310 pay (or tuition and stipend, for a student) in many African nations, highlighting the inaccessibility
311 of gold open access publishing in most of the world (Mekonnen et al. 2021). Green open access
312 does not charge fees to authors, however, and many other models of open access publication
313 exist. For example, “diamond” or “platinum” open access journals do not charge fees either to
314 authors or to subscribers. The initiative around Plan S (“cOAlition S”) and many others are
315 working to expand diamond open access journals, which currently serve “a fine-grained variety
316 of generally small-scale, multilingual, and multicultural scholarly communities” (Ancion et al.
317 2022).

318

319 Publishing preprints and taking an inclusive approach to authorship (for a review of issues and
320 solutions regarding authorship, see Cooke et al. 2021; Nakagawa et al. 2023) are necessary but
321 not sufficient conditions for advancing open science in marine science. The incentive and
322 reward structure for professional marine scientists must also adapt to this new paradigm.
323 Specifically, collaboration and non-publication outputs must be rewarded more and publishing in
324 prestige journals must be rewarded less in assessment, tenure, and promotion (Leonelli et al.
325 2015; Merow et al. 2023; National Academies of Sciences, Engineering, and Medicine 2018;
326 Nosek et al. 2015). Roadmaps to achieve this at the institutional level already exist
327 (<https://coara.eu/>, <https://sfdora.org/>). Early-career researchers still frequently receive the advice
328 that “one first-author paper is worth five or ten co-authored papers.” Especially at research-
329 intensive universities, publications – especially in prestige journals – massively outweigh other
330 open science contributions such as data, code, and educational materials. This mindset
331 devalues precisely the inclusive team spirit that we believe is a vital ingredient for high-quality,
332 data-intensive open science.

333

6. Engaging community science for the sea

334 While marine sciences has work to do in how we share, access, and reuse data, code, and
335 manuscripts, we also acknowledge that many marine research projects remain data-limited – a
336 gap that can be partially filled using data collected by non-professionals (Binley & Bennett
337 2023). The value of “community science” or “citizen science” to the natural sciences is
338 exemplified by the growing body of research using data from software applications like
339 iNaturalist (<https://www.inaturalist.org/>) and eBird (<https://ebird.org/>) (Binley et al. 2022). These
340 smartphone apps are both educational, teaching users to identify the taxa around them, and
341 scientific, allowing those users to log taxon identifications and associated metadata. Data from
342 these apps is freely available online and has been used for an enormous range of research and
343 conservation purposes (Callaghan et al. 2022; Sullivan et al. 2017). Being designed to help the
344 public interpret the natural world around them, it is no surprise that the data from eBird and
345 iNaturalist skew heavily toward terrestrial taxa. Community science has been slow in coming to
346 the oceans, but examples do exist like JellyWatch (<https://jellywatch.org>), Go-Sea
347 (<https://www.inaturalist.org/projects/go-sea>), Seafarer (Seafarers et al. 2017) and ocean-
348 focused “bioblitzes” (intensive biodiversity surveys done in a short time by community
349 scientists), and we suspect that explosive growth in marine science apps and community
350 engagement is right around the corner. Some of these initiatives developed new platforms and
351 others leveraged existing technology like iNaturalist with a new emphasis on marine systems.

352

353 One example of how valuable community science can be for marine science is Redmap
354 Australia, a tailored public outreach initiative and associated app that encourages users to
355 record any marine organism that seems uncommon, unfamiliar, or out of place. Similar to the
356 other apps mentioned, Redmap seeks both to collect data on range-shifting marine species and
357 to educate the public about the effects of climate change (Pecl et al. 2019). The program has
358 succeeded on both counts: the community scientists demonstrated more trust and social license
359 toward fellow marine stakeholders (Kelly et al. 2019), and the Redmap dataset has been used
360 to model fish species responding to warming oceans (Champion et al. 2018). With myriad social
361 and scientific benefits, we hope that these types of programs receive the sustained funding and
362 attention that they need to be implemented throughout global coastal communities.

