

1 Exploratory and confirmatory conservation research in the  
2 open science era

3 *Erlend B. Nilsen<sup>1</sup>, Diana E. Bowler<sup>1,2</sup> & John D.C. Linnell<sup>1</sup>*

4 1: Norwegian Institute for Nature Research, P.O. 5685 Torgarden, 7485 Trondheim, Norway

5 2: German Centre for Integrative Biodiversity Research (iDiv), Deutscher Pl. 5E, 04103

6 Leipzig, Germany

7

8 **Corresponding author:** Erlend B. Nilsen ([erlend.nilsen@nina.no](mailto:erlend.nilsen@nina.no))

9 **Word count:** 3999

10

11

12

13 Abstract:

- 14 1. Conservation biology is becoming a more open science, with an increasing focus on  
15 large-scale assessments of the patterns and processes of biodiversity dynamics.  
16 However, the new challenges arising when it comes to defining exploratory and  
17 confirmatory research practices, has been so far overlooked. We discuss how the  
18 research community could meet these new challenges by allowing full use of different of  
19 scientific approaches, without blurring the distinction between exploration and  
20 hypothesis-testing confirmatory research.
- 21 2. A rapid screening of a random selection of articles from the literature suggests that  
22 neither experimental protocols nor hypothesis testing *sensu stricto* are common in  
23 conservation biology. Most experiments are carried out on small spatial scales, which  
24 contrast with current global policy processes and research trends towards large spatial  
25 and temporal scales.
- 26 3. We suggest that a clearer distinction between exploratory and confirmatory research can  
27 be achieved by reevaluating the important, but different, role that each plays in the  
28 scientific process.
- 29 4. This clearer distinction could be facilitated by allocating journal sections to the different  
30 types of research, embracing new tools offered by the open science era, such as pre-  
31 registration of hypothesis, establishing new systems where posthoc hypothesis emerging  
32 through exploration can also be registered for later testing, and more broad adoption of  
33 causal inference methods that foster more structured establishment of hypotheses  
34 about causal mechanisms.

35 5. To fully gain the benefits from the open science era, researchers, funding bodies and  
36 journal editors should explicitly consider how incentives could encourage openness  
37 about methods and approaches, and value the full plurality of scientific approaches.

38

## 39 Rigorous science in conservation biology

40 As a response to global biodiversity loss, conservation biology is increasingly focused on  
41 detecting patterns of biodiversity change, isolating the factors that are causing this loss, and  
42 ultimately suggesting mitigation measures or management solutions. Conservation biology  
43 has inherited scientific tools and traditions from older sciences such as ecology and wildlife  
44 management (Caughley 1994), but is at its core an applied, interdisciplinary and mission  
45 driven science (Soulé 1996). Because biodiversity loss and ecosystem transformations are  
46 expected to cause major challenges to present and future human societies (Millennium-  
47 Ecosystem-Assessment 2005), the transparency and rigor of the science that underpins  
48 policy and management decisions is decisive to the wellbeing of future generations of  
49 humans.

50 Following some high-profile publications pointing towards a reproducibility crisis in fields like  
51 psychology (Nosek & Collaboration 2015) and social research (Camerer *et al.* 2018), there  
52 has been much focus on repeatability and reproducibility of scientific results (see e.g. the  
53 news feature in Nature by Baker 2016). One of the consequences of this renewed focus is  
54 the global focus on FAIR data management and open sharing of research data (Wilkinson *et*  
55 *al.* 2016), software and code used to perform statistical analysis. These changes are all parts  
56 of a more general movement towards “open science” (Nosek *et al.* 2015).

57 Increasing accessibility of new data sources allows researchers to apply a wide range of  
58 models to data for exploratory science. This contrast with the pleas for more widespread  
59 adoption of confirmatory research where hypotheses are described a-priori and then  
60 carefully tested based on empirical data (Caughley 1994; Houlahan *et al.* 2017). A rapid  
61 screening of a sample from the conservation literature (**Box 1**) suggest that conservation

62 biology researchers often do not follow the strong inference paradigm (Platt 1964; Sells *et*  
63 *al.* 2018), nor do they follow the hypothetico-deductive method. Our rapid screening of the  
64 literature also suggests that large scale studies often have large impacts if measured through  
65 citation rates (**Box 1**). Here, we discuss how both exploratory and confirmatory research is  
66 needed in the field of conservation biology, how we can improve understanding in the open  
67 science era, and how both scientists and journal editors should assist in the task of extracting  
68 the maximum value from different scientific approaches without blurring the distinction  
69 between exploration and confirmation.

