

1 Assessing Transparency and Reproducibility in Invasion Science

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22 **Abstract**

23 Policymakers and practitioners overseeing invasive species management depend on reliable
24 research for guidance. Transparency and reproducibility are core features of reliable research,
25 and prerequisites for successful study replication, but are evidently lacking in many science
26 disciplines. Whether this shortfall characterizes invasion science remains unknown. We
27 evaluated a sample of invasion science studies for their adherence to practices that enhance
28 transparency and reproducibility, such as making data and code available, and explicitly
29 considering statistical power. Our evaluation focused on published studies concerning two plant
30 species invasive to riparian ecosystems in British Columbia, Canada (*Elaeagnus angustifolia*
31 and *Phalaris arundinacea*), as these contributed to a broader systematic mapping initiative. Our
32 systematic literature search yielded 746 studies, of which 45 met our predefined inclusion
33 criteria (relevance rate = 6%). We assessed each study against a 14-item checklist (motivated
34 by the Transparency and Openness Promotions guidelines) and a corresponding scoring rubric.
35 On average, studies achieved a score of 26%, with no studies addressing statistical power, pre-
36 registering their plans, and few making data or code publicly available. There is a clear need
37 and opportunity for improving the transparency and reproducibility of invasion science research.
38 We refer researchers to resources aimed at improving research practices, and discuss two
39 practices that are especially important in the context of applied invasion science: power analysis
40 and sharing data and code. We echo recent calls for educational and research institutions to
41 expand access to training in open science, and urge policymakers and practitioners to consider
42 transparency and reproducibility when seeking guidance from invasion science research.

43
44 Keywords: Invasion science, Reproducibility, Open Science, Non-native species, Invasive
45 species, Impacts, Riparian ecosystems, Foreshore ecosystems, British Columbia, Reed
46 Canarygrass, Russian Olive

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49 Introduction

50 Globally, biological invasions are estimated to cost over 423 billion USD each year, a number
51 that has quadrupled every decade since 1970 (Roy et al., 2024). Ideally, policymakers and
52 managers tasked with addressing the threats and impacts of invasive species base their
53 decisions and actions on reliable research, thus maximizing potential success, and minimizing
54 waste in time, effort, and money. “Reliable” means that the research adheres to foundational
55 principles of scientific inquiry, i.e. it is reproducible (definition in Box 1), and that independent
56 efforts to replicate the findings would be successful. Worryingly, evidence suggests that a
57 sizeable fraction of published research, across a variety of disciplines including ecology (Kelly,
58 2019), is irreproducible (Baker, 2016), and replication efforts in fields such as psychology (Open
59 Science Collaboration, 2015) and medicine (Begley & Ioannidis, 2015) have met with limited
60 success.

61 A variety of shortfalls can render scientific research irreproducible (Parker, Forstmeier, et
62 al., 2016), including failing to fully and transparently report all procedures (e.g. details of
63 sampling and/or experimental design and statistical analyses, Davis & Kay, 2023; Parker et al.,
64 2016), and failing to make code and data available and useable (Culina et al., 2020; Popovic,
65 Mason, Drobniak, Marques, Potts, Joo, Altwegg, Burns, McCarthy, Johnston, Nakagawa,
66 McMillan, Devarajan, Taggart, Wunderlich, Mair, Martínez-Lanfranco, et al., 2024; Riordan-
67 Short et al., 2023; Roche et al., 2015a). Addressing such shortfalls does not, however,
68 guarantee that findings will be replicated. For example, a study may be reproducible thanks to
69 thorough and transparent reporting, but if it suffers from low statistical power its findings are less
70 likely to be replicated (Button et al., 2013; Yang et al., 2023). It is recommended that studies
71 achieve a power of at least 80% (Button et al., 2013), but a recent estimate of the average
72 power in primary ecology and evolutionary biology studies was 15% (Yang et al., 2023), slightly
73 lower than other disciplines (e.g. 20% in medical sciences, Lamberink et al., 2018).

74 Low statistical power in invasion science can reduce the likelihood of detecting impacts
75 of invasive species. Previous work evaluated 31 analyses testing for the ecological impacts of
76 nonnative species, finding that 97% (n=30) had insufficient power to detect if the reported
77 impact (or lack thereof) existed (or not, Davidson & Hewitt, 2014). Less well-known
78 consequences of low statistical power, and previously overlooked (Davidson & Hewitt, 2014),
79 are increases in the rate of false positives (Type-I errors), errors in effect magnitudes (Type M)
80 and errors in effect signs (Type S, (Button et al., 2013; Parker et al., 2018; Parker & Yang,
81 2023). When combined with the potential influences of P-hacking and HARKing (i.e.
82 hypothesizing after the results are known, Munafò et al., 2017), it is conceivable that a
83 substantial proportion of published studies in invasion science present findings that exaggerate
84 impacts or even misrepresent the true direction of effects.