363

364

7. Broadening participation in marine open data science

365 Surmounting the formidable challenges facing the marine sciences – as our field strives to
366 understand fundamental processes, conserve biodiversity, manage natural resources, and
367 forecast future states even while cumulative human impacts to the oceans mount – requires
368 widespread uptake of open science practices. The way that data are collected, stored, and
369 shared; the structure and inclusivity of collaborative teams; the computational methods we use;
370 and the pathways for communicating scientific results are all rapidly evolving as a result of the
371 open science movement. The tools for open marine science exist and are becoming easier to
372 learn and more accessible. The main barriers to uptake of open science in marine sciences, as
373 in some other fields (Hipsley & Sherratt 2019), are often are often social challenges: figuring out

374 which datasets to reuse and how to do so correctly, having supported time to learn new
375 technical skills, navigating the incentive structure of academic publishing, and reforming
376 institutions to encourage open science practices.

377
378 It can be overwhelming to realize that the marine science field now expects us to utilize cutting-
379 edge software tools and meet ever-higher standards for data and code quality and sharing while
380 continuing to push boundaries with our scientific questions and, often, advancing field and
381 laboratory research programs. We argue that the scope of questions enabled by the open
382 science movement has far outpaced researchers' skills in being able to answer them, and we
383 need more supported time to learn and teach these skills. Indeed, limited time and skills explain
384 why many researchers do not participate fully in open science practices (Gomes et al. 2022;
385 O'Dea et al. 2021). Skill-building is important for early-career researchers, who are central to
386 changing norms in science (Gownaris et al. 2022). And, skill-building is also important for
387 researchers at every career stage, who need continued learning time as part of their jobs so
388 they can continue to participate in modern open science practices throughout their careers,
389 whether as researchers, mentors, or/and supervisors (Robinson & Lowndes 2022).

390
391 Paid learning time is something that needs to be built into jobs across career stages, and is an
392 issue of diversity, equity, and inclusion: continuing the trend that scientists learn to code "on
393 their own time" exacerbates societal inequities. Open educational resources exist (two
394 examples: <https://stat545.com/> and <https://datacarpentry.org/semester-biology/>); what is needed
395 are paid time and career incentives for teaching and skillbuilding. Given the rapid pace of
396 progress in open science tools and the current lack of institutional incentives to engage with
397 them (Soeharjono & Roche 2021), we strongly advocate for giving working marine scientists
398 opportunities to be paid to learn regularly, including through paid leave or sabbatical.

399
400 In our experience, the conversation has shifted from "I don't want to do open science" to "I don't
401 know how" and "I don't have time". This shift deserves a celebration of the long-term work of the
402 global open science movement, such that the motivation and benefits of open science are
403 understood by researchers. Now, missing is how researchers are expected to learn the skills to
404 reap those benefits while continuing their disciplinary work and achieving a work-life balance.
405 Even if open science practices will eventually make researchers more efficient, the learning
406 curve may require an unacceptably high investment of time and effort that is unfeasible without
407 paid time across career stages. It is especially crucial to reckon with the way open science is
408 adding strain to researchers' workloads in the context of the burnout crisis in academia that was
409 accelerated by the COVID-19 pandemic (Gewin 2021).

410
411 In this article, we discussed some trends in the open science movement – particularly in the
412 context of data, code, publishing, and community science – that marine scientists can engage
413 in. We are grateful for the academic publishers and governments that are spearheading policy
414 to require open science practices, as well as groups changing tenure and promotion structure to
415 reward not only high-impact lead-author publications but all types of open science contributions
416 (e.g. software packages, data, open educational resources, and artwork)
417 (<https://sparcopen.org/>). And we call for more support for researchers to develop the skills they

418 need to meet those policies: transforming how we are educated so that marine scientists learn
419 data science tools and open science practices as part of our coursework, rather than solely via
420 voluntary participation in extracurricular groups, would lift a huge burden off of researchers at all
421 career stages. Since most university faculty are currently ill-equipped to teach these courses
422 (Emery et al. 2021), institutional incentives to support teachers and learners of open science
423 skills – and to hire open data science professors of practice – are key. Institutional support for
424 open science communities of practice like Eco-Data-Science can also provide a support system
425 and structured learning environment for scientists at all career stages.

