



Bayesian Statistics as an Alternative for Analyzing Data and Testing Hypotheses

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Can Social Norms Increase Towel Reuse?

Standard Environmental Message



Reuse:

35.1%

Descriptive Social Norm



44.1%

$$X^2(1, N = 433) = 3.72; p = .05$$

The Null Ritual

1. Set up a statistical null hypothesis of “no mean difference” or “zero correlation.” Do not specify any substantive alternative hypotheses.
2. Calculate a p-value.
3. If $p < .05$, refer to the result as “statistically significant” and interpret it as if you obtained evidence for the alternative hypothesis.

What is Wrong with this Approach?

- p-values do not quantify evidence for your null or alternative hypothesis
- With high enough sample size, you are guaranteed to get $p < .05$
- p-values depend on the intention of the researcher (identical data can yield different p values)
- p-values are not consistent: Little can be learned from non-significant results

When H_0 is True, p-Values are not Consistent

Alternative hypothesis is true

More data \rightarrow higher chance of identifying the true hypothesis

Null-Hypothesis is true

More data \rightarrow chance of identifying the true (null) hypothesis is constant (at $1-\alpha$)

Under H_0 , p-Values are uniformly distributed

p-Values Depend on the Intention of the Experimenter

- Default assumption: N is fixed

actually: $p = .053$

$$\chi^2(1, N = 433) = 3.72; p = .05$$

- What if the number of hotel guests is a random variable?

$$N_{\text{control}} \sim \mu = 211 \quad \sigma = 30$$

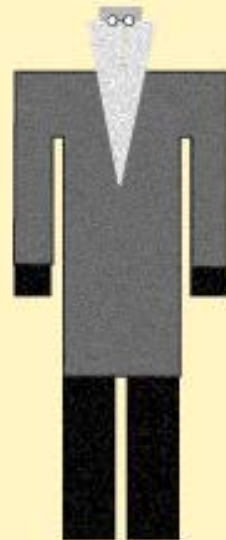
$$N_{\text{ex}} \sim \mu = 222 \quad \sigma = 30$$

$$\chi^2(1, N = 433) = 3.72; p = .06$$



19TH CENTURY SCIENTIST

I MUST FIND
THE EXPLANATION
FOR THIS
PHENOMENON IN
ORDER TO TRULY



**UNDERSTAND
NATURE...**



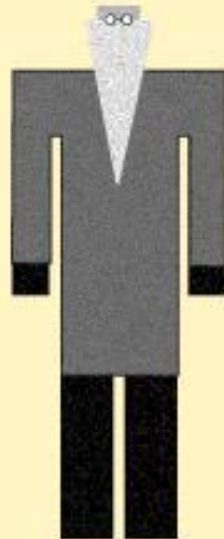
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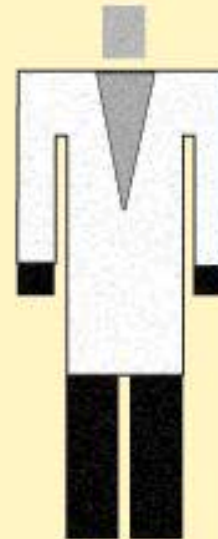
21ST CENTURY SCIENTIST

I MUST GET THE
RESULT THAT

FITS MY NARRATIVE

SO I CAN GET MY

**PAPER INTO
NATURE...**



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American Statistical Association Issues Warning

nature International weekly journal of science



Statisticians issue warning over misuse of P values

Policy statement aims to halt missteps in the quest for certainty.

“The widespread use of “statistical significance” (generally interpreted as “ $p \leq 0.05$ ”) as a license for making a claim of a scientific finding (or implied truth) leads to considerable distortion of the scientific process.”

Misuse of the P value — a common test for judging the strength of scientific evidence — is contributing to the number of research findings that **cannot be reproduced**, the American Statistical Association (ASA) warns in a **statement** released today¹. The group has taken the unusual step of issuing principles to guide use of the P value, which it says cannot determine whether a hypothesis is true or whether results are important.

This is the first time that the 177-year-old ASA has made explicit recommendations on such a foundational matter in statistics, says executive director Ron Wasserstein. The society's members had become increasingly concerned that the P value was **being misapplied** in ways that cast doubt on statistics generally, he adds.



U
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Message is not new...

“To report a significant result and reject the null in favor of an alternative hypothesis is meaningless [...].” (*Richard Feynman*)

“In the land of the blind, keep your eyes closed.”

[proverb from Kurdistan]



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Bayesian Approach Provides an Alternative



“Friends don’t let friends compute p-values”

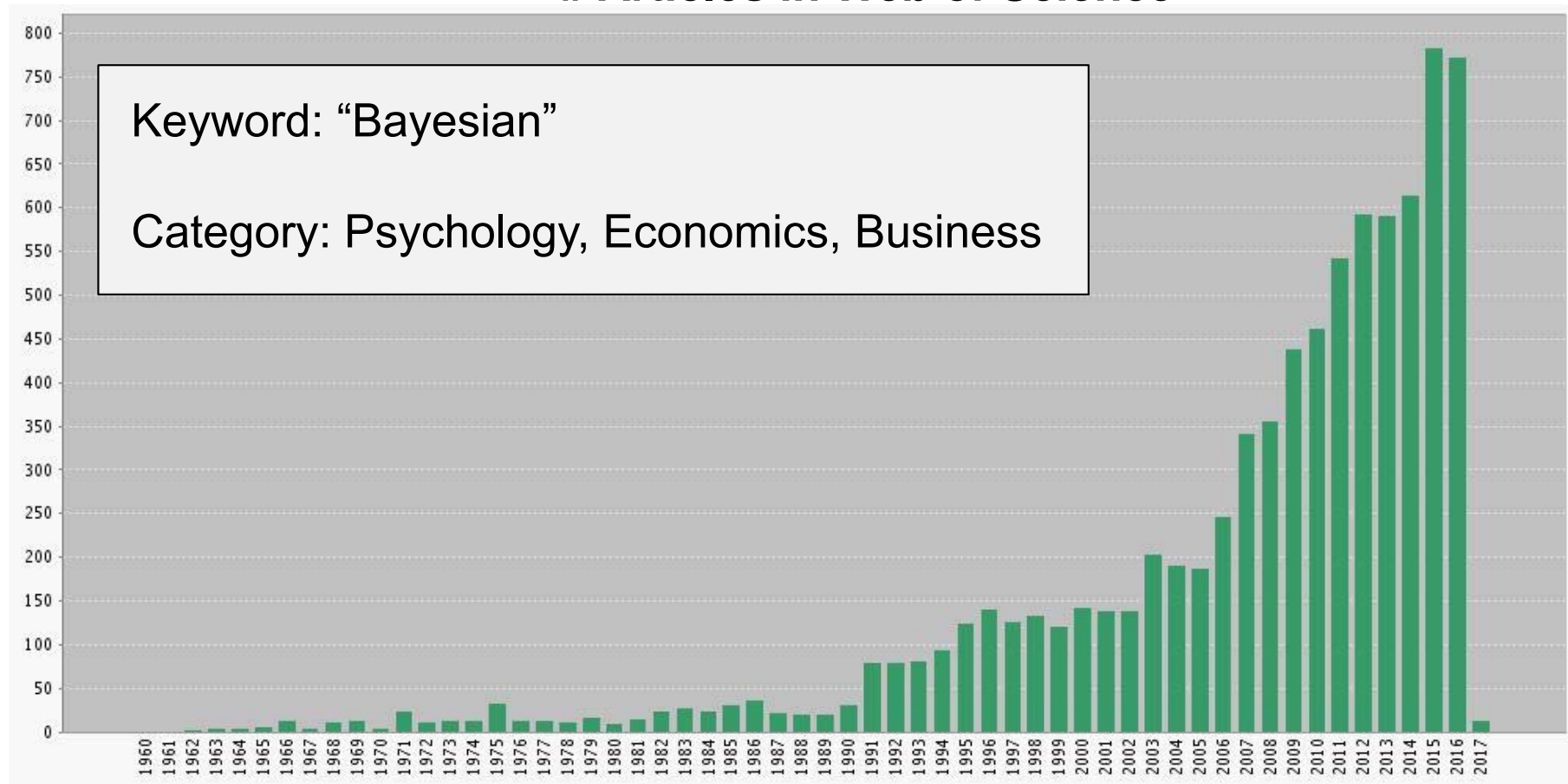


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Bayesian Approach Increasingly Attractive

Articles in Web of Science



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What Makes the Bayesian Approach Attractive?

