Evidence of misuse of nonparametric tests in the presence of heteroscedasticity within obesity research [version 1; peer review: 2 approved]

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Abstract

Background:
Classic nonparametric tests (cNPTs), like Kruskal–Wallis or Mann–Whitney U, are sometimes used to detect differences in central tendency (i.e., means or medians). However, when the tests’ assumptions are violated, such as in the presence of unequal variance and other forms of heteroscedasticity, they are no longer valid for testing differences in central tendency. Yet, sometimes researchers erroneously use cNPTs to account for heteroscedasticity.

Objective:
To document the appropriateness of cNPT use in obesity literature, characterize studies that use cNPTs, and evaluate the citation and public sharing patterns of these articles.

Methods:
We reviewed obesity studies published in 2017 to determine whether the authors used cNPTs: (1) to correct for heteroscedasticity (invalid); (2) when heteroscedasticity was clearly not present (correct); or (3) when it was unclear whether heteroscedasticity was present (unclear).

Open science R packages were used to transparently search literature and extract data on how often papers with errors have been cited in academic literature, read in Mendeley, and disseminated in the media.

Results:
We identified nine studies that used a cNPT in the presence of heteroscedasticity (some because of the mistaken rationale that the test corrected for heteroscedasticity), 25 articles that did not explicitly state whether heteroscedasticity was present when a cNPT was used,
and only four articles that appropriately reported that heteroscedasticity was not present when a cNPT was used. Errors were found in observational and interventional studies, in human and rodent studies, and only when studies were unregistered. Studies with errors have been cited 113 times, read in Mendeley 123 times, and disseminated in the media 41 times, by the public, scientists, science communicators, and doctors.

Conclusions:
Examples of inappropriate use of cNPTs exist in the obesity literature, and those articles perpetuate the errors via various audiences and dissemination platforms.

Keywords
Nonparametric tests, heteroscedasticity, research rigor, statistical methods, open science, nutrition, obesity
**Introduction**

Concerns have been raised as to whether scientific research generally\(^1\)–\(^3\) and nutrition and obesity research specifically\(^4\)–\(^6\) meet modern standards of research rigor and quality. Various suggestions for improvement in conceptualization, design, or analysis, have been published by the National Institutes of Health, National Academies of Sciences, Engineering, and Medicine, and within nutrition and obesity disciplines.\(^7\)–\(^10\) One domain that requires attention is using statistical procedures in ways that foster valid inferences. Indeed, it is not uncommon in nutrition and obesity research for inappropriate statistical models and procedures to be chosen and implemented,\(^11\)–\(^13\) which can erode the quality of the science and trust in the field.\(^13\)

We previously reported on the consequences of one area of statistical error: erroneously using classic nonparametric tests (cNPTs) to test for differences in central tendency (i.e., means or medians) when heteroscedasticity is present.\(^14\) Heteroscedasticity is a difference in statistical dispersion (e.g., variance, as used here) among groups. Therein, we defined cNPTs as “tests that do not rely on any assumptions about the data having a particular distribution, other than having a finite mean and a finite variance.” Such methods include Kruskal–Wallis, Wilcoxon signed-rank, and Mann–Whitney U-tests, and exclude methods that do not involve explicit resampling (e.g., bootstrap). We refer readers to our previous paper to learn more about further intricacies on cNPTs, the misconceptions of using nonparametric tests in the presence of heteroscedasticity, why this error can be of particular concern to nutrition and obesity research, and some guidance on how to avoid some errors.\(^14\)

The objective of the present study was to determine whether misuse of cNPTs in the presence of heteroscedasticity exists in the obesity literature. Specifically, we sought to identify publications where heteroscedasticity was used as a justification for choosing a cNPT. We also aimed to describe the characteristics of included studies, such as design and model type, open science practices adopted by included studies, and the extent to which the studies have been disseminated since publication, in order to understand the context in which this error occurred and its dissemination impact.