426
427 Our career paths provide examples of what the open data science vision can look like in marine
428 science. Fredston benefited enormously from colleagues whose employment was secure
429 enough that they could invest substantial time and energy in maintaining their own cutting-edge
430 skills and teaching her and others. Since then, she has been able to “pay it forward” by
431 becoming a teacher of these same methods and ideas, for example teaching GitHub and
432 speaking to audiences of programmers about environmental applications. Through open
433 science, Lowndes has engaged with global efforts supporting science and scientists. Through
434 Openscapes, her work helps researchers in academia, government, and non-profits explore
435 these resources together with their teams and a cohort of peers, developing habits for
436 collaboration and reducing the loneliness that is such an unfortunate common feeling of learning
437 to code alone (<https://openscapes.org>). Together, we have both conducted marine synthesis
438 research that relied on others making data publicly available along with associated metadata so
439 it could be reused accurately. Most importantly of all, we both have been part of marine data
440 science communities that celebrated our successes, normalized our failures, and opened up
441 pathways for us to lead. We recognize that to date that experience is relatively rare in marine
442 sciences. Our hope for this article is to provide a welcome to marine scientists so that everyone
443 in our field feels empowered and included with opportunities to join the open science movement.

444 Acknowledgments

445
446 We are grateful to Dominique Roche, Rachel Turba, and Amanda Whitmire, who provided
447 invaluable feedback on earlier drafts of the manuscript.

448 References

- 449 Ancion Z, Borrell-Damián L, Mounier P, Rooryck J, Saenen B. 2022. Action Plan for Diamond
450 Open Access
- 451 Barker M, Chue Hong NP, Katz DS, Lamprecht A-L, Martinez-Ortiz C, et al. 2022. Introducing
452 the FAIR Principles for research software. *Sci. Data*. 9(1):622
- 453 Barone L, Williams J, Micklos D. 2017. Unmet needs for analyzing biological big data: A survey
454 of 704 NSF principal investigators. *PLOS Comput. Biol.* 13(10):e1005755

455 Barros C, Luo Y, Chubaty AM, Eddy IMS, Micheletti T, et al. 2023. Empowering ecological
456 modellers with a PERFICT workflow: Seamlessly linking data, parameterisation, prediction,
457 validation and visualisation. *Methods Ecol. Evol.* 14(1):173–88

458 Bastille K, Hardison S, deWitt L, Brown J, Samhouri J, et al. 2021. Improving the IEA Approach
459 Using Principles of Open Data Science. *Coast. Manag.* 49(1):72–89

460 Berberi I, Roche DG. 2022. No evidence that mandatory open data policies increase error
461 correction. *Nat. Ecol. Evol.* 6(11):1630–33

462 Binley AD, Bennett JR. 2023. The data double standard. *Methods Ecol. Evol.*

463 Binley AD, Vincent JG, Rytwinski T, Proctor CA, Urness ES, et al. 2022. Patterns of community
464 science data use in peer-reviewed research on biodiversity

465 Blasco GD, Ferraro DM, Cottrell RS, Halpern BS, Froehlich HE. 2020. Substantial Gaps in the
466 Current Fisheries Data Landscape. *Front. Mar. Sci.* 7:

467 Bledsoe EK, Burant JB, Higinio GT, Roche DG, Binning SA, et al. 2022. Data rescue: saving
468 environmental data from extinction. *Proc. R. Soc. B Biol. Sci.* 289(1979):20220938

469 Braga PHP, Hébert K, Hudgins EJ, Scott ER, Edwards BPM, et al. 2023. Not just for
470 programmers: How GitHub can accelerate collaborative and reproducible research in ecology
471 and evolution. *Manubot*

472 Burton C, Duran G, Wright V, Chmiel R. 2023. Strategies for and Barriers to Collaboratively
473 Developing Anti-Racist Policies and Resources as Described by Geoscientists of Color
474 Participating in the Unlearning Racism in Geoscience (URGE) Program. *Earths Future.*
475 11(2):e2022EF002957

476 Butler L-A, Matthias L, Simard M-A, Mongeon P, Haustein S. 2022. The Oligopoly's Shift to
477 Open Access. How For-Profit Publishers Benefit from Article Processing Charges. Zenodo

478 Callaghan CT, Mesaglio T, Ascher JS, Brooks TM, Cabras AA, et al. 2022. The benefits of
479 contributing to the citizen science platform iNaturalist as an identifier. *PLOS Biol.*
480 20(11):e3001843

481 Carroll SR, Garba I, Figueroa-Rodríguez OL, Holbrook J, Lovett R, et al. 2020. The CARE
482 Principles for Indigenous Data Governance. . 19(1):43

483 Champion C, Hobday AJ, Tracey SR, Pecl GT. 2018. Rapid shifts in distribution and high-
484 latitude persistence of oceanographic habitat revealed using citizen science data from a climate
485 change hotspot. *Glob. Change Biol.* 24(11):5440–53