85 Here we report the results of an evaluation of invasion science research with respect to
86 its adherence to practices that enhance transparency and reproducibility. We were motivated to
87 undertake this assessment while conducting a separate project focused on producing a
88 systematic map of the evidence on the impacts of a selection of plant species invasive to
89 riparian ecosystems in British Columbia (Mologni et al., 2023a). While compiling the studies, we
90 found substantial variation in reporting practices, including among studies on the same species.
91 We therefore decided to more formally evaluate a subset of the publications with reference to a
92 checklist of practices that are recommended to maximize transparency and reproducibility.

93

94 **Methods**

95 Our assessment focused on studies concerning two plant species invasive to riparian
96 ecosystems in British Columbia, Canada: *Elaeagnus angustifolia* (Russian Olive) and *Phalaris*
97 *arundinacea* (Reed Canarygrass, thereafter RO and RCG respectively). We systematically
98 identified all studies investigating their impacts on riparian ecosystems within North America

99 (excluding Mexico), following the ROSES reporting standards for systematic maps and reviews
100 (Haddaway et al., 2018). A detailed description can be found in the protocol of our systematic
101 map (Mologni et al., 2023a, 2023b). For the present study, we applied an additional filter: we
102 removed studies published before the year 2000. This reflects a balance between excluding
103 years in which there was limited awareness around reproducibility in ecology, with the need for
104 a sufficient time period over which to reveal any obvious trends in reporting practices. We
105 assume that the literature we reviewed is representative of invasion science research in
106 general.

107 To appraise the level of adherence to practices that enhance transparency and
108 reproducibility, we assessed each article against a checklist and corresponding scoring rubric.
109 Our checklist is a streamlined version of a checklist developed previously in our lab (Kast et al.,
110 2023 see Appendix 1), which aimed to balance the operational simplicity of the “transparency
111 and openness promotion” guidelines developed by the Center for Open Science
112 (<https://www.cos.io/initiatives/top-guidelines>) with the more demanding evaluation steps of the
113 checklist for peer reviewers developed by (Parker et al., 2018). It was also informed by the
114 “transparency checklist” of Aczel et al. (2020).

115 The checklist includes 14 items, and to each we assigned a score of “0” if not
116 addressed, “1” if partially addressed, “2” if addressed or “n/a” if not applicable. We calculated
117 the final study score by summing the item scores and dividing by the number of applicable
118 questions. All items are equally weighted. The final score was halved and multiplied by 100 to
119 convert it to a percentage.

120 The items fall into four categories and include the following questions:

- 121 - Pre-registration (see Discussion): was the study pre-registered and were deviations from
122 planned methods described?

- 123 - Materials availability: were availability statements included? Were data and code stored
124 in openly accessible repositories and links and licences provided? Was the description
125 of the experimental protocol sufficiently comprehensive to allow for replication?
126 - Study design: was sample size determined (e.g. with a priori power analysis) and
127 randomization described? If applicable, as the study authorized (i.e. a relevant permit
128 was obtained if required)?
129 - Data analysis: were statistical methods justified and associated assumptions checked?

130 The full list and description of items can be found in the supplementary material (Appendix 1).

131 We calculated the overall, by-year and by-item average scores. We identified areas of
132 weakness and strength using average scores by item. We arbitrarily categorized scores below
133 20% as “*overlooked*”, between 20 and 80% as “*in progress*” and above 80% as “*addressed*”.

134 We did not complete a pre-registration for this study because it does not include any
135 formal statistical analyses or inferences, and we initiated this study after some of the literature
136 search and evaluation activities had already been completed as part of a concurrent literature
137 review (Mologni et al., 2023a).

138

139 **Results**

140 Our systematic literature search yielded a total of 746 studies, 335 for RO and 411 for CRG
141 (Figure 1). The majority were extracted from Web of Science (n = 405), followed CABI Invasive
142 Species Compendium (n = 175). Two reviews were identified for RO and one for CRG. **Katz &**
143 **Shafroth (2003)** focused on RO and all references were screened (n=162). **Lalk et al. (2021)**
144 investigated the impacts of several woody species and we selected only studies referring to RO
145 (n=6). **Zedler & Kercher (2004)** focused on wetlands and only references regarding CRG were
146 extracted (n = 7). After removing studies published before the year 2000, we obtained 46

147 studies. One appeared in both searches and was thus excluded. The final list contained 45
148 studies, with an overall relevance rate (i.e. the percentage of studies included after screening
149 and removing duplicates, Ridley et al., 2022) of 6%.

