- 1 Covered with P: The consequences of null hypothesis significance testing on point and
- interval estimates
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10	Abstract

11 Confidence intervals (CI) do not allow post-data inference on parameter values.

12 Keywords: confidence interval, NHST

Word count: Short and sweet.

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Covered with P: The consequences of null hypothesis significance testing on point and interval estimates

17 Introduction

Scientists, psychological or otherwise, routinely use null hypothesis significance testing procedures (NHSTP) to move from data to conclusions—a practice whose applicability has been debated since its inception. Recently, concerns about the replicability and reliability of empirical findings (Collaboration, 2015) have reignited questions about the validity of NHSTP as they are implemented and interpreted in practice (Cumming, 2014; Gelman & Loken, 2014).

- One response to the growing concerns regarding the reliability of NHSTP has been an appeal to effect size and interval estimation in addition—or as replacement—to NHSTP test statistics (Cumming, 2014). For example, many journals in psychology and neuroscience now ask authors to include *confidence intervals* (CI) with their test statistics. These confidence intervals are intended, by practitioners, to communicate estimates of likely ranges of parameter values, although the common computations simply result in ranges of parameter values that would not be rejected by the very statistical test the CI is supposed to supplement. This approach is problematic from at least two perspectives:
 - CIs do not support probability statements about parameters

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- CIs are designed (Neyman, 1957) to be a procedure of generating a set of intervals
 that, as a whole set, contain the true parameter value X% of the time
 - This property is severely compromised by the current practice of using CIs as a post-data inferential tool, as we show in this paper

While the first of these perspectives is misunderstood by many researchers in psychology and other fields (citing stuff), we focus here on the second, which is more subtle and even less appreciated among practitioners. A search of [searching for textbooks in a

principled way shows that ~X% [surely some depressing number] endorse the view that the frequency coverage or "confidence" of a CI refers to the specific values obtained for a particular interval. This is an incorrect interpretation to begin with, but given its 42 widespread acceptance it is important to describe specific examples in which such an interpretation produces misleading results. In this paper, we report an unappealing property of obtained confidence intervals for 45 results that pass a significance test. Because the claim of confidence intervals is to have a coverage proportion of the true parameter value equal to the nominal value (usually 95%), it is crucial that this claim is substantiated in its long-run property for a CI to be what it claims to be (not a confidence interval.) We show that using confidence intervals in addition to P values leads to an undesirable distortion of the coverage proportion. This is a direct result of the more general problem with interpreting any particular obtained CI in terms of frequency coverage, but we feel the specific case of significance thresholding on this interpretation is worth describing, given the pervasive reporting and interpretation of such intervals following NHSTP in a wide range of research areas. We demonstrate analytically and with simulations that the proportion of confidence intervals that are significant and 55 contain the true parameter in question is a function of the power of a statistical procedure. Given the low power of many psychological studies (cite), and the widespread belief that 57 obtained confidence intervals' nominal coverage provides valid information about parameters 58 of interest (see blank for recommendations along this line), this demonstration may be of use 59 to practitioners who feel confused or even strongly about the benefits of confidence interval estimation in NHSTP. 61

62 P stains the nominal coverage proportion

63 <!— Here we describe the gist of the paper. —>

64 Methods

We performed a simulation study...

66 Results

[1] 0.6457245

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Discussion

hypothesis significance threshold. We don't know if this result is well known to statisticians, 70 but from the perspective of practitioners, we found it suprising. This paper was motivated in 71 part by the discussions with colleagues who were equally suprised by the biases induced by 72 hypothesis testing, especially on interval estimation. The "significance filter" has been 73 discussed previously (Gelman, 2011), but to our knowledge there have been no discussions of the effect of this filter on the frequency properties of confidence intervals. 75 We note that, while the issues discussed in this paper are related to questionable 76 research practices as well as known issues in null hypothesis testing such as alpha inflation due to multiple comparisons, the biased point estimates and interval coverage for significant results we discuss here are present in expectation even for single tests. Thus this bias will be more severe for significant results which have been filtered through such processes, but...

Here we show a pervasive bias in the paramaters and intervals passing a null

References

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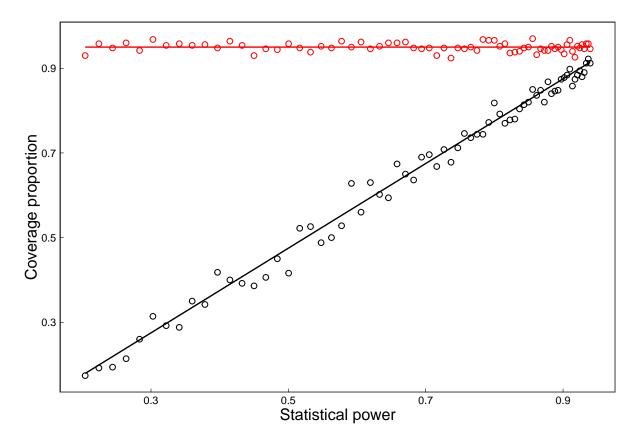


Figure 1. Coverage proportion of confidence intervals versus statistical power. The red line represents the nominal coverage proportion (95%), black line is the actual coverage proportion when CIs are conditioned on the result being significant.