486 Chytrý M, Pillar VD, Price JN, Wagner V, Wiser SK, Zelený D. 2023. The benefits of publishing
487 in society-owned scientific journals. *Appl. Veg. Sci.* 26(1):e12705

488 Cooke SJ, Nguyen VM, Young N, Reid AJ, Roche DG, et al. 2021. Contemporary authorship
489 guidelines fail to recognize diverse contributions in conservation science research. *Ecol. Solut.*
490 *Evid.* 2(2):e12060

491 Culina A, Baglioni M, Crowther TW, Visser ME, Woutersen-Windhouver S, Manghi P. 2018.
492 Navigating the unfolding open data landscape in ecology and evolution. *Nat. Ecol. Evol.*
493 2(3):420–26

494 Czapanskiy MF, Ponganis PJ, Fahlbusch JA, Goldbogen JA. 2022. Compendium of R code and
495 data for “An accelerometer-derived ballistocardiogram method for detecting heart rates in free-
496 ranging marine mammals”.

497 Emery NC, Crispo E, Supp SR, Farrell KJ, Kerkhoff AJ, et al. 2021. Data Science in
498 Undergraduate Life Science Education: A Need for Instructor Skills Training. *BioScience*.
499 (biab107):
500 Fenwick I, Davis A, Williams A, Butland S, Lowndes JSS. 2023. Community building for Black
501 environmental and marine researchers: Pathways to Open Science
502 Filazzola A, Lortie C. 2022. A call for clean code to effectively communicate science. *Methods*
503 *Ecol. Evol.* 13(10):2119–28
504 Fu DY, Hughey JJ. 2019. Releasing a preprint is associated with more attention and citations for
505 the peer-reviewed article. *eLife*. 8:e52646
506 Gaynor KM, Azevedo T, Boyajian C, Brun J, Budden AE, et al. 2022. Ten simple rules to
507 cultivate belonging in collaborative data science research teams. *PLOS Comput. Biol.*
508 18(11):e1010567
509 Geiger RS. 2018. *Reports from the BIDS Best Practices in Data Science Series*.
510 <https://stuartgeiger.com>
511 Gewin V. 2021. Pandemic burnout is rampant in academia. *Nature*. 591(7850):489–91
512 Gil Y, David CH, Demir I, Essawy BT, Fulweiler RW, et al. 2016. Toward the Geoscience Paper
513 of the Future: Best practices for documenting and sharing research from data to software to
514 provenance. *Earth Space Sci.* 3(10):388–415
515 Gomes DGE, Pottier P, Crystal-Ornelas R, Hudgins EJ, Foroughirad V, et al. 2022. Why don't
516 we share data and code? Perceived barriers and benefits to public archiving practices. *Proc. R.*
517 *Soc. B Biol. Sci.* 289(1987):20221113
518 Gownaris NJ, Vermeir K, Bittner M-I, Gunawardena L, Kaur-Ghumaan S, et al. 2022. Barriers to
519 Full Participation in the Open Science Life Cycle among Early Career Researchers. *Data Sci. J.*
520 21(1):2
521 Halpern BS, Berlow E, Williams R, Borer ET, Davis FW, et al. 2020. Ecological Synthesis and
522 Its Role in Advancing Knowledge. *BioScience*. 70(11):1005–14
523 Hipsley CA, Sherratt E. 2019. Psychology, not technology, is our biggest challenge to open
524 digital morphology data. *Sci. Data*. 6(1):41
525 Hörstmann C, Buttigieg PL, Simpson P, Pearlman J, Waite AM. 2021. Perspectives on
526 Documenting Methods to Create Ocean Best Practices. *Front. Mar. Sci.* 7:1260
527 Jenkins GB, Beckerman AP, Bellard C, Benítez-López A, Ellison AM, et al. 2023.
528 Reproducibility in ecology and evolution: Minimum standards for data and code. *Ecol. Evol.*
529 13(5):e9961
530 Kelly R, Fleming A, Pecl GT. 2019. Citizen science and social licence: Improving perceptions
531 and connecting marine user groups. *Ocean Coast. Manag.* 178:104855
532 Leonelli S, Spichtinger D, Prainsack B. 2015. Sticks and carrots: encouraging open science at
533 its source. *Geo Geogr. Environ.* 2(1):12–16
534 Lowndes J, Robinson E. 2022. Supporting open science as a daily practice
535 Lowndes JSS. 2019. Open Software Means Kinder Science. *Sci. Am. Blog Netw.*, .
536 <https://blogs.scientificamerican.com/observations/open-software-means-kinder-science/>
537 Lowndes JSS, Best BD, Scarborough C, Afflerbach JC, Frazier MR, et al. 2017. Our path to
538 better science in less time using open data science tools. *Nat. Ecol. Evol.* 1(6):1–7
539 Lowndes JSS, Froehlich HE, Horst A, Jayasundara N, Pinsky ML, et al. 2019. Supercharge your
540 research: a ten-week plan for open data science. *Nature*