70

71 [A mature research community should value both exploration and confirmation](#)

72 Many earlier authors, including Caughley (1994), Sells *et al.* (2018) and Betini, Avgar and  
73 Fryxell (2017), have called for more formal use of the hypothetico-deductive method and the  
74 strong inference paradigm (*sensu* Platt 1964) within conservation biology and wildlife  
75 management. We agree with that plea, but also underline the fundamental role that  
76 descriptive studies documenting the state or trends of local or global biodiversity, or the  
77 natural history of species, has for conservation biology (Beissinger & Peery 2007; Pereira *et*  
78 *al.* 2013; Lehtikoinen *et al.* 2019). Recently, the emergence of Essential Biodiversity Variables  
79 (EBV) emphasises that robust descriptive research combined with observational data is still  
80 fundamental to our scientific progress (Pereira *et al.* 2013). Moreover, the developments of  
81 the United Nation's Sustainability Goals (SDG) and a movement towards more planetary  
82 scale assessments, such as those carried out by the Intergovernmental Panel on Biodiversity  
83 and Ecosystem Services (IPBES), makes it unfeasible for policy to rely mainly on insights  
84 gained from experimental research (Mazor *et al.* 2018; Box 1).

85 Nevertheless, to avoid an ever-growing list of un-tested hypothesis emerging from  
86 exploratory research, we must also reevaluate the fundamental (but different) role that  
87 hypothesis-testing and prediction play in conservation biology research (Houlahan *et al.*  
88 2017). Only by testing a-priori articulated hypothesis can we robustly confirm or reject the  
89 potential of a scientific hypothesis to describe natural phenomena. However, studies do not  
90 always follow such protocols and surveys have revealed the existence of a number of  
91 questionable research practices (Ioannidis *et al.* 2014; Fraser *et al.* 2018). Such practices  
92 include both “*harking*” (Hypothesis After Results Are Known), where ad-hoc postdictions are  
93 presented as if they were already planned before the study was conducted, and “*p-hacking*”  
94 where researchers carelessly search for significant associations in the data (and often  
95 present them as if they were from a-priori hypotheses). Recent surveys suggest that they  
96 might be common also among ecologists and evolutionary biologists (Fraser *et al.* 2018).  
97 Without more frequent use of prediction, we risk that confirmation bias and the personal  
98 beliefs of the scientists will result in overly self-confident ‘storytelling’ with weak scientific  
99 support (Hayward *et al.* 2019). Basing conservation planning and mitigation actions on such  
100 research may lead to costly mis-management.

101

## 102 [Novel ways to test ecological theories](#)

103 Our survey of the literature (**Box 1**) (see also Betini, Avgar & Fryxell 2017; Sells *et al.* 2018)  
104 suggest that conservation biology research most often does not confirm to strict hypothesis  
105 testing. In the open science era, there are ample possibilities to increase the use and impact  
106 of confirmatory research, by more widely embracing new tools and methods, and especially  
107 increased data availability.

108 Strict experiments in conservation biology (**Box 1**) are generally conducted at small local  
109 spatial scales (although there are some very notable exceptions, e.g. Krebs, Boutin & Boonstra  
110 1995; Wiik *et al.* 2019). This contrasts with the fact that many ecological and policy  
111 processes operate at far larger scales (Estes *et al.* 2018). Better utilization of large-scale  
112 unreplicated natural experiments could facilitate an improved understanding of causal  
113 relationships in ecological systems (Barley & Meeuwig 2017; Serrouya *et al.* 2019), especially  
114 the impacts of rare and extreme events (e.g. Gaillard *et al.* 2003). A complementary  
115 approach, when experiments are not feasible, would be to apply methods that allow  
116 integration of findings from small-scale manipulative experiments into large-scale synthesis  
117 of drivers of biodiversity change. Such integration will necessitate closer collaboration  
118 between ecologists working on different spatial scales, and between experimentalists and  
119 modellers (Heuschele *et al.* 2017). The increased popularity of hierarchical statistical models  
120 and methods to integrate data from disparate data sources (Nilsen & Strand 2018; Miller *et*  
121 *al.* 2019) facilitate such an integration. In the new era of open science, large amounts of  
122 data from both field surveys and experiments are now becoming available making such  
123 integration much more feasible.