New Results

- Evidence for Hypotheses (incl. H_0)

Better Results

- More accurate estimates and predictions
- Intuitive interpretation

Compare different models and hypotheses

- Take model complexity into account

Broad application

- “One-stop shop“



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Can Social Norms Increase Towel Reuse?

Standard Message

Descriptive Social Norm



What is the evidence?

Ex1: 35.1%

44.1% N = 433 p = .05

Ex2: 37.2%

44.5% N = 1,595 p = .03



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Bayesian Hypothesis Test

- Suppose we have two hypotheses: H_1 and H_0
- Which hypothesis is better supported by the data?
- The model that makes the best prediction for the data!
- Ratio of predictive performance = **Bayes Factor**

Bayes Factor

Prior odds

$$\frac{p(H_1)}{p(H_0)}$$

•

Bayes Factor

X

=

Posterior odds

$$\frac{p(H_1 | D)}{p(H_0 | D)}$$

Change in probability of one hypothesis/ model over another after seeing the data

posterior probability of one model over the other



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Bayes Factor

Prior odds

$$\frac{p(H_1)}{p(H_0)}$$

Bayes Factor

$$\frac{\int p(D | H_1) \cdot p(H_1)}{\int p(D | H_0) \cdot p(H_0)}$$

Posterior odds

$$= \frac{p(H_1 | D)}{p(H_0 | D)}$$

Change in probability of one hypothesis/ model over another after seeing the data

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Bayes Factor

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Change in probability of one hypothesis/ model over another after seeing the data

posterior probability of one model over the other

Simple Interpretation:

BF = 3: After seeing the data, H_1 became 3 times more probable than H_0

Bayes Factor

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Bayes Factor

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Change in probability of one hypothesis/ model over another after seeing the data

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Simple Interpretation:

BF = 3: After seeing the data, H_1 became 3 times more probable than H_0

1 to 3	Anecdotal
3 to 10	Substantial
10 to 100	Strong
>100	Decisive



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Advantages of Bayes Factors

- Quantifies evidence instead of forcing an all-or-none decision
- Easy to interpret
- All assumptions are explicit
- Able to distinguish between “data support H_0 ” and “data are not diagnostic”

Bayes Factor for Towel Data

Ex1	Reuse	Throw
Standard message	74	137
Social norm	98	124

Bayes Factor for Towel Data

Assumptions:

- Sample size in each condition (row) is fixed
- Probability of reuse = π_{standard} , $\pi_{\text{social norm}}$
- Prior: Each value for π is equally probable a-priori
- $H_0: \pi_{\text{standard}} = \pi_{\text{social norm}}$
- $H_1: \pi_{\text{standard}} < \pi_{\text{social norm}}$

Ex1	Reuse	Throw
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Independent multinomial distribution

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Ex1	Reuse	Throw
Standard message	74	137
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Independent multinomial distribution

Result:

- $BF = 1.46$
- BF for undirected hypothesis ($\pi_{\text{standard}} \neq \pi_{\text{social norm}}$) = 0.75

Can Social Norms Increase Towel Reuse?

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Descriptive Social Norm



What is the evidence?

Ex1: 35.1% 44.1% N = 433 p = .05 **BF = 1.46**

Ex2: 37.2% 44.5% N = 1,595 p = .03 **BF = 2.03**



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Can Social Norms Increase Towel Reuse?

Lots of failed replications...