150 The average score across all 45 studies was 26% and ranged from 11 to 50%. The
151 number of publications and scores increased slightly over time (Figure 2). Ten items averaged a
152 score below 20% and were categorized as “*overlooked*” (Figure 3). Four items scored 0 across
153 all studies: pre-registration, code link, code licence and sample size determination. The other 6
154 items averaged between 0 and 20%: the material availability statement (score = 5%), data
155 availability statement (16%), data link (7%), data licence (2%), code availability statement (3%),
156 and study authorization (19%, Figure 3). Two items were between 20% and 80%: randomization
157 (40%) and assumptions (71%) and were categorized as “*in progress*” (Figure 3). Two items
158 were above 80% and were experimental protocol (96%) and statistics (89%), and marked as
159 “*addressed*” (Figure 3).

160

161 **Discussion**

162 The cost of managing biological invasions has quadrupled every decade since the 1970s, with
163 no signs of slowing down (Roy et al., 2024). To optimize the allocation of limited financial and
164 personnel resources, the research informing managers and policy-makers must be transparent,
165 reproducible, and ideally replicable. Our findings clearly show that there is much room for
166 improvement. On average, only a quarter of the reproducibility criteria were satisfactorily
167 addressed across 45 studies in invasion science. The majority were addressed only
168 sporadically. This is consistent with previous findings in psychology (Open Science
169 Collaboration, 2015), medicine (Begley & Ioannidis, 2015), and ecology (Parker, Forstmeier, et
170 al., 2016; Roche et al., 2015a).

171 Two items were categorized as “*addressed*”: describing experimental design and
172 statistics. The importance of reporting detailed experimental designs and statistical analyses
173 has long been recognized and thus generally implemented in scientific research. We did not
174 evaluate the appropriateness of the experimental design and statistical analyses, as this was
175 beyond the scope of our study. Two items were categorized as “*in progress*”: on average, more
176 than two-thirds of the studies sufficiently described and justified meeting assumptions of
177 statistical tests. More than a third stated if randomization occurred and described why or why
178 not.

179 Most items (10 out of 14) scored below 20% and were deemed as “*overlooked*”. Material
180 availability statements were rarely provided, highlighting the inability to recover original
181 specimens and samples. This includes, for example, no mention of the existence of voucher
182 specimens when these are relevant. Surprisingly, less than a third of the studies provided a data
183 availability statement. Of these, the majority (n=9) provided data as supplementary information
184 and only three a proper data availability statement, two linked to a repository (Price et al., 2018;
185 Rowe et al., 2019). Price et al. (2018) was the only study that provided a data licence too.
186 These findings align with a previous review of data availability statements (Federer et al., 2018)
187 and contrast with current journal policies, which generally require data availability statements
188 and encourage storing and sharing data in a repository. Code availability statements were
189 largely absent too, despite code availability being identified as one of the principal barriers to
190 reproducibility in science (Kambouris et al., 2024). Two studies included a code availability
191 statement, but only one provided access to code; however, shared as supplementary
192 information. No study provided a link and licence to code.

193 No study conducted power analyses to pre-determine the sample size. Low statistical
194 power increases the chances of missing (Type II error, (Davidson & Hewitt, 2014) and
195 overestimating impacts (Type M error, Parker & Yang, 2023; Yang et al., 2022). Similarly, no
196 study completed a pre-registration. Pre-registration reduces bias and encourages thorough pre-

197 planning (Yang et al., 2023), including power analyses. Low average scores were also identified
198 for study authorization.

199 Open science policies of the main journals in invasion science vary considerably (i.e.
200 Biological Invasions, Neobiota, Invasive Plant Science and Management, and Wetlands, the
201 latter being the most common journal in our bibliographic sample). These journals consistently
202 encourage data sharing, sometimes in specialized repositories; however, this is not mandatory.
203 Biological Invasions is the only journal that requires a data availability statement, albeit this does
204 not require data themselves to be made available. Invasive Plant Science and Management
205 encourages best practices in reporting methodologies, while Neobiota strongly encourages
206 depositing methods and “protocols” in a repository (we consider a protocol akin to a pre-
207 registration). Yet again, these policies are not mandatory. On the other hand, Springer journals
208 (Biological Invasions and Wetlands) require sharing voucher specimens and identifiers.
209 Wetlands also requires the submission of specific data and materials (i.e. proteins, DNA, and
210 RNA sequences) to appropriate repositories. All in all, policies vary greatly across journals, are
211 typically non-mandatory and cover only a fraction of open science good practices.

212 Other journals in ecology are adopting stronger open science policies. For instance,
213 some journals allow for the automatic publication of studies as preprints or allow direct transfer
214 from preprint repositories upon submission (Proceedings of the Royal Society B). Other journals
215 have partnerships with data repositories (Journal of Biogeography) or require publicly
216 accessible archiving of the data (Ecology). In the last few years, data availability statements
217 have increasingly become mandatory, and many journals now require authors to share their
218 code too (Proceedings of the Royal Society B), to the point that the American Naturalist now
219 enlists data editors to evaluate data and code sharing of each manuscript submitted to the
220 journal. Finally, Environmental Evidence requires registering a protocol (i.e. preregistration) in
221 an open-access repository. Still, consistency in journal policies concerning open science good
222 practices is lacking.