124 Given our reliance on observational data, conservation biology research could gain more  
125 insight into causal processes by more widely applying novel statistical methods that are seek  
126 to establish causality from observational data (Law *et al.* 2017). A side effect of adopting  
127 causal inference approaches is forcing researchers to think more deeply about the direct and  
128 indirect relationships of variables in their study systems (Ferraro, Sanchirico & Smith 2019).  
129 Causal inference methods aiming at controlling for confounding factors include matching (to  
130 control observable confounders) and use of panel data and synthetic controls to control for  
131 unobservable confounders, as well as instrumental variables to eliminate unobservable

132 confounders (reviewed by Law *et al.* 2017), and time series methods such as convergent  
133 cross mapping (Sugihara *et al.* 2012). Time-series data might be particularly useful because  
134 they are unidirectional implying that cause must precede effect (Dornelas *et al.* 2013).  
135 Triangulation, whereby several approaches are formally applied to the same problem, could  
136 serve as another model for increasing the reliability of causal claims (Munafo & Smith 2018).  
137 Finally, to effectively synthesize evidence from causal claims across studies, a wider adoption  
138 of systematic reviews and other structured evidence synthesis methods would allow more  
139 robust assessment of the evidence base (Pullin & Stewart 2006). In the open science era, the  
140 time is now ripe to develop models and procedures that conduct evidence synthesis based  
141 directly on open data rather than published effect sizes (Culina *et al.* 2018).

142

#### 143 [Journals, editors, and reviewers should assist in the change](#)

144 Science is not conducted in isolation in research labs, but rather represents a collective social  
145 endeavour involving many people with different roles to fill. Journals could play an  
146 important role facilitating scientific rigor of the studies that underpin real-life conservation  
147 decisions. This could partly be achieved by creating new incentives for more honest and  
148 open reporting from the research process.

149 Pre-registration of research hypothesis has been advocated (Nosek *et al.* 2018), partly to  
150 distinguish exploration and confirmation research. In the open science era, studies are  
151 increasingly based on pre-existing data, and even data that have been previously analysed  
152 and with results published in a scientific journal. This should however not discourage *a priori*  
153 hypothesis development and pre-registration (Nosek *et al.* 2018). Journal editors could  
154 facilitate this shift by applying a model where authors declare their study design and identify



155 at which stage in the process they developed their hypothesis (e.g. before or after data  
156 collection, before or after initial data analysis etc). This could include a link to the pre-  
157 registered hypothesis that might be hosted on e.g. Open Science Framework ([www.osf.io](http://www.osf.io)),  
158 and potentially an associated “open science badge” (Kidwell *et al.* 2016) as a sign of an open  
159 research practice.

160 We also encourage journal editors to more actively encourage fair valuation of case studies  
161 that mainly describe and document the state of local and global biodiversity. To  
162 accommodate this, we suggest that journals should more explicitly allocate different  
163 sections to different types of studies (exploratory, methods, confirmatory/hypothesis testing  
164 etc). This will make the publication process more transparent and facilitate more honest  
165 reporting of how the study was performed, especially reducing the incentives for *harking*,  
166 and lessen publication bias towards significant studies.

167 Finally, we propose (as a counterpart to pre-registration of hypotheses) a model where  
168 hypotheses arising from explicit exploratory research could also be registered so that they are  
169 readily available for testing in subsequent studies. Given the rise of global databases and  
170 repositories, such a model could make it feasible to track hypothesis to their source, which  
171 would allow for fair attribution of credit to those that originally proposed the hypothesis,  
172 and it would provide a clearer link between exploratory (hypothesis generating) and  
173 confirmatory (hypothesis testing) research.