Schultz et al. (2007) *Social Influence*

Mair & Bergin-Seers (2010) *Tourism and Hospitality Research*

Bohner & Schlüter (2014) *PloS ONE*

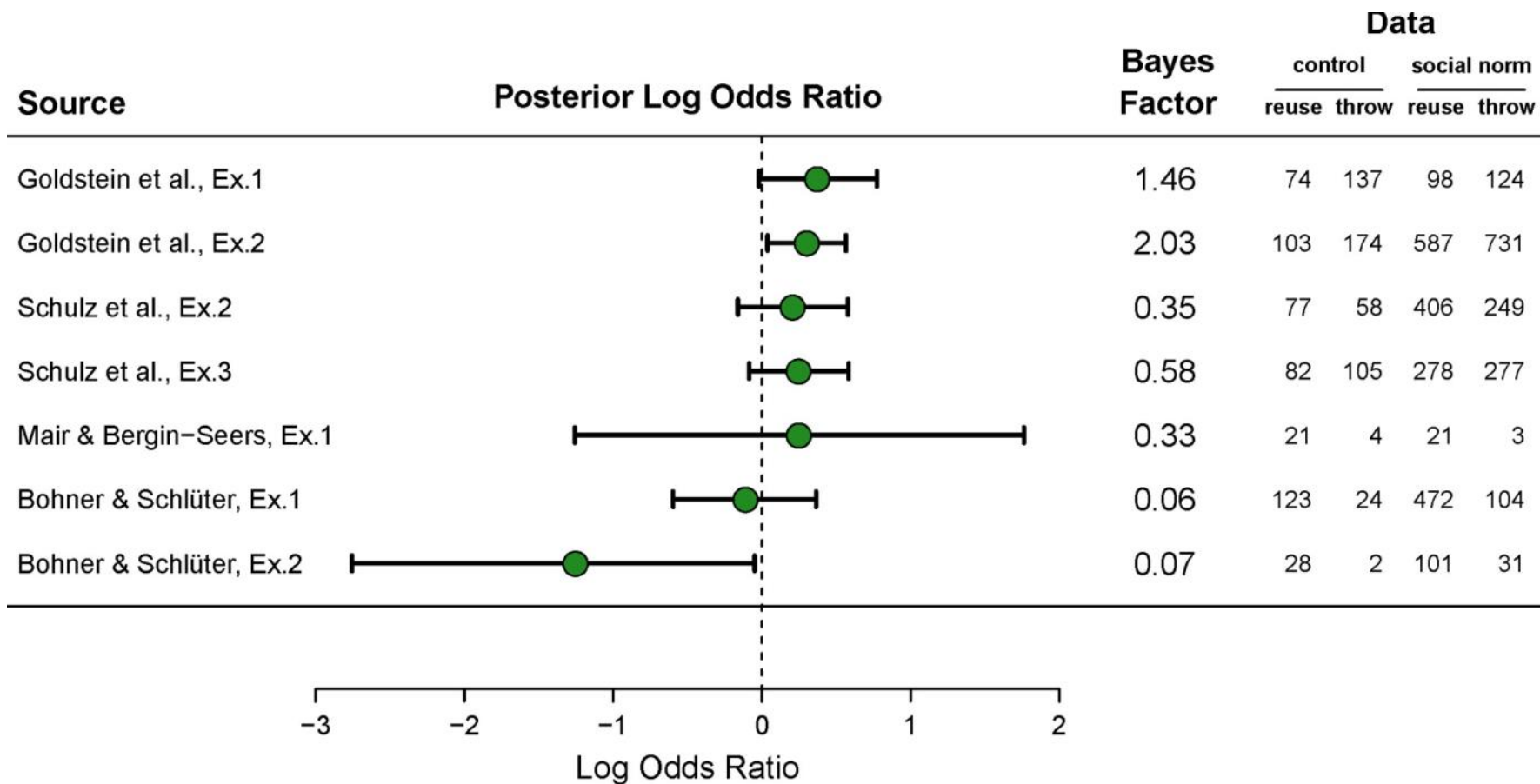
} all p's > .14



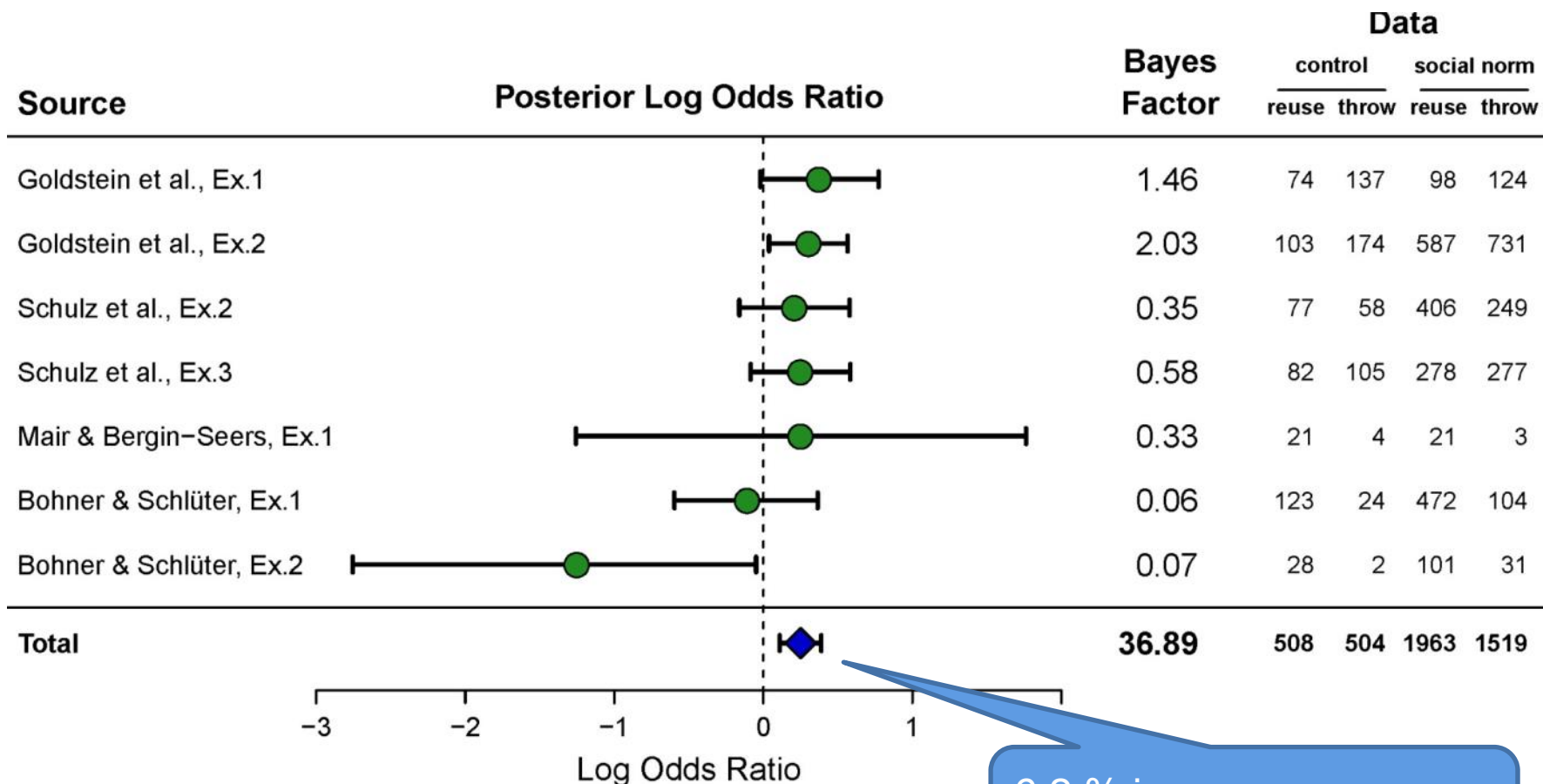
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Can Social Norms Increase Towel Reuse?



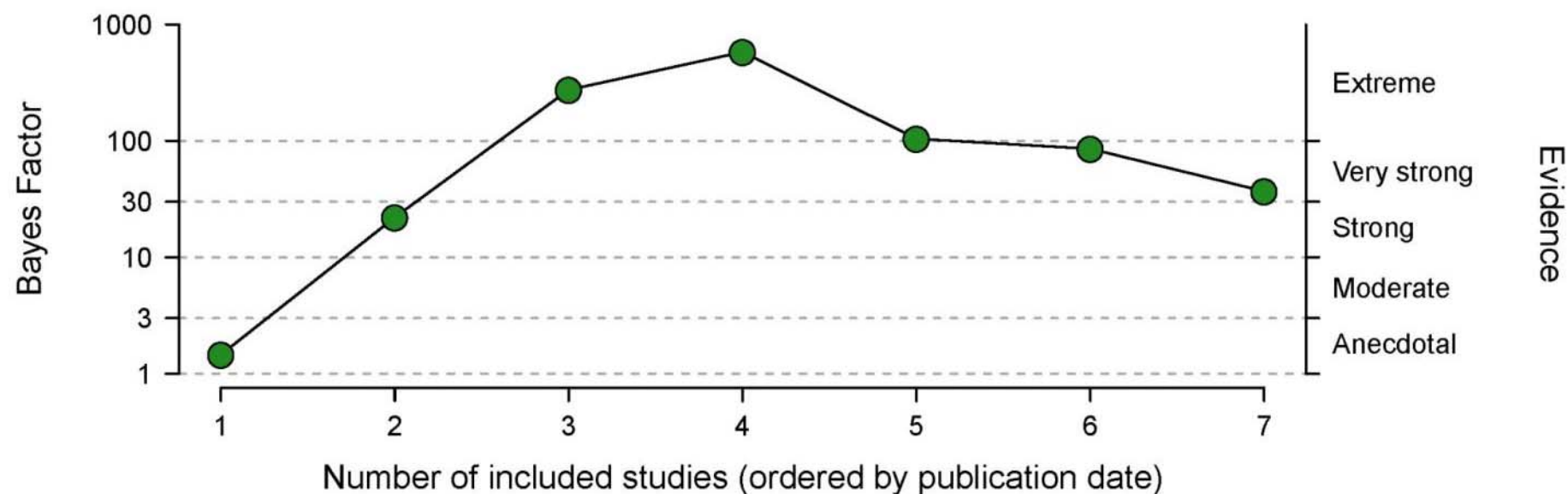
Can Social Norms Increase Towel Reuse?



6.2 % increase
Weighted with BF: 6.0%

Can Social Norms Increase Towel Reuse?