**Methods**

**Search methods**

A pilot search was conducted in Google Scholar (Google LLC) using combinations and permutations of the terms “nonparametric,” “bootstrapping,” “obesity,” “overweight,” and “adiposity”. Herein, we use the word ‘term’ to mean words and phrases (that is, in contrast to a statistical ‘term’ in a model). This search provided evidence of the use of cNPTs in the presence of heteroscedasticity and helped to inform our search terms for this study. We searched the literature for studies 1) reporting obesity outcomes, 2) mentioning use of nonparametric tests, 3) acknowledging the presence or potential for unequal variances, and 4) published in 2017. Our use of the year 2017 coincided with preparing these materials for our tutorial article\(^14\) and allows for the follow up period of three years for the assessment of dissemination impact. The citation and dissemination impacts of those articles are current as of December 9, 2020 (see Screening and Data Extraction). Search terms were expanded based on input from J.J. Pionke, a librarian at the University of Illinois at Urbana-Champaign. The PubMed Central platform was used because it allows for full-text searches. Nutrition and obesity outcome search terms were limited to abstracts, whereas the article body was also searched for heteroscedasticity and nonparametric terms. Because heteroscedasticity and nonparametric terms are key inclusion criteria and are often not included in abstracts, this methodological choice was key to improve the feasibility of this study, considering the large number of nutrition and obesity papers published each year. To make our search reproducible, we used the Open Science R package rEntrez (SCR_021062)\(^15\) to conduct our search within PubMed Central. The full search strategy can be found in the study repository.\(^16\)

**Screening and data extraction**

Full texts were obtained and assessed for eligibility and data extraction. Articles were considered eligible if they contained an obesity outcome, a heteroscedasticity term (e.g., heterogeneous variances, unequal variances, or homogeneity of variance), and a nonparametric term (e.g., Kruskal–Wallis, rank-sum test, or Wilcoxon test), and were published in 2017. An obesity outcome was defined as a body composition outcome within a study that aimed to address the problem of overweight or obesity. Heteroscedasticity and nonparametric terms were defined by those used for the literature search, which can be found in the study repository.\(^16\) Articles were independently screened, and the data independently extracted by two investigators (CMK and either BAH or TMH). Discrepancies were resolved by consensus among CMK, BAH, and TMH.

The following data were manually extracted from each eligible article: study design (intervention or observational), subject type (human, rodent, or other), type of cNPT used, reason for use of heteroscedasticity terms, and whether findings obtained from the cNPT were statistically significant. The reason the authors referenced a heteroscedasticity term was categorized according to Table 1.
To determine whether improper use of a cNPT was potentially avoided and to gain insight on the prevalence of responsible research practices that help to improve the transparency of data analysis, we also extracted the following information from each included article: whether a study was registered in a study registry (e.g., clinicaltrials.gov), whether a study was preregistered (i.e., before the start of data collection), whether statistical analysis plans and outcomes were prespecified, and whether the study data or analysis code were described as “publicly available”, “available upon request,” or not.

Open Science R packages were used to automatically extract data on the relative impact of articles containing errors and ambiguity since their publication in 2017 to December 9, 2020. For instance, rAltmetric17 was used to extract the articles’ overall Altmetric Attention Score (Altmetric LLP), their total number of mentions (e.g., in news outlets, social media, blogs), and the cohorts giving the articles attention (e.g., members of the public, doctors, scientists, and science communicators). The Altmetric Attention Score is made up of weighted approximations for volume (total number of original mentions), sources (e.g., mentions in newspaper articles weigh more than blog posts, which weigh more than tweets), and authors (e.g., articles shared by doctors to other doctors weigh more than automatic posts from an academic journal account). The package rCrossref17 was used to extract the number of times articles were cited in academic literature. The package rEntrez was used to automatically extract publication date to determine the number of days since publication. The full code for the implemented search strategy, automatic data extraction, included studies, manually obtained data, and associated figure generation can be found at our study repository.16

Given the descriptive nature of this study and having not conducted any formal a priori power calculations nor establishing any a priori hypotheses, results are presented as counts, frequencies, and other summaries, sometimes stratified by cNPT appropriateness or study characteristics. No formal between-group comparisons or statistical inferences are calculated herein.

Results

Figure 1 depicts the flow diagram of included studies, with 38 ultimately being included. Inclusion of heteroscedasticity and nonparametric terms were not considered if the study did not contain a nutrition obesity outcome or heteroscedasticity term, respectively. Figure 2 shows the proportion of included studies that used cNPTs in clear error,9 ambiguously (25), or correctly.4

A breakdown of study characteristics by appropriateness of cNPT use is depicted in Figure 3. Study designs consisted of interventional (n = 14) or observational (n = 24), and model types of rodents (n = 10) and human participants (n = 28). P-values obtained from the cNPT were statistically significant in n=18 studies, statistically insignificant in n = 2 studies, or a mix of statistical significance depending on the outcome (n = 18). The prevalence of responsible research practices among studies by appropriateness of cNPT use can be viewed in Figure 4. Clear errors were only found in studies that were not registered. Only two studies were preregistered, and no studies preregistered a statistical analysis plan. Raw data were indicated as available for seven studies, including one study with a clear error. Analysis code was only listed as available for two studies but for none of the studies with clear errors.