174

## 175 Outlook

176 We should value the unique contributions of exploratory and confirmatory studies, but be  
177 much clearer about the fundamental differences between them. In the open science era

178 (Nosek *et al.* 2015), where more and more research is based on pre-existing (and often  
179 open) data, and where large scale studies are needed to address key conservation policy  
180 challenges, a simple plea to follow the strong inference paradigm (Platt 1964) might not be  
181 sufficient. However, current incentives that promote the presentation of studies that are by  
182 design and conduct exploratory as if they were confirmatory is a disservice to scientific  
183 progress. In applied fields like conservation biology, this will also delay progress to solve real  
184 conservation problems. The open science era has already radically improved the  
185 reproducibility of research; however, we argue that a cultural shift, involving researchers,  
186 journals, and funding bodies, is still needed towards full transparency and valuation of  
187 diverse research methods.

188

189

190 [Acknowledgement](#)

191 We are grateful to many people at our research department for fruitful discussions about  
192 this topic over the last years. EBN, DB and JDCL received funding from the Research Council  
193 of Norway.

194

195 **Data accessibility:** Data and R-scripts used to perform the randomization routines and  
196 produce figures for Box 1 is currently available here: <https://osf.io/n8fum/>

197

198 **Authors contributions:** EBN conceived the idea for this work, after discussions with DB and  
199 JDCL. EBN and DB performed the literature survey for Box 1. EBN were responsible for  
200 writing the manuscript, with inputs from JDCL and DB. All authors edited and approved the  
201 final version of the manuscript.

202    **References**

203    Baker, M. (2016) 1,500 scientists lift the lid on reproducibility. *Nature*, **New feature**.

204    Barley, S.C. & Meeuwig, J.J. (2017) The Power and the Pitfalls of Large-scale, Unreplicated Natural  
205        Experiments. *Ecosystems*, **20**, 331-339.

206    Beissinger, S.R. & Peery, M.Z. (2007) Reconstructing the historic demography of an endangered  
207        seabird. *Ecology*, **88**, 296-305.

208    Betini, G.S., Avgar, T. & Fryxell, J.M. (2017) Why are we not evaluating multiple competing  
209        hypotheses in ecology and evolution? *Royal Society Open Science*, **4**.

210    Camerer, C.F., Dreber, A., Holzmeister, F., Ho, T.H., Huber, J., Johannesson, M., Kirchler, M., Nave, G.,  
211        Nosek, B.A., Pfeiffer, T., Altmejd, A., Buttrick, N., Chan, T.Z., Chen, Y.L., Forsell, E., Gampa, A.,  
212        Heikensten, E., Hummer, L., Imai, T., Isaksson, S., Manfredi, D., Rose, J., Wagenmakers, E.J. &  
213        Wu, H. (2018) Evaluating the replicability of social science experiments in Nature and Science  
214        between 2010 and 2015. *Nature Human Behaviour*, **2**, 637-644.

215    Caughley, G. (1994) Directions in conservation Biology. *Journal of Animal Ecology*, **63**, 215-244.

216    Culina, A., Crowther, T.W., Ramakers, J.J.C., Gienapp, P. & Visser, M.E. (2018) How to do meta-  
217        analysis of open datasets. *Nature Ecology & Evolution*, **2**, 1053-1056.

218    Dornelas, M., Magurran, A.E., Buckland, S.T., Chao, A., Chazdon, R.L., Colwell, R.K., Curtis, T., Gaston,  
219        K.J., Gotelli, N.J., Kosnik, M.A., McGill, B., McCune, J.L., Morlon, H., Mumby, P.J., Øvreås, L.,  
220        Stuedeny, A. & Vellend, M. (2013) Quantifying temporal change in biodiversity: challenges and  
221        opportunities. *Proceedings of the Royal Society B: Biological Sciences*, **280**.

222    Estes, L., Elsen, P.R., Treuer, T., Ahmed, L., Caylor, K., Chang, J., Choi, J.J. & Ellis, E.C. (2018) The  
223        spatial and temporal domains of modern ecology. *Nature Ecology & Evolution*, **2**, 819-826.

224    Ferraro, P.J., Sanchirico, J.N. & Smith, M.D. (2019) Causal inference in coupled human and natural  
225        systems. *Proceedings of the National Academy of Sciences of the United States of America*,  
226        **116**, 5311-5318.