- Cumulative evidence “as the data comes in”
- Can be updated in the future



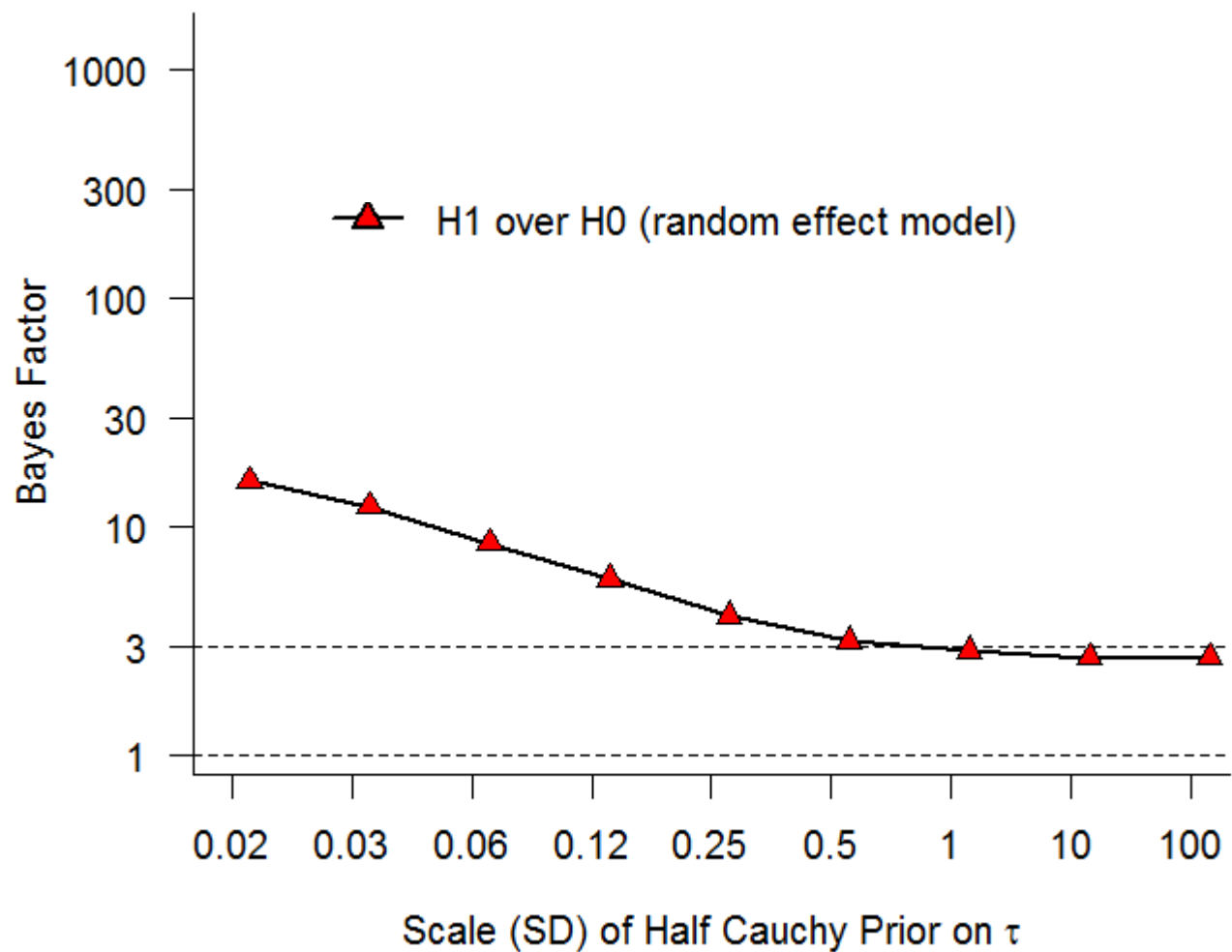
Repeated Testing is Allowed

- Multiple tests in NHST = risk of false-positive findings
- In a Bayesian framework, optional stopping does not distort the results
- No correction for repeated testing required

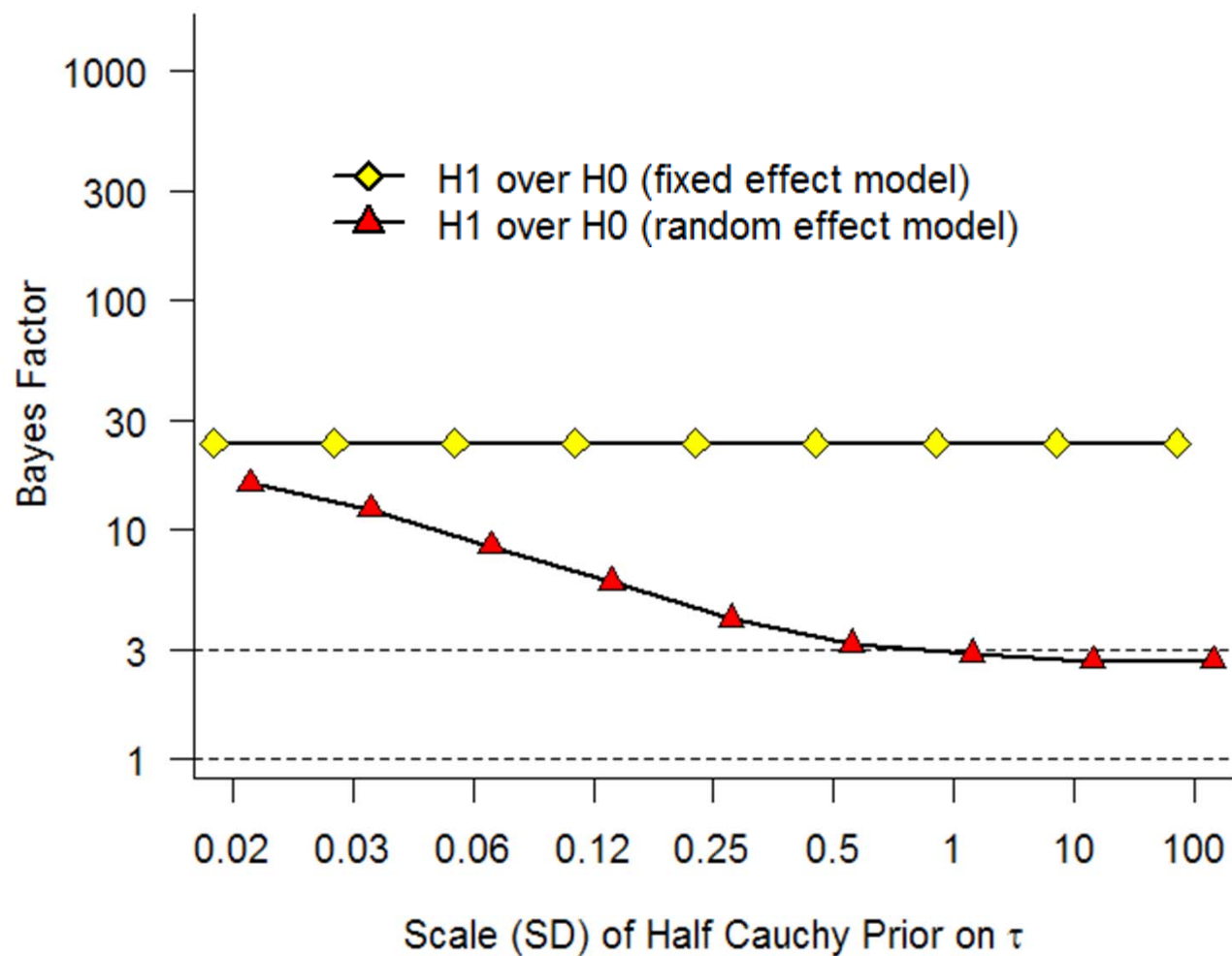
“It follows that any significance test based on Bayes’s theorem does not depend on the sequential stopping rule used [...]”

