

Prior Criticism in Bayesian Meta-Analysis

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Estimation vs. Criticism

J. R. Statist. Soc. A, (1980),
143, Part 4, pp. 383–430

Sampling and Bayes' Inference in Scientific Modelling and Robustness

By GEORGE E. P. BOX

University of Wisconsin–Madison

[Read before the ROYAL STATISTICAL SOCIETY at a meeting organized by the South Wales Local Group on Thursday, May 15th, 1980, the President SIR CLAUS MOSER in the Chair]

Box (1980) distinguishes

- Estimation based on posterior distribution $f(\theta \mid \text{data}, \text{assumptions})$

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- **Estimation** based on **posterior distribution** $f(\theta \mid \text{data}, \text{assumptions})$
- **Criticism** based on **prior-predictive distribution** $f(\text{data} \mid \text{assumptions})$
- The assumptions include both **model** and **prior** assumptions.

Box's Tail Probability

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$$p_{\text{Box}} = \Pr \{ f(\text{data} \mid \text{assumptions}) < f(\text{observed data} \mid \text{assumptions}) \}$$

- Quantifies **compatibility** of model/prior with **observed data**.
- Small values of p_{Box} indicate that model or prior are discredited by the observed data.

The Q-Test

- Model: Normal-normal model
 - Prior: $\tau^2 = 0$
 - Data: Differences between effect estimates $\hat{\theta}_i$ with standard errors σ_i
- $f(\text{observed data} \mid \text{assumptions})$ is the **Q-statistic**

$$Q = \frac{\sum_{i < j} w_i w_j (\hat{\theta}_i - \hat{\theta}_j)^2}{\sum_{i=1}^k w_i}$$

where $w_i = 1/\sigma_i^2$ are “**fixed-effect**” weights.

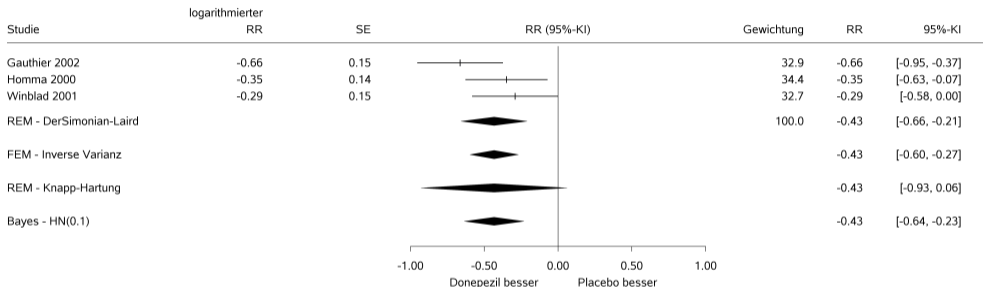
→ p_{Box} simplifies to the p -value obtained from the null distribution $Q \sim \chi_{k-1}^2$

The Generalized Q-Test

- Model: Normal-normal model
- Null hypothesis: $\tau^2 = \tau_0^2$
- The generalized Q-statistic $Q(\tau_0^2)$ now uses “random-effects” weights $w_i = 1/(\sigma_i^2 + \tau_0^2)$.
- We still have $Q(\tau_0^2) \sim \chi_{k-1}^2$ if $\tau^2 = \tau_0^2$
- Solving $Q(\tau^2) = k - 1$ for τ^2 gives the Paule-Mandel estimate.