227    Fraser, H., Parker, T., Nakagawa, S., Barnett, A. & Fidler, F. (2018) Questionable research practices in  
228        ecology and evolution. *Plos One*, **13**.

229    Gaillard, J.M., Duncan, P., Delorme, D., van Laere, G., Pettorelli, N., Maillard, D. & Renaud, G. (2003)  
230        Effects of hurricane Lothar on the population dynamics of European roe deer. *Journal of*  
231        *Wildlife Management*, **67**, 767-773.

232    Hayward, M.W., Edwards, S., Fancourt, B.A., Linnell, J.D.C. & Nilsen, E.B. (2019) Top-down control of  
233        ecosystems and the case for rewilding: does it all add up? *Rewilding* (eds J.T. du Toit, N.  
234        Pettorelli & S.M. Durant), pp. 325-354. Cambridge University Press, Cambridge.

235    Heuschele, J., Ekvall, M.T., Mariani, P. & Lindemann, C. (2017) On the missing link in ecology:  
236        improving communication between modellers and experimentalists. *Oikos*, **126**, 1071-1077.

237    Houlahan, J.E., McKinney, S.T., Anderson, T.M. & McGill, B.J. (2017) The priority of prediction in  
238        ecological understanding. *Oikos*, **126**, 1-7.

239    Ioannidis, J.P.A., Munafò, M.R., Fusar-Poli, P., Nosek, B.A. & David, S.P. (2014) Publication and other  
240        reporting biases in cognitive sciences: detection, prevalence, and prevention. *Trends in*  
241        *Cognitive Sciences*, **18**, 235-241.

242    Kidwell, M.C., Lazarević, L.B., Baranski, E., Hardwicke, T.E., Piechowski, S., Falkenberg, L.-S., Kennett,  
243        C., Slowik, A., Sonnleitner, C., Hess-Holden, C., Errington, T.M., Fiedler, S. & Nosek, B.A.  
244        (2016) Badges to Acknowledge Open Practices: A Simple, Low-Cost, Effective Method for  
245        Increasing Transparency. *PLOS Biology*, **14**, e1002456.

246    Krebs, C.J., Boutin, S. & Boonstra, R. (1995) *Ecosystem dynamics of the boreal forest: The Kluane*  
247        *project*. Oxford University Press, Oxford.

248    Law, E.A., Ferraro, P.J., Arcese, P., Bryan, B.A., Davis, K., Gordon, A., Holden, M.H., Iacona, G.,  
249        Martinez, R.M., McAlpine, C.A., Rhodes, J.R., Sze, J.S. & Wilson, K.A. (2017) Projecting the  
250        performance of conservation interventions. *Biological Conservation*, **215**, 142-151.

251 Lehtikoinen, A., Brotons, L., Calladine, J., Campedelli, T., Escandell, V., Flousek, J., Grueneberg, C.,  
252 Haas, F., Harris, S., Herrando, S., Husby, M., Jiguet, F., Kalas, J.A., Lindstrom, A., Lorrilliere, R.,  
253 Molina, B., Pladevall, C., Calvi, G., Sattler, T., Schmid, H., Sirkia, P.M., Teufelbauer, N. &  
254 Trautmann, S. (2019) Declining population trends of European mountain birds. *Global*  
255 *Change Biology*, **25**, 577-588.

256 Mazor, T., Doropoulos, C., Schwarzmuller, F., Gladish, D.W., Kumaran, N., Merkel, K., Di Marco, M. &  
257 Gagic, V. (2018) Global mismatch of policy and research on drivers of biodiversity loss.  
258 *Nature Ecology & Evolution*, **2**, 1071-+.

259 Millennium-Ecosystem-Assessment (2005) *Ecosystems and Human Well-being: Synthesis*. Island  
260 Press, Washington, DC.

261 Miller, D.A.W., Pacifici, K., Sanderlin, J.S. & Reich, B.J. (2019) The recent past and promising future for  
262 data integration methods to estimate species' distributions. **10**, 22-37.