541 Lowndes JSS, Robinson EM. 2021. The NASA-Openscapes Framework | Zenodo
542 Marín-Spiotta E, Barnes RT, Berhe AA, Hastings MG, Mattheis A, et al. 2020. Hostile climates
543 are barriers to diversifying the geosciences. *Adv. Geosci.* 53:117–27
544 Maureaud AA, Palacios-Abrantes J, Kitchel Z, Mannocci L, Pinsky M, et al. 2023.
545 FishGlob_data: an integrated database of fish biodiversity sampled with scientific bottom-trawl
546 surveys. Open Science Framework
547 McKiernan EC, Bourne PE, Brown CT, Buck S, Kenall A, et al. 2016. How open science helps
548 researchers succeed. *eLife.* 5:e16800
549 Mekonnen A, Downs C, Effiom EO, Razafindratsima O, Stenseth NC, Chapman CA. 2021.
550 What costs half a year’s pay for African scholars? Open access. *Nature.* 596(7871):189–189
551 Merow C, Boyle B, Enquist BJ, Feng X, Kass JM, et al. 2023. Better incentives are needed to
552 reward academic software development. *Nat. Ecol. Evol.* 7(5):626–27
553 Nakagawa S, Ivimey-Cook ER, Grainger MJ, O’Dea RE, Burke S, et al. 2023. Method Reporting
554 with Initials for Transparency (MeRIT) promotes more granularity and accountability for author
555 contributions. *Nat. Commun.* 14(1):1788
556 NASA. 2022. *2023: Year of Open Science.* NASA TOPS. <https://nasa.github.io>
557 National Academies of Sciences, Engineering, and Medicine. 2018. *Open Science by Design:
558 Realizing a Vision for 21st Century Research.* National Academies Press
559 Nosek BA, Alter G, Banks GC, Borsboom D, Bowman SD, et al. 2015. Promoting an open
560 research culture. *Science.* 348(6242):1422–25
561 NumFOCUS. *Case Study: First Photograph of a Black Hole, Enabled by NumFOCUS Tools.*
562 NumFOCUS. <https://numfocus.org>
563 O’Dea RE, Parker TH, Chee YE, Culina A, Drobniak SM, et al. 2021. Towards open, reliable,
564 and transparent ecology and evolutionary biology. *BMC Biol.* 19(1):68
565 Ovando D, Caselle JE, Costello C, Deschenes O, Gaines SD, et al. 2021. Assessing the
566 population-level conservation effects of marine protected areas. *Conserv. Biol.* 35(6):1861–70
567 Pecl GT, Stuart-Smith J, Walsh P, Bray DJ, Kusetic M, et al. 2019. Redmap Australia:
568 Challenges and Successes With a Large-Scale Citizen Science-Based Approach to Ecological
569 Monitoring and Community Engagement on Climate Change. *Front. Mar. Sci.* 6:349
570 Pesant S, Not F, Picheral M, Kandels-Lewis S, Le Bescot N, et al. 2015. Open science
571 resources for the discovery and analysis of Tara Oceans data. *Sci. Data.* 2(1):150023
572 Poisot T, Bruneau A, Gonzalez A, Gravel D, Peres-Neto P. 2019. Ecological Data Should Not
573 Be So Hard to Find and Reuse. *Trends Ecol. Evol.* 0(0):
574 Primack RB, Miller TK, Terry C, Marín-Spiotta E, Templer PH, et al. 2023. Historically excluded
575 groups in ecology are undervalued and poorly treated. *Front. Ecol. Environ.*
576 Ram K. 2013. Git can facilitate greater reproducibility and increased transparency in science.
577 *Source Code Biol. Med.* 8:7
578 Ramachandran R, Bugbee K, Murphy K. 2021. From Open Data to Open Science. *Earth Space
579 Sci.* 8(5):e2020EA001562
580 Reichman OJ, Jones MB, Schildhauer MP. 2011. Challenges and Opportunities of Open Data in
581 Ecology. *Science.* 331(6018):703–5
582 Robinson E, Lowndes JSS. 2022. The Openscapes Flywheel: A framework for managers to
583 facilitate and scale inclusive Open science practices