Dennis Lindley (1957)

Would a Random-Effect Model be Better Suited?



Would a Random-Effect Model be Better Suited?



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Compare different models and hypotheses

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Broad application

- “One-stop shop“



How can we Compare Models?

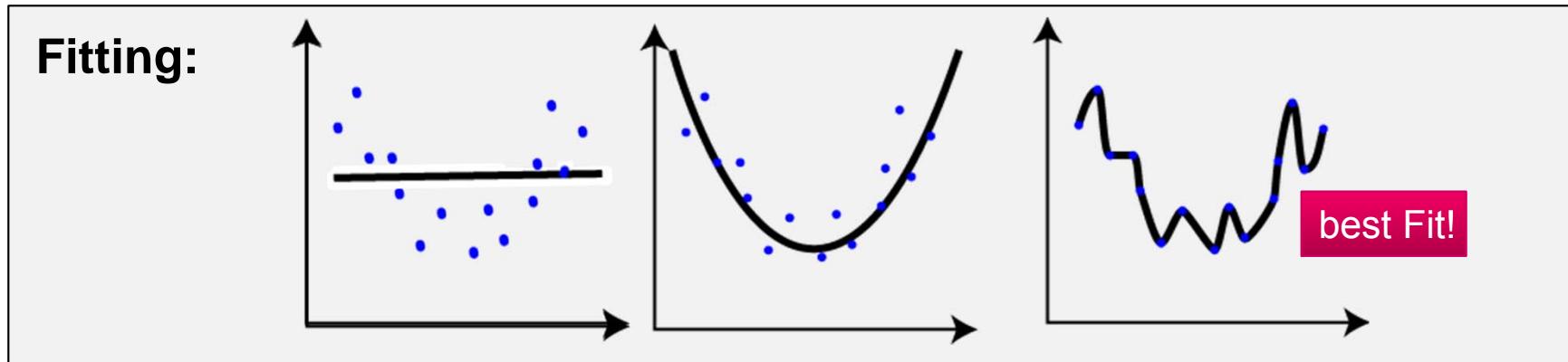


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How can we Compare Models?

More complex models provide a better *description* of the data...

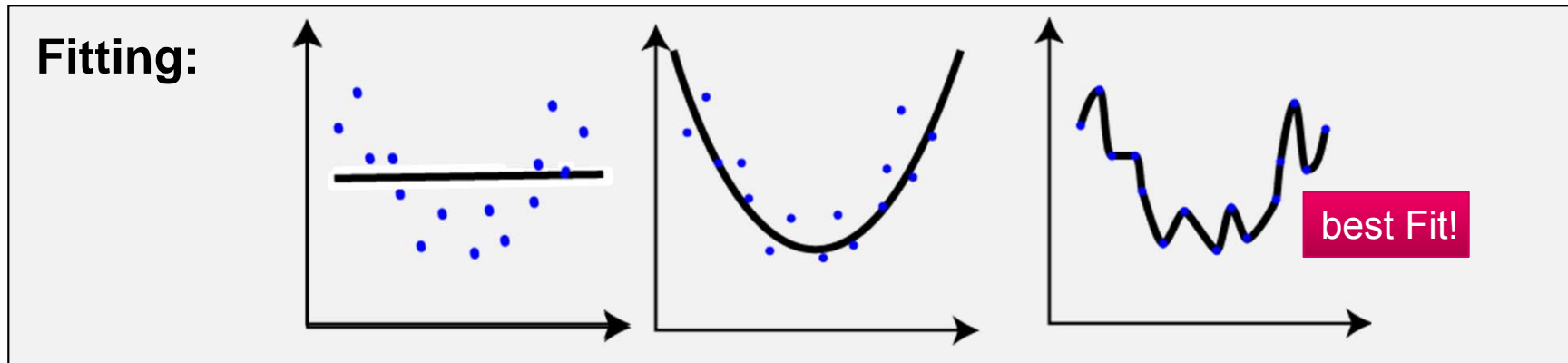


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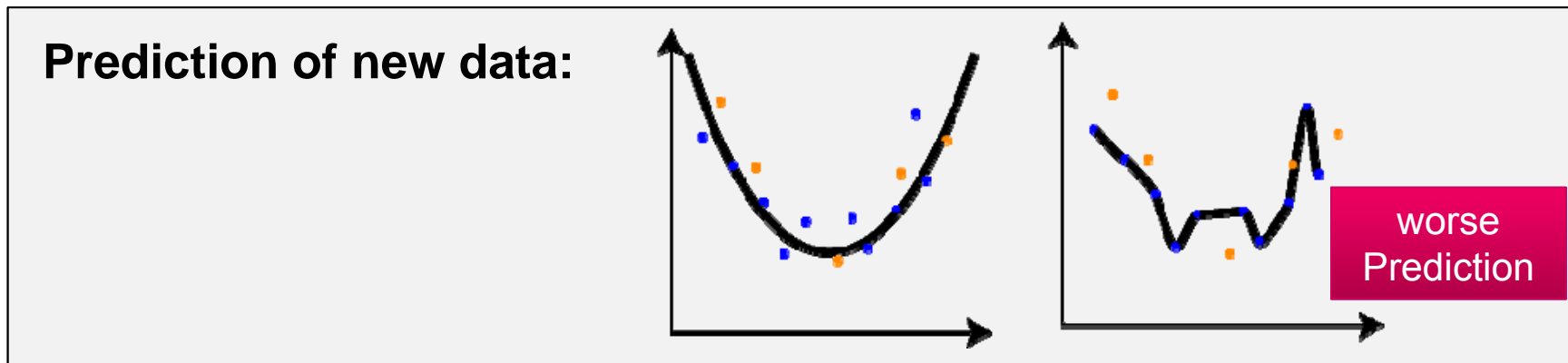
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...but not necessarily a better *explanation*:

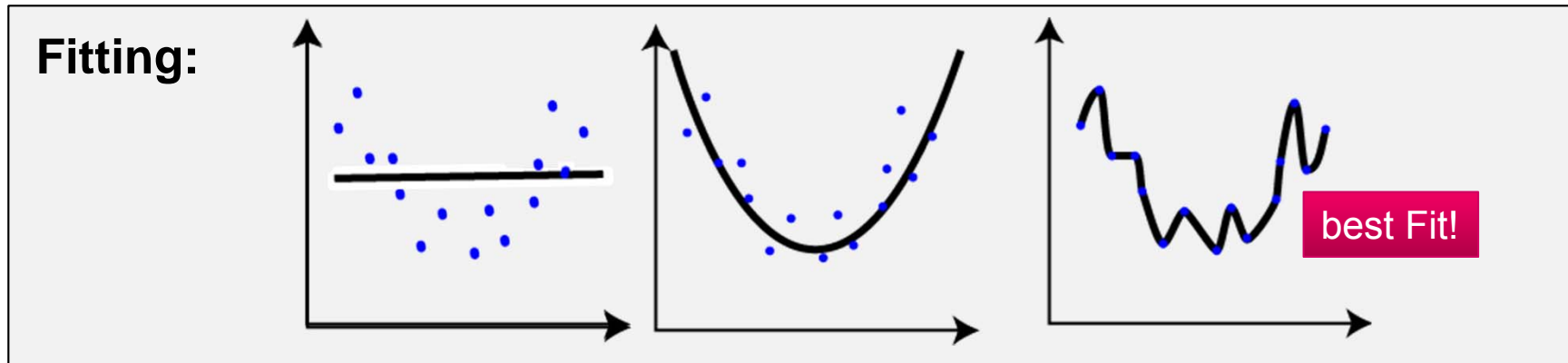


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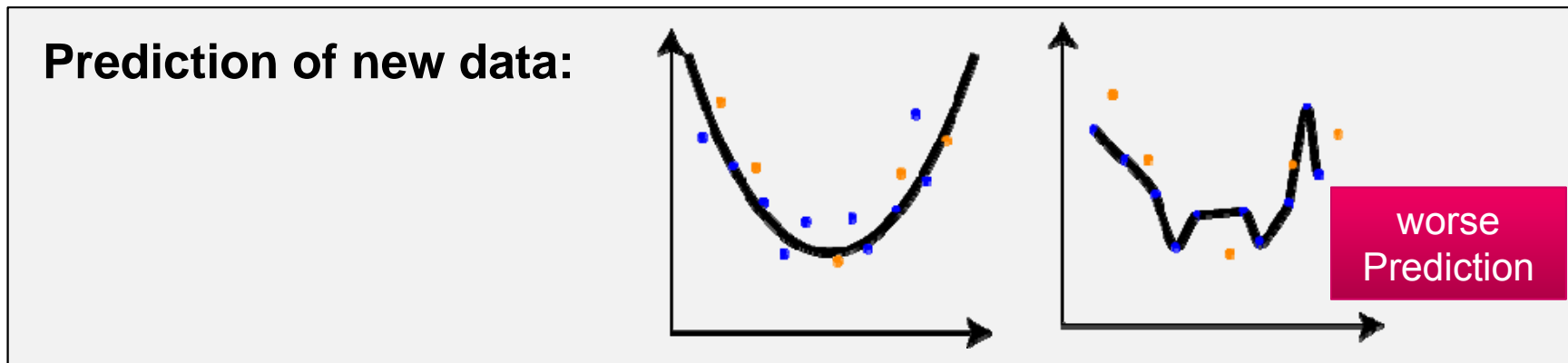
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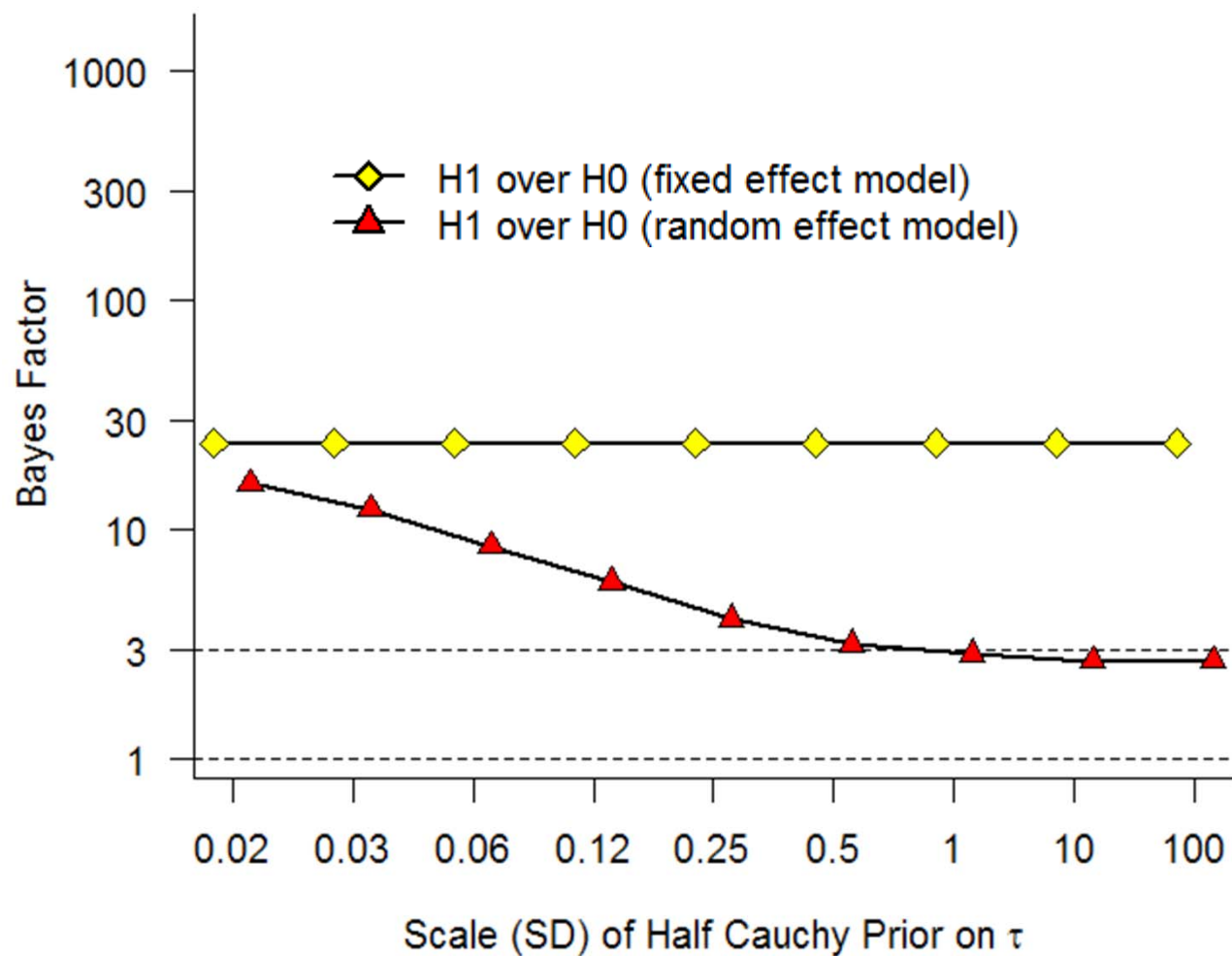
Bayes Factor