263 Munafo, M.R. & Smith, G.D. (2018) Repeating experiments is not enough. *Nature*, **553**, 399-401.

264 Nilsen, E.B. & Strand, O. (2018) Integrating data from multiple sources for insights into demographic  
265 processes: Simulation studies and proof of concept for hierarchical change-in-ratio models.  
266 *PLoS ONE*, **13**, e0194566.

267 Nosek, B.A., Alter, G., Banks, G.C., Borsboom, D., Bowman, S.D., Breckler, S.J., Buck, S., Chambers,  
268 C.D., Chin, G., Christensen, G., Contestabile, M., Dafoe, A., Eich, E., Freese, J., Glennerster, R.,  
269 Goroff, D., Green, D.P., Hesse, B., Humphreys, M., Ishiyama, J., Karlan, D., Kraut, A., Lupia, A.,  
270 Mabry, P., Madon, T., Malhotra, N., Mayo-Wilson, E., McNutt, M., Miguel, E., Paluck, E.L.,  
271 Simonsohn, U., Soderberg, C., Spellman, B.A., Turitto, J., VandenBos, G., Vazire, S.,  
272 Wagenmakers, E.J., Wilson, R. & Yarkoni, T. (2015) Promoting an open research culture. **348**,  
273 1422-1425.

274 Nosek, B.A. & Collaboration, O.S. (2015) Estimating the reproducibility of psychological science. **349**,  
275 aac4716.

276 Nosek, B.A., Ebersole, C.R., DeHaven, A.C. & Mellor, D.T. (2018) The preregistration revolution.  
277 *Proceedings of the National Academy of Sciences of the United States of America*, **115**, 2600-  
278 2606.

279 Pereira, H.M., Ferrier, S., Walters, M., Geller, G.N., Jongman, R.H.G., Scholes, R.J., Bruford, M.W.,  
280 Brummitt, N., Butchart, S.H.M., Cardoso, A.C., Coops, N.C., Dulloo, E., Faith, D.P., Freyhof, J.,  
281 Gregory, R.D., Heip, C., Höft, R., Hurtt, G., Jetz, W., Karp, D.S., McGeoch, M.A., Obura, D.,  
282 Onoda, Y., Pettorelli, N., Reyers, B., Sayre, R., Scharlemann, J.P.W., Stuart, S.N., Turak, E.,  
283 Walpole, M. & Wegmann, M. (2013) Essential Biodiversity Variables. **339**, 277-278.

284 Platt, J.R. (1964) Strong Inference. *Science*, **146**, 347-353.

285 Pullin, A.S. & Stewart, G.B. (2006) Guidelines for systematic review in conservation and  
286 environmental management. *Conservation Biology*, **20**, 1647-1656.

287 Sells, S.N., Bassing, S.B., Barker, K.J., Forshee, S.C., Keever, A.C., Goerz, J.W. & Mitchell, M.S. (2018)  
288 Increased scientific rigor will improve reliability of research and effectiveness of  
289 management. *Journal of Wildlife Management*, **82**, 485-494.

290 Serrouya, R., Seip, D.R., Hervieux, D., McLellan, B.N., Mcnay, R.S., Steenweg, R., Heard, D.C.,  
291 Hebblewhite, M., Gillingham, M. & Boutin, S. (2019) Saving endangered species using  
292 adaptive management. *Proceedings of the National Academy of Sciences of the United States*  
293 *of America*, **116**, 6181-6186.

294 Soulé, M.E. (1996) *Conservation Biology: The Science of Scarcity and Diversity*. Sinauer & Associates,  
295 Sunderland, MA.

296 Sugihara, G., May, R., Ye, H., Hsieh, C.H., Deyle, E., Fogarty, M. & Munch, S. (2012) Detecting  
297 Causality in Complex Ecosystems. *Science*, **338**, 496-500.

298 Wiik, E., d'Annunzio, R., Pynegar, E., Crespo, D., Asquith, N. & Jones, J.P.G. (2019) Experimental  
299 evaluation of the impact of a payment for environmental services program on deforestation.  
300 **1**, e8.