223 We encourage invasion science researchers to consult the commentary by O’Dea et al.
224 (2021), who provide a thoughtful assessment of the ways researchers, publishers, and funders
225 can facilitate progress towards more open, transparent, and reliable research, and also
226 underscore the importance of adjusting incentives and removing barriers for researchers (Baker,
227 2016; Strømme et al., 2022). We also highlight the contributions of Filazzola and Cahill (2021)
228 and (Cooke et al., 2017), the former providing insightful discussion about study design and
229 different forms of “replication” in ecology, and the latter presenting cogent advice on
230 experimental design for informing conservation ecology - which has clear parallels to invasion
231 science. Below we highlight (i) addressing statistical power and (ii) sharing code and data as
232 key steps within the context of invasion science research, which often seeks to inform
233 management policy and allocation of resources.

234 An adequate statistical power (e.g. a goal of 80% power) reduces the chances of
235 missing and overestimating impacts. Although individual research groups often lack funding and
236 resources to achieve adequate statistical power, it is nevertheless worthwhile to undertake a
237 prospective power analysis and clearly report its outcomes. If the desired power can be
238 achieved, then communicating this will strengthen confidence in the study outcomes.
239 Acknowledging that a study falls short of the desired power is equally important and beneficial,
240 and this acknowledgment can readily be included in the standard “study limitations” section of
241 the Discussion. More generally, we suggest a section on statistical power should be routine in
242 every invasion science study, and ideally communicate what implications the achieved power
243 (be it lower than desired or not) may have for their study. Doing so will help raise awareness
244 among scientists and practitioners about the critical importance of statistical power, and how, for
245 example, it influences the risk of missing, overestimating, or underestimating impacts (Davidson
246 & Hewitt, 2014; Parker & Yang, 2023). It is noteworthy that as early as the 1990s the Journal of
247 Wildlife Management required authors to address statistical power, recognizing its importance to
248 management-related research.

249 Given that individual ecological studies rarely achieve the desired statistical power
250 (Davidson & Hewitt, 2014; Yang et al., 2022), it is especially important for researchers to make
251 their data openly available and useable (when ethically reasonable to do so), such that they can
252 contribute to meta-analyses. When the desired power cannot be achieved due to resource
253 constraints (as is commonplace), coordinated distributed experiments offer a solution by
254 combining the efforts of multiple teams (Filazzola & Cahill Jr, 2021; Fraser et al., 2013; Yahdjian
255 et al., 2021). Such collaborative approaches are becoming increasingly common in invasion
256 science and ecology in general (e.g. Knollová et al., 2024; van Kleunen et al., 2019) and should
257 be prioritized by funding agencies (Packer et al., 2017).

258 Invasive species managers and policymakers rely on research outputs to inform their
259 decisions and actions. Our findings highlight an opportunity and need to improve the
260 transparency, openness, and reproducibility of invasion science research. Ultimately,
261 addressing this challenge in invasion science will help ensure resources are allocated optimally,
262 and in ways that maximize success. For researchers, checklists (e.g. Aczel et al., 2020; Parker
263 et al., 2016), recent guidelines (Popovic, Mason, Drobniak, Marques, Potts, Joo, Altwegg,
264 Burns, McCarthy, Johnston, Nakagawa, McMillan, Devarajan, Taggart, Wunderlich, Mair,
265 Martínez-Lanfranco, et al., 2024), and example workflows (e.g. Pither, 2024) can facilitate
266 progress toward better practices, and the Society for Open, Reliable, and Transparent Ecology
267 and Evolutionary Ecology (SORTEE) provides a list of helpful resources
268 (https://www.sortee.org/other_resources/). We highlight Roche et al. (2015) and (Abdill et al.,
269 2024) for advice on sharing data and code, respectively. Lastly, we emphasize that adapting
270 research protocols towards the ideal takes time, and researchers should embrace and celebrate
271 every advance made, however small.

272

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446

447 **Box 1** Definitions of reproducible and replicable according to the National Academies of
448 Sciences, Engineering, and Medicine (2019) and Nosek et al. (2022)

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450 **Reproducible:** The ability to independently re-do the research using the same data and
451 materials and obtain the same results.

452

453 **Replicable:** The ability to independently re-do the research using new materials and data and
454 obtain consistent results.

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456

457 **Figure Captions**

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459 **Figure 1** Screening process following ROSES reporting standards (Haddaway et al.,
460 2018). 'Records extracted from bibliographic searches' refers to studies extracted from WOS,
461 CABI and reviews identified from those sources. "Extracted from other sources" refers to studies
462 not identified through the systematic search of bibliographic databases.