As of December 9, 2020, the included studies had been cited a total of 531 times across all studies after an average of 1268 days since publication. A total of 323 of these citations were of articles containing ambiguities as to whether heteroscedasticity was present when a cNPT was used, 113 citations were of articles that erroneously used a cNPT to correct for heteroscedasticity or in the known presence of heteroscedasticity, and 95 citations were from articles that transparently reported correct use of nonparametric tests. Altmetrics were available for 28 studies, which had been read and mentioned 1578 and 564 times, respectively, and have received a net Altmetric Attention Score of 634. The cohorts that have been documented engaged in dissemination are the public (n = 225), scientists (n = 62), science

Table 1. Categories of use of nonparametric tests in the included studies.

| Description of Practice | Abbreviation | Validity |
|-------------------------|--------------|----------|
| Used a nonparametric test; explicitly stated that heteroscedasticity was not present | Correct | Valid |
| Used a nonparametric test; explicitly stated that the test was used in the presence of heteroscedasticity or to correct for heteroscedasticity | Error | Invalid |
| Used at least one nonparametric test; no explicit statement of whether heteroscedasticity was tested for or present | No Link | Unclear |
| Used a nonparametric test; stated that variance assumptions were tested for, but did not explicitly state whether heteroscedasticity was present | Ambiguous | Unclear |
communicators (n = 16), and medical doctors (n = 79). Dissemination data are depicted in Figure 5, where they are grouped by appropriateness of cNPT use. Papers with clear errors have been cited in peer-reviewed journals 113 times, read in Mendeley 123 times, obtained an Altmetric Attention Score of 131.8, and shared on the internet by the public (n = 11), science communicators (n = 4), doctors (n = 2), and research scientists (n = 6).

Discussion

Examples of improper use of cNPTs in nutrition and obesity research were easy to find among articles indexed in PubMed Central in 2017. A common reason for use of cNPTs in statistical analysis was the presence of heteroscedasticity, because assumptions of equal variances were not met. Over half of the included studies also did not clearly state the rationale for use of cNPTs with respect to heteroscedasticity or that heteroscedasticity was not present when a cNPT was used. These studies 1) did not link the rationale for using a cNPT with the use of the heteroscedasticity term, 2) reported that variance assumptions were tested for or met but did not specify for which outcomes or for all cases a cNPT was used, or 3) reported that variance assumptions were tested for or met but did not explicitly state that heteroscedasticity was not present. This lack of transparency in reporting creates ambiguity when interpreting such results, as the reader cannot determine whether a cNPT was used appropriately. The use of inappropriate tests may also lead researchers to draw the wrong conclusion, further muddying the scientific record and increasing mistrust. Without clear explanation of statistical methodology and results, these findings cannot be readily reproduced. Further, because some authors clearly and erroneously used a cNPT with heteroscedasticity, concern is raised as to whether similar misunderstandings are present in cases where reporting is vague, which in our sample occurred in two-thirds of cases. Responsible research practices, such as preregistering statistical analysis plans or clarifying whether data and analysis code are publicly available or available upon request, were not prevalent. Such practices are tools that can help improve replication of statistical findings and make details of the statistical methods more transparent.

Despite containing misused and unclearly used statistical methods, many of the included articles are being cited within both scientific and lay communities. Dissemination of research with improper statistical methods may reinforce the misuse of cNPTs in the presence of heteroscedasticity within the scientific community. If misused or unclearly used cNPTs led to invalid conclusions, both the lay and scientific communities and those they share with can become misinformed of the magnitude or statistical significance of such results.