301 Wilkinson, M.D., Dumontier, M., Aalbersberg, I.J., Appleton, G., Axton, M., Baak, A., Blomberg, N.,  
302 Boiten, J.-W., da Silva Santos, L.B., Bourne, P.E., Bouwman, J., Brookes, A.J., Clark, T., Crosas,

303 M., Dillo, I., Dumon, O., Edmunds, S., Evelo, C.T., Finkers, R., Gonzalez-Beltran, A., Gray,  
304 A.J.G., Groth, P., Goble, C., Grethe, J.S., Heringa, J., 't Hoen, P.A.C., Hooft, R., Kuhn, T., Kok, R.,  
305 Kok, J., Lusher, S.J., Martone, M.E., Mons, A., Packer, A.L., Persson, B., Rocca-Serra, P., Roos,  
306 M., van Schaik, R., Sansone, S.-A., Schultes, E., Sengstag, T., Slater, T., Strawn, G., Swertz,  
307 M.A., Thompson, M., van der Lei, J., van Mulligen, E., Velterop, J., Waagmeester, A.,  
308 Wittenburg, P., Wolstencroft, K., Zhao, J. & Mons, B. (2016) The FAIR Guiding Principles for  
309 scientific data management and stewardship. *Scientific Data*, **3**, 160018.

310

311

312

## 313 Box 1: State of conservation biology as a science

314 In a seminal paper from 1994, G. Caughley (Caughley 1994) was concerned that parts of  
315 conservation biology (the branch concerned with declining populations) had a very thin  
316 theoretical basis, was carried out mainly as a series of case studies, and therefore often had  
317 limited generalisable value. In line with many other philosophers of science, Caughley  
318 suggested that much more rapid progress would be made if conservation biologists applied  
319 the strong inference paradigm (sensu Platt 1964) when designing and conducting research.

320 To gain a rapid insight into the current state of affairs in the scientific conservation  
321 literature, we randomly sampled 160 papers published in eight journals covering  
322 conservation biology, applied ecology and wildlife management. We only included studies  
323 from terrestrial ecology, that were data-driven (i.e. not reviews or pure simulation studies),  
324 that presented the results from at least one statistical test, that presented original data or  
325 data from literature surveys, and focused on conservation biology. From these studies, we  
326 assessed i) to which extent one or more clearly stated hypotheses were presented in the  
327 introduction, ii) whether there were multiple competing hypothesis and, iii) whether they  
328 applied an experimental study design. In addition, we extracted the number of citations  
329 registered by Web of Science. A more comprehensive description of the inclusion criteria  
330 and data extraction procedures can be found in **Appendix S1**.

331 Based on our sample of research papers, it seems that clearly stating a research hypothesis  
332 in the introduction is surprisingly rare in the literature (**Fig 1a**). Overall, only about 19% of  
333 the studies presented clear hypothesis, whereas about 26% presented what we term  
334 “implied hypotheses” or “partly”, where the hypothesis could be inferred from the text but  
335 was not presented clearly. After removing articles mainly focusing on methods development,

336 the corresponding proportions were 22% (clear hypothesis) and 28% (implied), respectively.  
337 Presenting multiple competing hypothesis, as described in the original presentation of the  
338 strong inference paradigm (Platt 1964) is even rarer, and only 2 of the studies we reviewed.  
339 Another hallmark of science is the use of well planned, randomized and replicated  
340 experimental manipulation to test for causal relationships (Platt 1964; Caughley 1994).  
341 Based on our review, however, the use of full experimental designs are rare, and only 12% of  
342 the studies we reviewed were based on randomized controlled experimental designs. In  
343 addition, 15% of the studies in our sample included Before-After-Control-Impact (BACI) or  
344 Quasi-experimental protocols. The majority of the randomized controlled experiments were  
345 performed on a local spatial scale (**Fig 1b**), although a few studies presented landscape scale  
346 experiments. In our sample, local scale studies in general received less attention in the  
347 literature compared to studies spanning larges spatial scales (**Fig 1b**).