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464 **Figure 2** Number of studies (grey bars) and average score (black line) by publication year.

465

466 **Figure 3** Plots displaying average scores by item category. Colours indicate *overlooked*
467 (white, < 0.2), *in progress* (grey, 0.2-0.8), and *addressed* (black, >0.8) items.

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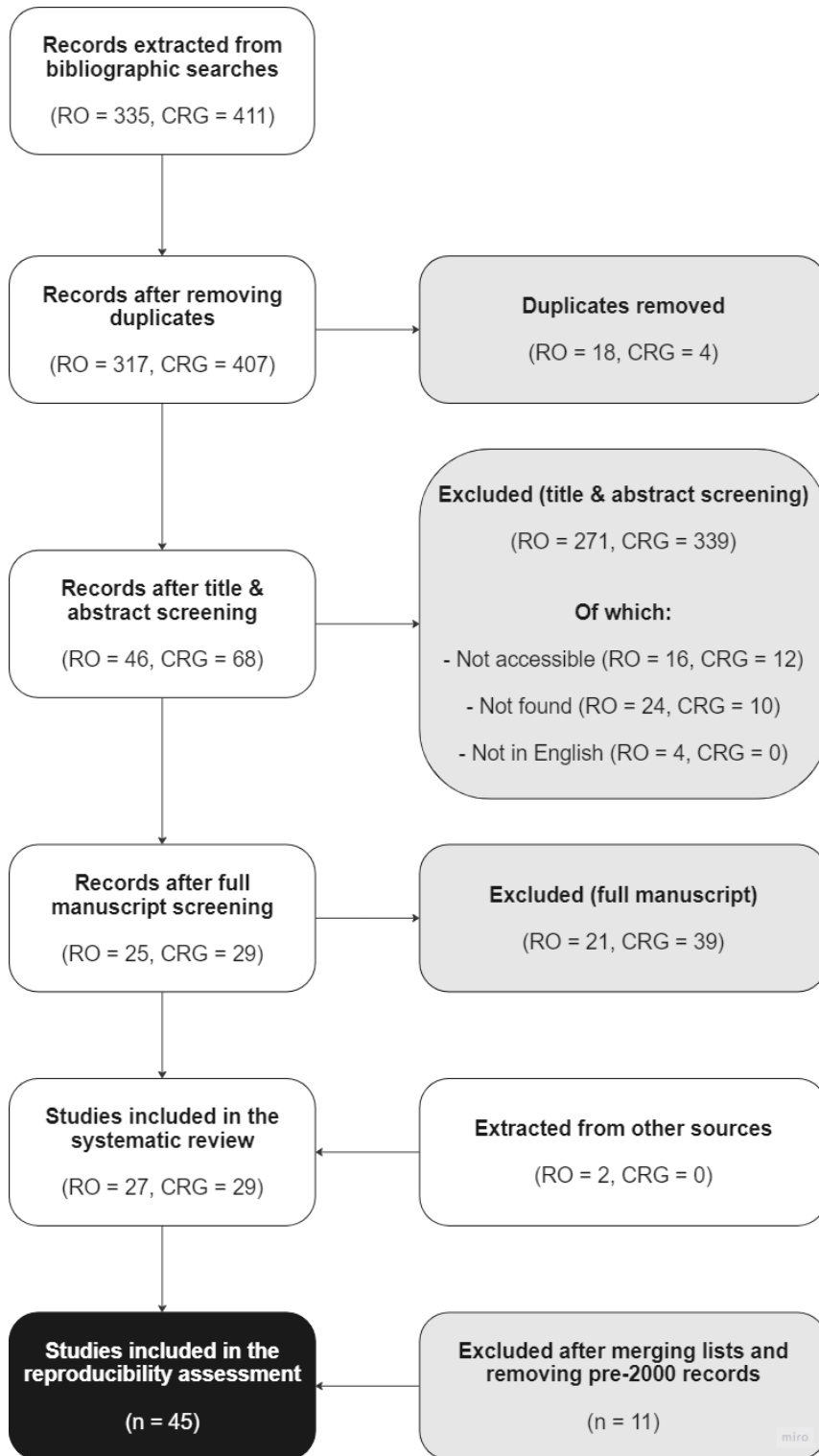
