

The airport fallacy

Justin Smith's response ("An embarrassment of riches", page 46, October 2017) to our article "What are the odds!? The 'airport fallacy' and statistical inference" (August 2017) defends frequentist inference, stating that: "I simply fail to see how frequentism is a fallacy." Strangely, Smith did not actually address any of the arguments we made.

He first describes a simplistic coin-flipping experiment where there are little or no model selection issues that are typical of real data. But our central point was that data-dependent choices made during the model-building process for real, complex data afford a legion of opportunities for overfitting and, consequently, flawed inference. Moreover, without strict adherence to a pre-specified experimental protocol and statistical analysis plan – which would suffocate the exploratory nature of most scientific research – there isn't even a well-defined probability space on which to make such inferences. This is the airport fallacy. Coin flipping does not usefully represent this situation.

Smith next cites an example of 20 historical estimates of the solar deflection of light. He states that while individual experiments may have flaws, the estimates are consistent with general relativity, and confidence intervals are shrinking. Similar examples abound in the physics literature (e.g. the "Review of Particle Physics").¹ In his classic 1972 paper, "Enduring values", W. J. Youden discussed laboratory measurement examples in which different laboratories' estimates often disagreed by more than their reported error bars.² He interpreted this as evidence of unmeasured systematic error. This reinforces our point – statistical inference characterises uncertainty based on models of putative error sources, but such models are inherently oblivious to extra experimental systematic errors. This again implies that the greatest source of uncertainty stems from model choice: error bars obtained conditional on a particular chosen model understate this true uncertainty.

Smith next cites Deborah Mayo's work on severe testing. But Mayo defines severity using probability, thus requiring a model, so model uncertainty and overfitting issues still apply. Finally, Smith cites the argument that "likelihood swamps the prior" so that frequentist and Bayesian approaches usually tend to give similar results. Yet again this assumes reliable probability models and likelihood functions, ignoring the model choice and uncertainty issues that we argued are the heart of the problem.

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References

1. C. Patrignani *et al.* (Particle Data Group) (2016) Review of Particle Physics. *Chinese Physics C*, **40**, 100001. See 2017 update at pdg.lbl.gov.
2. Youden, W. J. (1972) Enduring values. *Technometrics*, **14**, 1–11.

In their August 2017 article, Gunter and Tong question the widespread use of frequentist statistical inference. They identify

just one place where the methodological problems that concern them may be avoided by rigorous design and pre-specification, namely in the conduct of pivotal clinical trials in medicine. They believe that the difficulties and resources required to protect reliable inferences are far too expensive and time-consuming for other areas of investigation. I write to contest this point.

I was heavily involved in the development of international guidelines for the design, conduct and analysis of clinical trials carried out for the purpose of licensing new medicines. These were the global ICH E9 guideline¹ and its European predecessor.² The inference problems that Gunter and Tong highlight were at the forefront of our minds in drafting these guidelines so that clinical trials worldwide would lead to reliable regulatory decisions.

In ICH E9 an important distinction is made between confirmatory and exploratory trials. As the name implies, a confirmatory trial is carried out to test a well-developed hypothesis generated by, and relying on, earlier work. This earlier work should provide all necessary information to aid the design of the confirmatory study and to allow the pre-specification of all important aspects of the conduct of the trial and its analysis. Some of the earlier work is likely to have been carried out in earlier exploratory trials, less rigorously designed and analysed.

It is hard to see why such an approach is not possible in most areas of scientific work covered by standard texts on the design and analysis of experiments. Gunter and Tong point to the time and cost involved. However, the cost of clinical trials is heavily influenced by the need to carry them out in a large number of patients, often in an international multi-centre setting, requiring major resources to recruit and manage the patients, monitor the study and collect data uniformly and reliably. Difficulties of this nature do not arise, for example, in studies carried out in laboratories, in many agricultural experiments, or in studies of manufacturing facilities. In addition, time invested at the time of design is always amply rewarded by savings of time in gaining acceptance of results, in implementing conclusions and in avoiding repetition of experiments.

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References

1. Lewis, J. A. (1999) Statistical principles for clinical trials (ICH E9): An introductory note on an international guideline. *Statistics in Medicine*, **18**, 1903–1904.
2. Lewis, J. A., Jones, D. R. and Röhmel, J. (1995) Biostatistical methodology in clinical trials – a European guideline. *Statistics in Medicine*, **14**, 1655–1682.

On sample size

There are two points that one might add to Deirdre Toher's otherwise excellent and very thorough article ("Help! Is my sample big enough?", August 2017).

She points out that grinding up all the 40 flowers in the treatment

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