348

## 349 Figure Legends

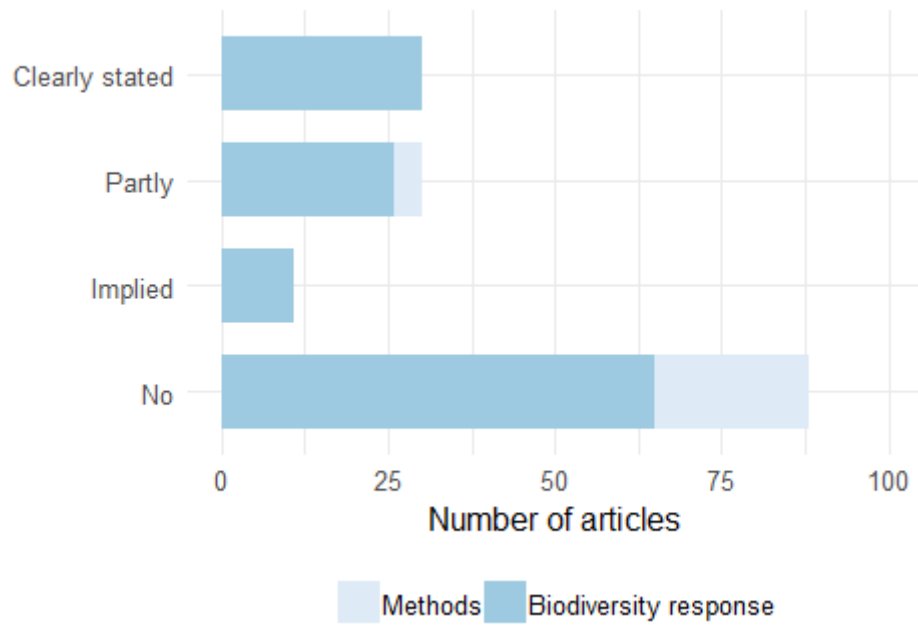
350 **Figure 1.** In **a)** the proportion of articles that reported clear hypotheses, implied or partly indicated  
351 hypotheses that were tested, and articles that did not present hypotheses. In **b)** the proportion of  
352 articles that used experimental, quasi-experimental/BACI or no experimental designs are matched  
353 with the corresponding spatial scales of the studies. The size of the circles indicates the number of  
354 studies. The colour key indicates citation rates (mean annual number of citations since the year of  
355 publication).

356



357 Figures

358



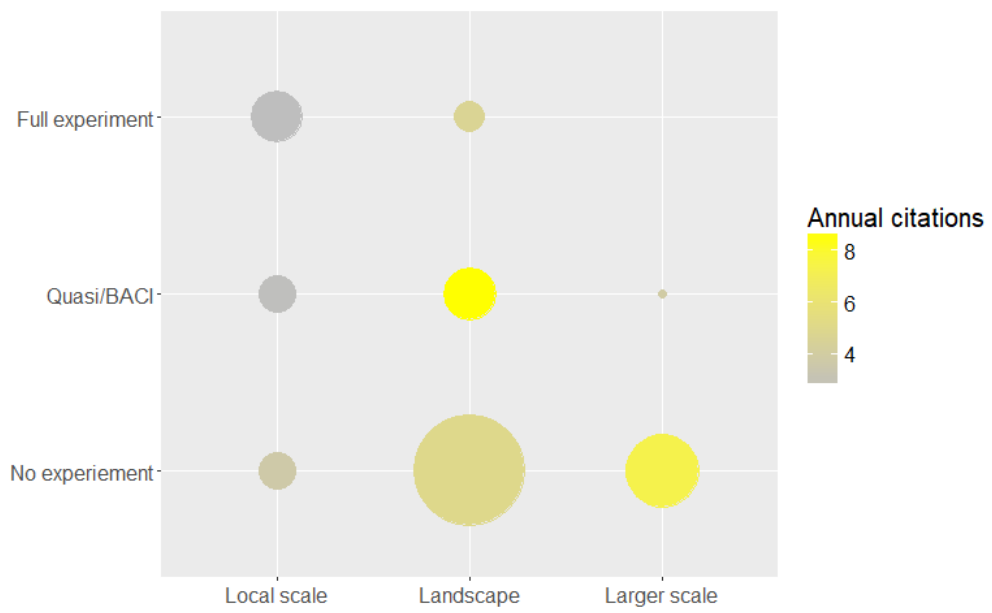
359

360 **Fig 1a**

361

362

363



364

365 **Fig 1b**