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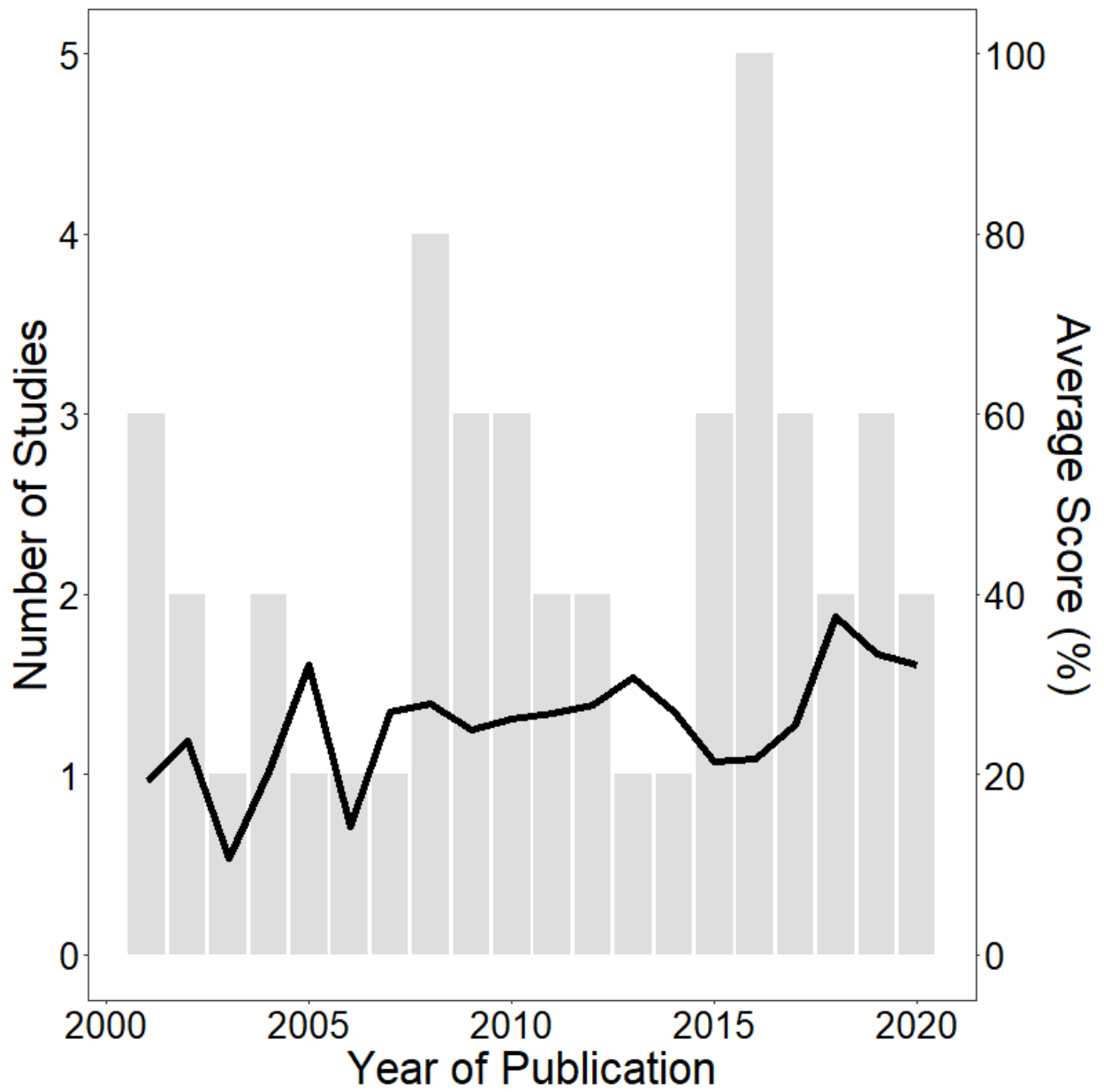
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479 **Figure 1**
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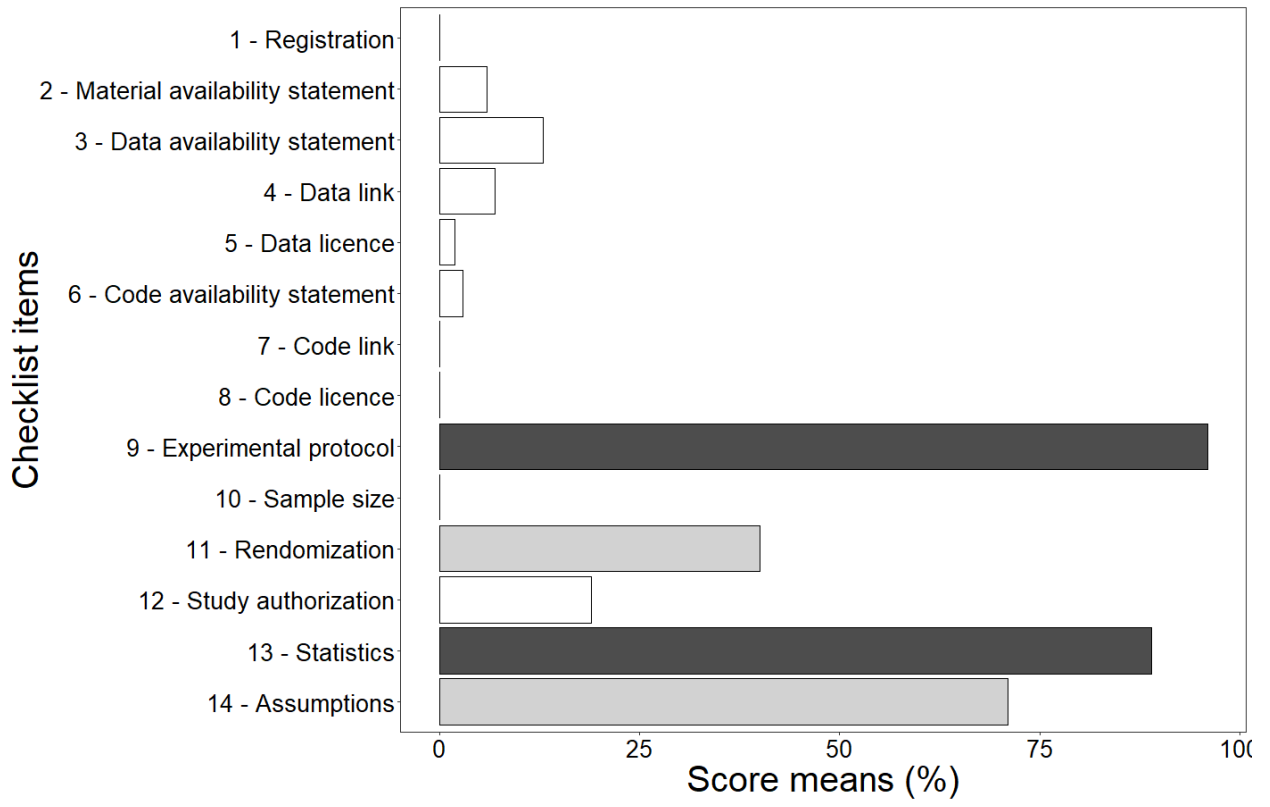
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485 Figure 2
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498 **Figure 3**
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504 **Data availability statement**