average probability of the observed data across all predictions the model makes

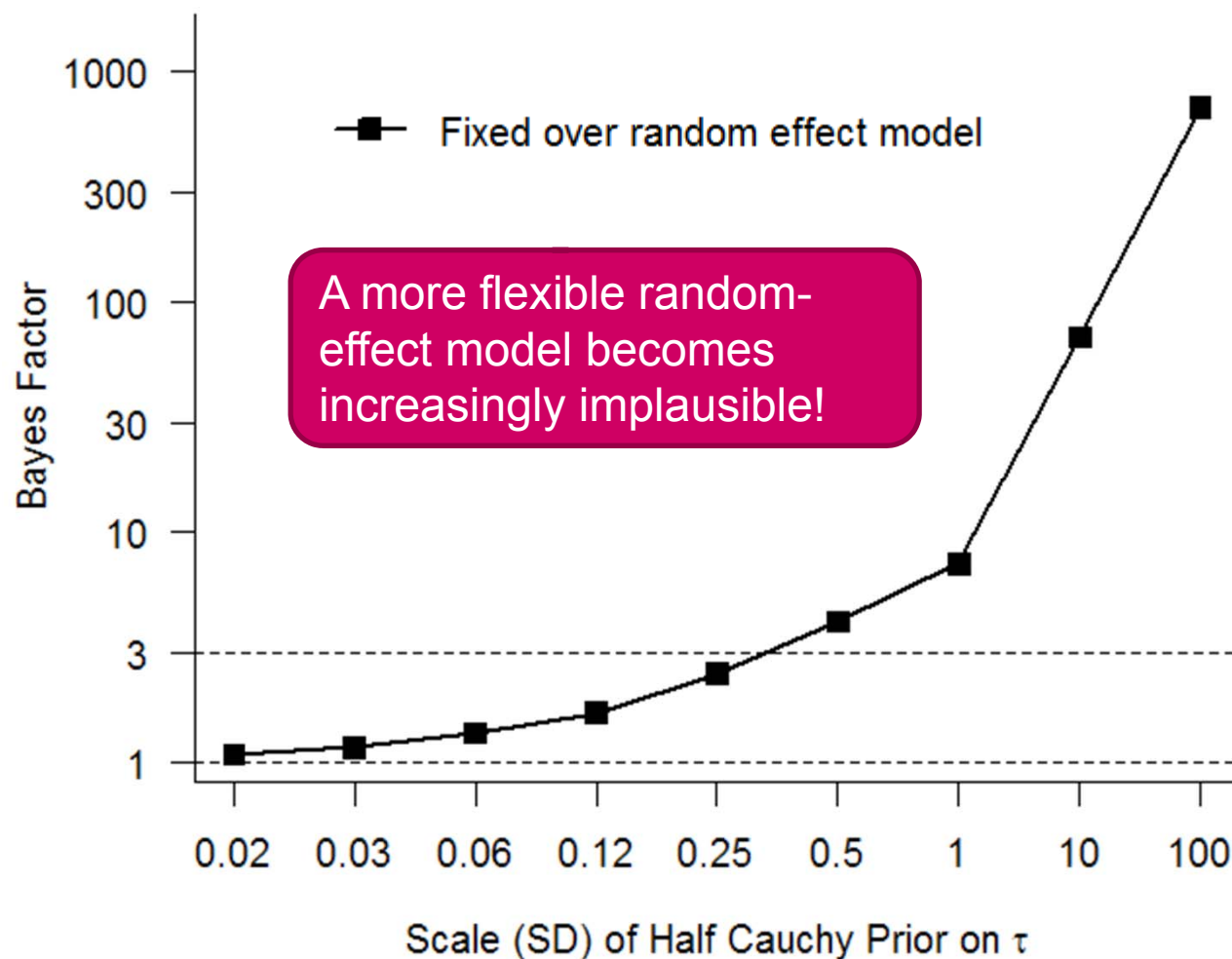
$$\frac{\int p(D|\theta_1, M_1) \cdot p(\theta_1, M_1)}{\int p(D|\theta_2, M_2) \cdot p(\theta_2, M_2)}$$

- Flexible models make many predictions
- If most of these predictions are wrong, this drives down the average probability

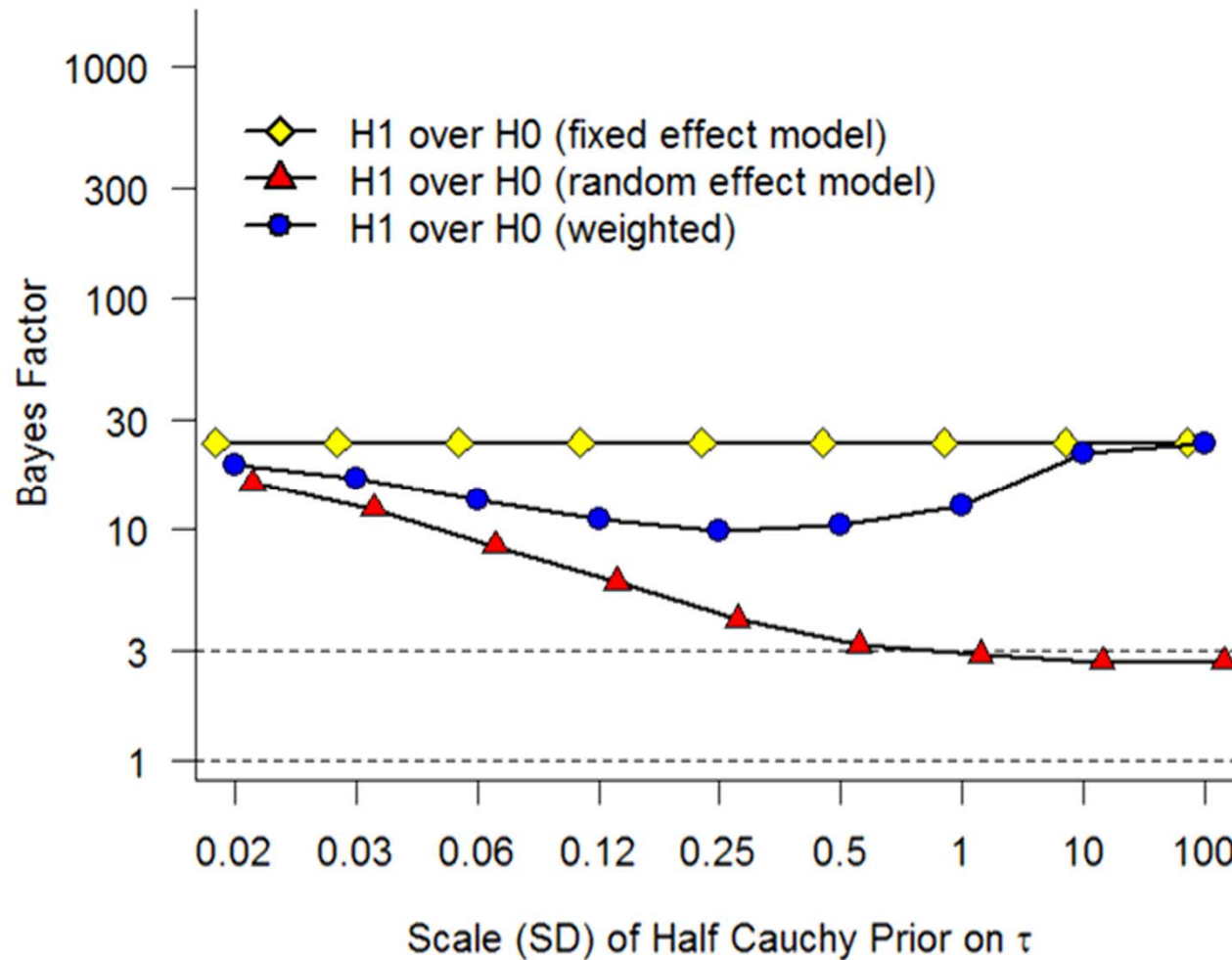
Would a Random-Effect Model be Better Suited?