In some cases, authors explicitly used cNPTs to erroneously attempt to address deviations from homoscedasticity. The use of cNPTs in the presence of unequal variances among experimental groups, especially if sample sizes are imbalanced, will result in an increased type I (false-positive) error rate, as shown using simulations in our previous work. These improper statistical methods can lead to incorrect conclusions and dissemination of misleading findings.
Figure 2. Appropriateness of nonparametric test use in the included studies. Error, used a nonparametric test and explicitly stated that the test was used in the presence of heteroscedasticity or to correct for heteroscedasticity; No Link, used at least one nonparametric test but made no explicit statement of whether heteroscedasticity was tested for or present; Ambiguous, used a nonparametric test and stated that variance assumptions were tested for, but did not explicitly state whether heteroscedasticity was present; Correct, used a nonparametric test and explicitly stated that heteroscedasticity was not present.

Strengths and limitations

Strengths of our approach include the transparent and reproducible research process. Conducting the literature search programmatically and sharing the code in a study repository allows the possibility of exact replication, as well as clarity around how the search was done. Further, extracting citation counts, date published, and Altmetric data programmatically may help improve the speed of data acquisition while minimizing extraction error that comes with independently manually extracting data. Our study repository transparently describes the research process for this project. Dissemination data were extracted for a conference presentation in 2018, simulations on why this error is a problem statistically were published in 2020, and dissemination data were updated for submission of this manuscript and data for peer review publication. The repository explains the differences and reasons for each publication, shows how the dissemination data have changed over time, and foster transparency and reproducibility of each step of the project process. In addition, the use of papers in 2017 is relatively recent to investigate the field’s use of such tests, while also allowing enough lag-time to observe accumulation of dissemination trends.

Limitations of our approach include that we do not know whether our findings are representative of all nutrition and obesity papers, even those published in 2017, but only to those that met our inclusion criteria in PubMed Central. Limiting our review to PubMed Central allowed us to use the R packages “rentrez” and “fulltext” to programmatically search full texts of articles for key inclusion criteria, but excludes studies not indexed in PubMed Central and thus may not be representative. Future work might expand these methods to other databases by using packages to download full texts and text mine them. Furthermore, our approach used terms for heteroscedasticity, nonparametric tests, nutrition, and obesity from a preliminary search of the literature, but may not capture all search terms, and thus may have resulted in some relevant literature being excluded from our review. Consistent with these limitations, we do not present our numbers as evidence of the prevalence of the problem within the literature, but rather as a systematically obtained case series to illustrate issues with cNPT reporting and use within the literature.
Figure 3. Appropriateness of nonparametric test use in nutrition and obesity studies published in 2017 according to study characteristics. *Nonparametric Tests: Description of the test used was only “non-parametric tests.” Observed statistical significance: Number of articles that report results obtained from a nonparametric test as statistically significant (significant), mixed (mixed significance), or not significant. Correct, used a nonparametric test and explicitly stated that heteroscedasticity was not present; Ambiguous, used a nonparametric test and stated that variance assumptions were tested for but did not explicitly state whether heteroscedasticity was present; No Link, used at least one nonparametric test, but made no explicit statement of whether heteroscedasticity was tested for or present; Error, used a nonparametric test and explicitly stated that the test was used in the presence of heteroscedasticity or to correct for heteroscedasticity.

Figure 4. Appropriateness of nonparametric test use in nutrition and obesity studies published in 2017 according to reported responsible research practices. Correct, used a nonparametric test and explicitly stated that heteroscedasticity was not present; Ambiguous, used a nonparametric test and stated that variance assumptions were tested for but did not explicitly state whether heteroscedasticity was present; No Link, used at least one nonparametric test, but made no explicit statement of whether heteroscedasticity was tested for or present; Error, used a nonparametric test and explicitly stated that the test was used in the presence of heteroscedasticity or to correct for heteroscedasticity. For Raw Data Available, “No” means the authors stated the data are not available, the paper had no data availability statement, or the data availability statement was left blank.
Conclusion
The use of cNPTs in the presence of heteroscedasticity was present among nutrition and obesity articles indexed in PubMed Central and published in 2017. In some cases, the cNPT was erroneously used to correct for heteroscedasticity, while many of the statistical methods and results sections of the included articles were ambiguous. Better reporting and appropriate use of cNPTs is needed in nutrition and obesity literature.