505 Data and code are available at <https://doi.org/10.5281/zenodo.14288882>.
506

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515

516 **Competing interests**

517 The authors declare that they have no competing interests.

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537 **Supplementary material**

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539 **Appendix 1** - Full scoring rubric.

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541 **Appendix 2** - List of studies included in this article and associated scores by rubric item.

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543 **Appendix 3** - Supplementary information

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580 **Appendix 1 – Checklist of Open Science Best Practices**

581 **Scoring**

582 0 = Not completed

583 1 = Partially completed (does not complete all aspects of a given criterion)

584 2 = Fully completed

585 NA = This criterion is not relevant

586

587

588 **Criteria**

589 **Pre-Registration**

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591 1. Pre-Registration: Was the study pre-registered? Are protocol deviations/changes from
592 the pre-registration fully described (e.g. experimental procedures)? (1 point each
593 question)

594

595 **Data, Methods, and Materials Availability**

596 Note: A proper availability statement is a dedicated section that tries to improve the
597 reproducibility of a manuscript by stating what data/code/materials were used during the study
598 and where said data/code/materials can be found/obtained. Availability statements almost
599 always have a dedicated header.

600 Note: The “code availability statement” refers to newly generated custom computer code (or the
601 software or mathematical algorithm or LCA models).

602

603 2. Material_Availability¹: Is there a materials availability statement (materials to include in a
604 material availability statement are: plants, microbe strains, cell lines, primary cell
605 cultures, and primers, as well as all unique chemical and biological materials, including
606 commercially and non-commercially obtained materials, newly created materials, and
607 unique specimens)?

608 a. Sequence_Availability: Is there an accession number in repository or a supplier
609 name, catalog number, clone number, or RRID for DNA/RNA sequences? Is
610 metadata (e.g. hosts, source, collection location, etc.) available for DNA/RNA
611 sequences?

612 b. Plant_Availability: If plants were used, are the species, strain, ecotype, cultivar
613 and source (including location for collected wild specimens) provided where
614 relevant? Is there a unique accession number (if available)?

615

616 3. Data_Availability: Is there a data availability statement?

¹ Points are given even if material availability is discussed in the text, rather than as a separate section.

- 617
618 4. Data_Link: Is there an accession number or DOI or URL (data)?
619
620 5. Data_License: If there are newly created datasets, are licensing details for said datasets
621 provided?
622
623 6. Code_Availability: Is there a code availability statement (Often included in data
624 availability statement)?
625
626 7. Code_Link: Is there an accession number or DOI or URL (code)?
627
628 8. Code_License: If there is newly created code, are licensing details for said code
629 provided?
630
631 9. Experimental_Protocol: Were detailed step-by-step experimental protocols made
632 available to allow for replication? Are experimental protocols sufficiently described to
633 allow for replication?