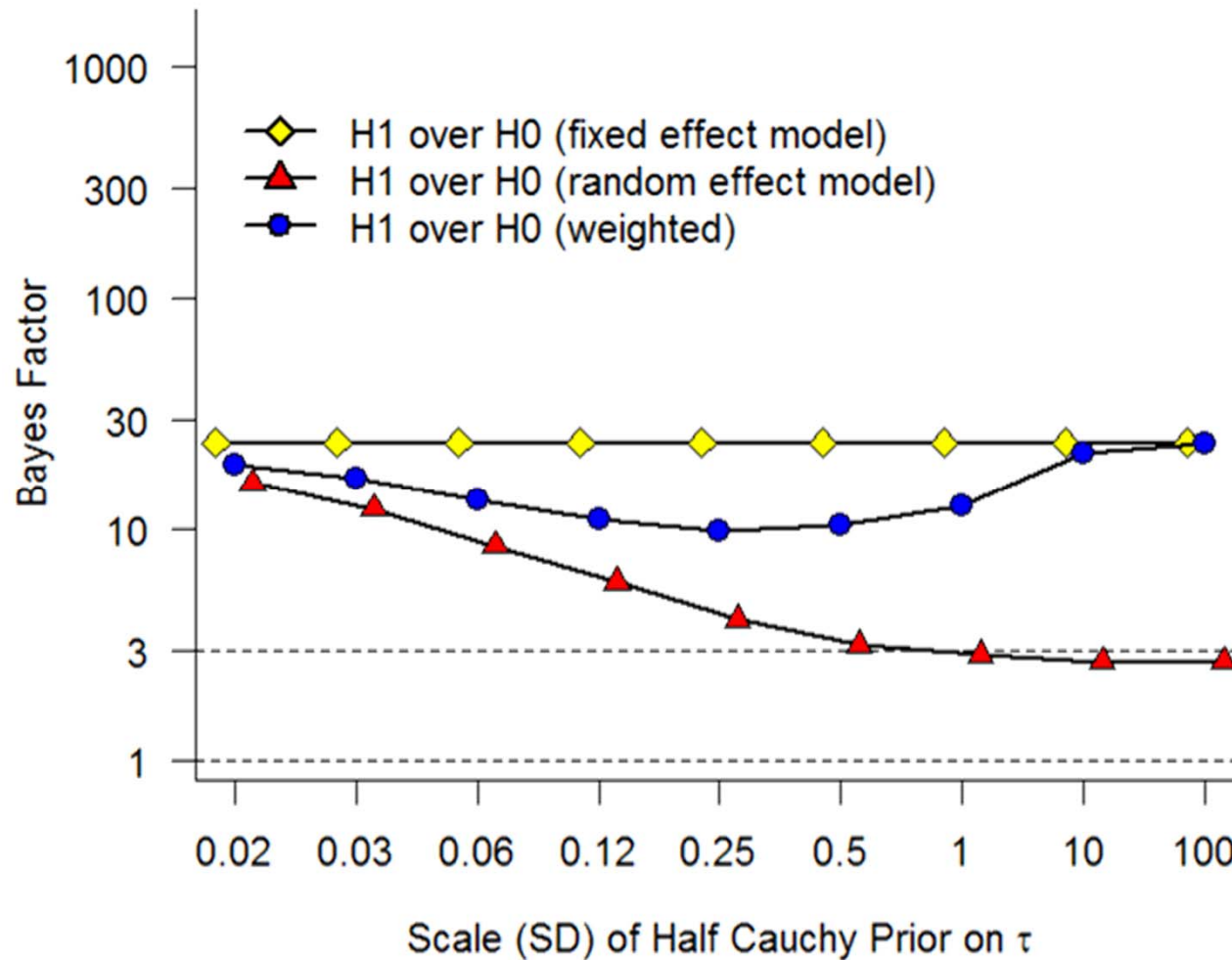
Would a Random-Effect Model be Better Suited?



Resolution: Bayesian Model Averaging

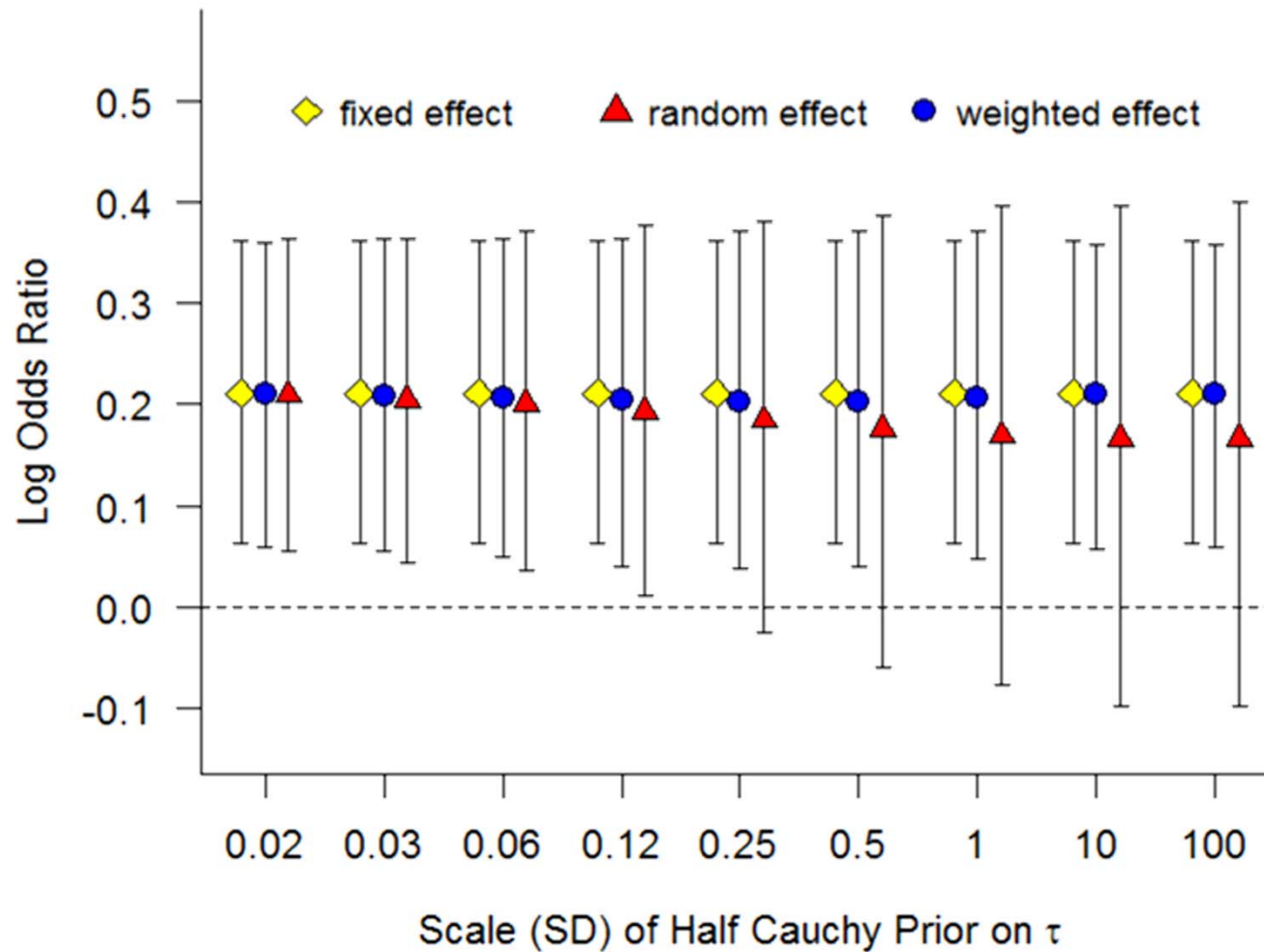


Resolution: Bayesian Model Averaging



= Prior on the between-study variance

Resolution: Bayesian Model Averaging



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Broad Application

- Well suited for estimating more complex models
- Lends itself to hierarchical designs



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Advantages of Bayesian Approach

- Probabilities for hypotheses
- Able to distinguish between “data support H” and “data are not diagnostic for H”
- Easy to interpret
- Can be updated as more data are collected
- Possibility to include prior knowledge
- All assumptions are explicit
- Model comparison: Occam's razor
- Well-suited for big data
- High "street credibility"





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