Acknowledgments
We would like to thank J.J. Pionke (University of Illinois) for his insight on development of the search strategy. We also would like to thank David B. Allison, John A. Dawson, David B. King, and Bryan McComb for their helpful feedback on earlier drafts of materials presented herein as part of our tutorial paper on cNPT.14

Data availability
Zenodo: Underlying data for ‘Evidence of misuse of nonparametric tests in the presence of heteroscedasticity within obesity research’. https://doi.org/10.5281/zenodo.4733330.16

This project contains the following underlying data:

- Raw data
- Code book
- Analytical code

Data are available under the terms of the Creative Commons Attribution-ShareAlike 4.0 International license (CC BY-SA 4.0).
References

1. Munafò MR, Nosek BA, Bishop DVM, et al.: A manifesto for reproducible science. Nat Hum Behav. 2017; 1: 0021. PubMed Abstract | Publisher Full Text | Free Full Text

2. Hsieh T, Vaickus MH, Remick DG: Enhancing Scientific Foundations to Ensure Reproducibility. A New Paradigm. Am J Pathol. 2018; 188(1): 6-10. PubMed Abstract | Publisher Full Text | Free Full Text

3. Allison DB, Brown AW, George BJ, et al.: Reproducibility: A tragedy of errors. Nature. 2016; 530(7588): 27. PubMed Abstract | Publisher Full Text | Free Full Text

4. Allison DB, Brown AW, George BJ, et al.: Common scientific and statistical errors in obesity research. Obesity (Silver Spring). 2016; 24(4): 781–90. PubMed Abstract | Publisher Full Text | Free Full Text

5. Byrne JL, Yee T, O'Connor K, et al.: Registration status and methodological reporting of randomized controlled trials in obesity research: A review. Obesity (Silver Spring). 2017; 25(4): 665–70. PubMed Abstract | Publisher Full Text | Free Full Text

6. Brown AW, Ioannidis JP, Cope MB, et al.: Unscientific Beliefs about Scientific Topics in Nutrition. Oxford University Press; 2014.

7. Brown AW, Kaiser KA, Allison DB: Issues with data and analyses: Errors, underlying themes, and potential solutions. Proc Natl Acad Sci U S A. 2018; 115(11): 2563–70. PubMed Abstract | Publisher Full Text | Free Full Text

8. Wood AC, Wren JD, Allison DB: The Need for Greater Rigor in Childhood Nutrition and Obesity Research. JAMA Pediatr. 2019; PubMed Abstract | Publisher Full Text | Free Full Text

9. National Institutes of Health: Implementing Rigor and Transparency in NIH & AHRQ Research Grant Applications. 2015.

10. National Academies of Sciences, Engineering, and Medicine: Reproducibility and Replicability in Science. Washington, DC: The National Academies Press; 2019.

11. George BJ, Brown AW, Allison DB: Errors in statistical analysis and questionable randomization lead to unreliable conclusions. J Paramed Sci. 2015; 6(3): 153–4. PubMed Abstract | Free Full Text

12. Mehta T, Allison DB: From Measurement to Analysis Reporting: Grand Challenges in Nutritional Methodology. Front Nutr. 2014; 1(6). PubMed Abstract | Publisher Full Text | Free Full Text

13. Kroeger CM, Garza C, Lynch C, et al.: Scientific rigor and credibility in the nutrition research landscape. Am J Clin Nutr. 2018; 107(3): 484–94. PubMed Abstract | Publisher Full Text | Free Full Text

14. Kroeger CM, Ejima K, Hannon BA, et al.: Persistent confusion in nutrition and obesity research about the validity of classic nonparametric tests in the presence of heteroscedasticity: evidence of the problem and valid alternatives. Am J Clin Nutr. 2021. PubMed Abstract | Publisher Full Text | Free Full Text

15. Winter D: rentrez: Entrez in R. R package version 1.1.0 ed, 2017: RRID: SCR_021062.

16. Kroeger CM: Data and analysis code for manuscript: Evidence of misuse of nonparametric tests in the presence of heteroscedasticity within obesity research. Zenodo. 2021. Publisher Full Text

17. Chamberlain S, Boettiger C, Hart T, et al.: Client for various ‘CrossRef’ APIs: A package version 0.7.0 ed. 2017.

18. Buscemi N, Hartling L, Vandermeer B, et al.: Single data extraction generated more errors than double data extraction in systematic reviews. J Clin Epidemiol. 2006; 59(7): 697–703. PubMed Abstract | Publisher Full Text

19. Kroeger CM: Data and analysis code for published abstract: Misuse of nonparametric tests with heteroscedasticity: A semi-automated review of obesity research. Zenodo. 2018. Publisher Full Text