634 Experimental study design

- 635 10. Sample_Size: Is there a description of sample size/replicate determination (e.g., a priori
636 power analysis)?
637
638 11. Randomization: Do the authors state if randomization occurred and justify why or why
639 not? (1 point each question)
640
641 12. Study_Authorization: If the study involves field sampling, do the authors state if relevant
642 permits were obtained and provide details of authority approving study? If none were
643 required, is there an explanation?

644 Data Analysis

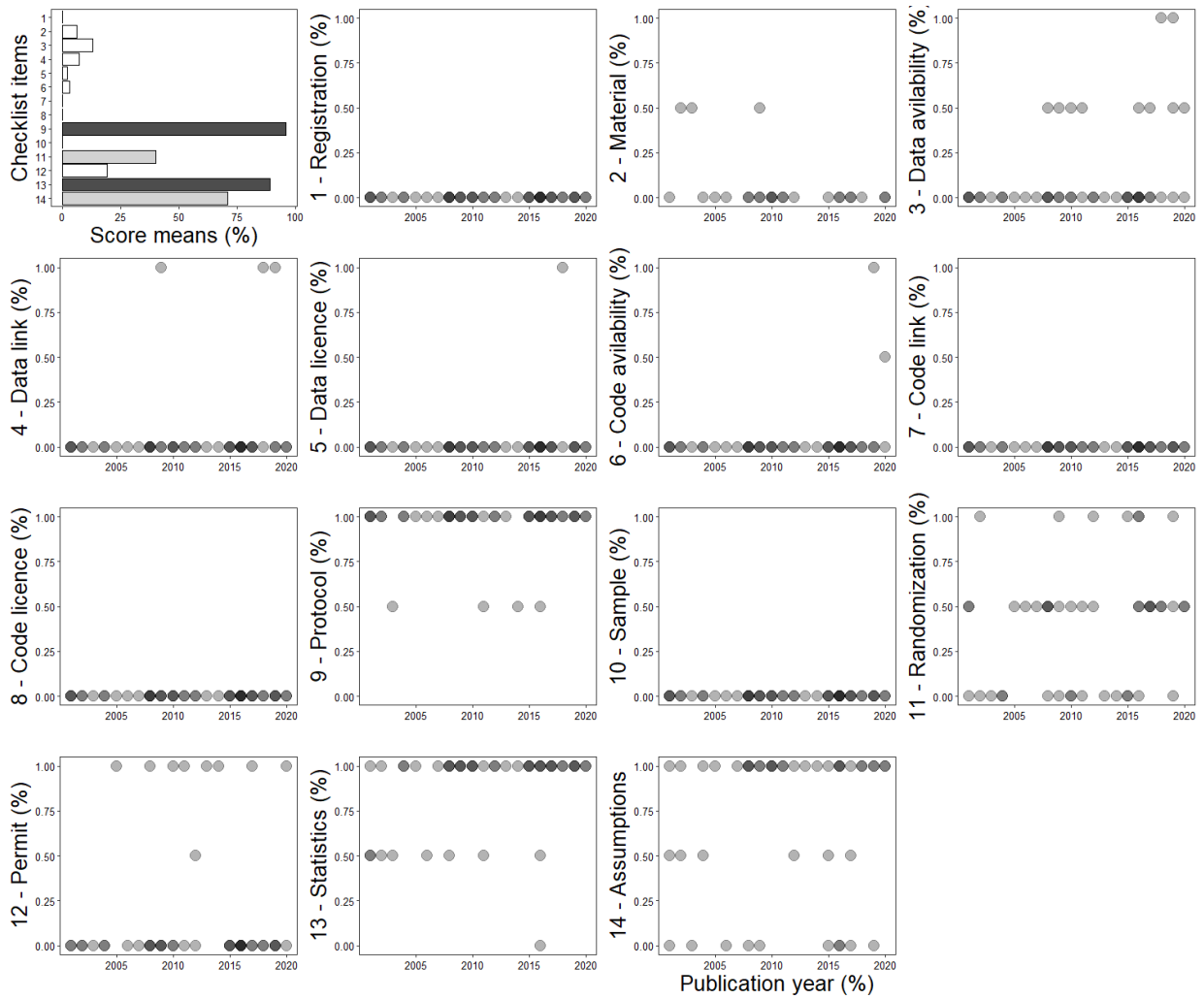
- 645 13. Statistics: Do the authors describe their use of statistical tests and justify their choice of
646 tests?
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648 14. Assumptions: Are appropriate assumptions of the applied statistical tests discussed
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658 **Appendix 3**

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660 **Figure S1** Plots displaying scores by item category. First is the average of each score.
661 Colours indicate *overlooked* (white, < 20%), *in progress* (grey, 20-80%), and *addressed* (black, >80%) items.
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