- 1. The power of simulating data: a tool to design experiments,
- ² understand data limitations and improve scientific reasoning
- 2. Between noise and patterns: the power of data simulation in science to overcome perception biases
- 3. The power of simulating data: a tool for scientists from designing experiments to drawing/reaching reasonable conclusions
- 4. Between noise and patterns: Overcoming perception biases through data simulation

Frederik Baumgarten¹, Elizabeth Wolkovich, invite Andrew Gelman May 30, 2024

2 Abstract

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- Advertise simulation (for all: frequentists, bayesian, machine learning)
- What's the Problem, what do we want to save and what's the solution
- Werwolf: p-values, 'sloppy', simplified stats, over-interpretation of patterns that emerge by chance
- Baby: scientific standards, correct conclusions, knowing the potential and limits of a dataset
- $_{\rm 17}\,$ Silverbullet: Data simulation through formulation of a mathematical model and playing with the parameters and the replication

$_{\circ}$ 1 Introduction

21 Opening example

- 22 Human desire for patterns even in pure noise
- 23 Confirmation bias
- 24 But noisy data
- 25 how can we ensure a standard?

Current solutions

- 28 scientific workflow (fig)
- include experiments, null hypothesis testing and their limits(not meaningull, not testable, not interesting), analysis -> conclusions

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Growing evidence that this is not enough

- 33 p-values/replication crisis/overconfidence/overinterpretation/mathematical artefacts
- Show famous examples. -> could expand into box

Some people hope to address this through machine learning

- machines search for patterns without or less bias (or at least in a systematic/objective way)
- machine learning is usually amechanistic
- problem with hypothesis remain. searching for a model so it includes much of the assumptions/hy-
- 40 potheses expected from a useful model

42 **Aim**

- We propose an updated approach focussed on simulation + show how to do it
- 44 Helps to address current gaps/limitations:
- build hypothesis, then formulation of mathematical model
- 46 better design experiments
- 47 avoid overconfidence/overinterpretation and mathematical artefacts

Bus example with simulation workflow

50 Situation

51 model - show poisson distr. simulate

52 no what?

- still waiting for the bus...outlayer?
- ⁵⁴ update the model assumptions
- 55 add variable traffic jam

3 Biological example - how to do it

 $_{58}$ question based + we walk through each one.

what influences y?

- 61 nitrogen
- what form? linear/nonlinear, near Gaussian, Poisson
- 63 linear

What assumptions are reasonable?

- 65 for y, alpha, beta
- 66 for effect sizes (parameters)
- 67 for x data
- 68 -> pick some for this example

70 Simulate!

71 How to use this - Play!

72 so many ways, we highlight just a few

73 Power analysis

74 to better design experiments

75

76 Avoiding overconfidence

77 play with replication while holding variance and effect size constant. p-value figure

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₇₉ Increasing importance in the future

80 Avoiding overconfidence

- 81 evergrowing Lit with AI we must learn to ask better questions
- the right questions and hypothesis that are testable with current + new methods. Build up house of
- knowledge instead of using new pattern finding algorithms
- 84 how to integrate with AI?

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