584 Roche DG, O’Dea RE, Kerr KA, Rytwinski T, Schuster R, et al. 2022a. Closing the knowledge-
585 action gap in conservation with open science. *Conserv. Biol.* 36(3):
586 Roche DG, Raby GD, Norin T, Ern R, Scheuffele H, et al. 2022b. Paths towards greater
587 consensus building in experimental biology. *J. Exp. Biol.* 225(Suppl_1):jeb243559
588 Seafarers SD, Lavender S, Beaugrand G, Outram N, Barlow N, et al. 2017. Seafarer citizen
589 scientist ocean transparency data as a resource for phytoplankton and climate research. *PLOS*
590 *ONE.* 12(12):e0186092
591 Shea MM, Kuppermann J, Rogers MP, Smith DS, Edwards P, Boehm AB. 2023. Systematic
592 review of marine environmental DNA metabarcoding studies: toward best practices for data
593 usability and accessibility. *PeerJ.* 11:e14993
594 Soeharjono S, Roche DG. 2021. Reported Individual Costs and Benefits of Sharing Open Data
595 among Canadian Academic Faculty in Ecology and Evolution. *BioScience.* 71(7):750–56
596 Stoudt S, Vásquez VN, Martinez CC. 2021. Principles for data analysis workflows. *PLOS*
597 *Comput. Biol.* 17(3):e1008770
598 Sullivan BL, Phillips T, Dayer AA, Wood CL, Farnsworth A, et al. 2017. Using open access
599 observational data for conservation action: A case study for birds. *Biol. Conserv.* 208:5–14
600 The White House. 2023. FACT SHEET: Biden-Harris Administration Announces New Actions to
601 Advance Open and Equitable Research | OSTP
602 U.S. Office of Science and Technology Policy. 2022. OSTP Issues Guidance to Make Federally
603 Funded Research Freely Available Without Delay. Aug. 25, .
604 [https://www.whitehouse.gov/ostp/news-updates/2022/08/25/ostp-issues-guidance-to-make-](https://www.whitehouse.gov/ostp/news-updates/2022/08/25/ostp-issues-guidance-to-make-federally-funded-research-freely-available-without-delay/)
605 [federally-funded-research-freely-available-without-delay/](https://www.whitehouse.gov/ostp/news-updates/2022/08/25/ostp-issues-guidance-to-make-federally-funded-research-freely-available-without-delay/)
606 Walter P, Mullins D. 2019. From symbiont to parasite: the evolution of for-profit science
607 publishing. *Mol. Biol. Cell.* 30(20):2537–42
608 Watson C. 2022. Rise of the preprint: how rapid data sharing during COVID-19 has changed
609 science forever. *Nat. Med.* 28(1):2–5
610 Williams JJ, Drew JC, Galindo-Gonzalez S, Robic S, Dinsdale E, et al. 2019. Barriers to
611 integration of bioinformatics into undergraduate life sciences education: A national study of US
612 life sciences faculty uncover significant barriers to integrating bioinformatics into undergraduate
613 instruction. *PLOS ONE.* 14(11):e0224288
614 Williams JJ, Tractenberg RE, Batut B, Becker E. 2023. Optimizing Short-format Training: an
615 International Consensus on Effective, Inclusive, and Career-spanning Professional
616 Development in the Life Sciences and Beyond | bioRxiv. *BioRxiv Prepr.*
617 Wilson G, Bryan J, Cranston K, Kitzes J, Nederbragt L, Teal TK. 2017. Good enough practices
618 in scientific computing. *PLOS Comput. Biol.* 13(6):e1005510
619 Wolkovich EM, Regetz J, O’Connor MI. 2012. Advances in global change research require open
620 science by individual researchers. *Glob. Change Biol.* 18(7):2102–10
621 Wong APS, Wijffels SE, Riser SC, Pouliquen S, Hosoda S, et al. 2020. Argo Data 1999–2019:
622 Two Million Temperature-Salinity Profiles and Subsurface Velocity Observations From a Global
623 Array of Profiling Floats. *Front. Mar. Sci.* 7:
624 Zastrow M. 2020. Open science takes on the coronavirus pandemic. *Nature.* 581(7806):109–10
625