20. Kroeger CM, Ejima K: Data and analysis code for manuscript: Persistent confusion in nutrition and obesity research about the validity of classic nonparametric tests in the presence of heteroscedasticity: Evidence of the problem and valid alternatives. Zenodo. 2019. Publisher Full Text
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Yes, these classic methods are indeed inappropriate for dealing with measures of central tendency under general conditions. This has been known in the statistics literature for many years. This paper underscores the fact that these insights remain unknown by many researchers. Generally, there is a serious gap between advances relevant to basic techniques and how the typical researcher is trained. Hopefully this paper will help motivate others to deal with this problem.

There is a vast literature dealing with robust methods for making inferences about measures of central tendency (Wilcox (2022¹)). They deal with heteroscedasticity as well as concerns about non-normality. It might help readers if this is pointed out.

References
1. RR, Wilcox: Introduction to Robust Estimation and Hypothesis Testing. Academic Press. 2022.

Is the work clearly and accurately presented and does it cite the current literature?
Yes

Is the study design appropriate and is the work technically sound?
Yes

Are sufficient details of methods and analysis provided to allow replication by others?
Yes

If applicable, is the statistical analysis and its interpretation appropriate?
Yes

Are all the source data underlying the results available to ensure full reproducibility?
No source data required

Are the conclusions drawn adequately supported by the results?
Yes

**Competing Interests:** No competing interests were disclosed.

**Reviewer Expertise:** Robust statistical methods

I confirm that I have read this submission and believe that I have an appropriate level of expertise to confirm that it is of an acceptable scientific standard.

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**Joscha Krause**

Department of Economic and Social Statistics, Trier University, Trier, Germany

**Short summary:**
The presented paper provides a literature review on the use of nonparametric tests for the detection of central tendency differences in grouped obesity research data. In particular, the authors investigate how often corresponding tests have been applied in the presence of heteroscedasticity. Since many common tests (e.g., Kruskal-Wallis, Mann-Whitney-U) assume variance homogeneity between groups, their application in such circumstances can lead to false conclusions. To evaluate how common such misuse is in the literature, the authors review articles published in 2017. They evaluate whether the nonparametric tests were applied correctly, incorrectly, or whether it cannot be determined.

**Comments:**
- Being a statistician, I can only comment on the statistical aspects of the presented paper.
- The paper is compact and well-written. It is well-known that a significant number of empirical studies in this field suffer from inaccuracies due to a lack in statistical rigor. Accordingly, a quantitative evaluation providing evidence on this problem is a welcome contribution.
- The authors give a transparent description on how they searched for relevant papers, and how they categorized the studies. They further provided the necessary programming code for reproduction. The presentation of the results is clear and expedient. The authors point out that their study is purely descriptive and does not provide any statistical inference. Against this background, there is not much room for methodological points of critique. Within this scope, the study is well done.
- Maybe the authors would like to consider the subsequent minor points as a proposal for some additions to the manuscript.
o For readers from statistics, it would have been interesting to see how exactly a violation of the homogeneity assumption affects the test outcomes. An increase in type-1-errors is mentioned. However, is this due to the asymptotic distributions of the test statistics being invalid, are the tests biased, do they lose testing power? I understand that this paper is a ‘follow-up’ to a previous technical study on this matter by the authors. Nevertheless, maybe they can add a few more words on that aspect.

O Another interesting point would have been to see how often the false application of the tests in the literature has actually led to seriously wrong conclusions. However, even without that aspect, the presented paper makes a convincing case for addressing this important issue.

Recommendation:
 O I have no serious concerns from a statistical perspective and, thus, approve the indexing of the paper.

Is the work clearly and accurately presented and does it cite the current literature?
Yes

Is the study design appropriate and is the work technically sound?
Yes

Are sufficient details of methods and analysis provided to allow replication by others?
Yes

If applicable, is the statistical analysis and its interpretation appropriate?
Yes

Are all the source data underlying the results available to ensure full reproducibility?
Yes

Are the conclusions drawn adequately supported by the results?
Yes

Competing Interests: No competing interests were disclosed.

Reviewer Expertise: Statistics, data science, Monte Carlo simulations

I confirm that I have read this submission and believe that I have an appropriate level of expertise to confirm that it is of an acceptable scientific standard.
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