Example: Donepezil vs. Placebo

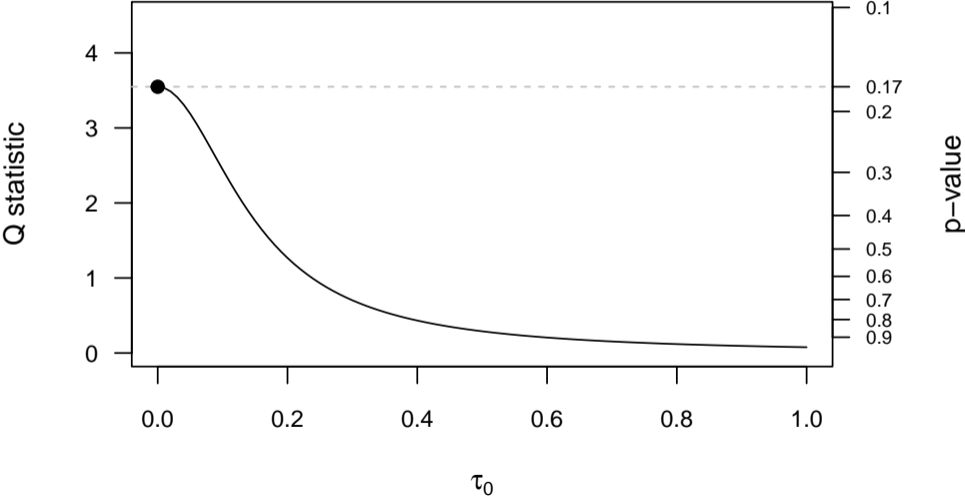
Donepezil vs. Placebo
DAD, CMCS, PDS



Heterogenität: $Q=3.70$, $df=2$, $p=0.157$, $I^2=45.9\%$

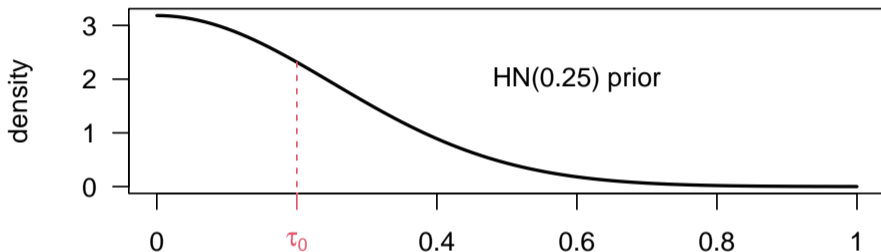
Gesamteffekt (REM - DerSimonian-Laird): $Z\text{-Score}=-3.78$, $p<0.001$, $\text{Tau}=0.135$

Example: Donepezil vs. Placebo



Checking the Heterogeneity Prior

- Suppose we now have a **half-normal prior** $f(\tau)$ with $E(\tau) = \tau_0$

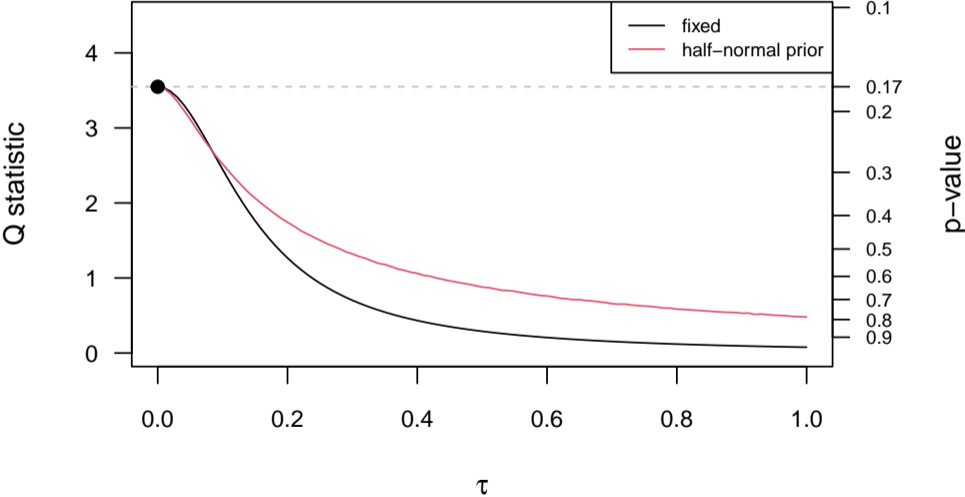


→ p_{Box} is now based on

$$\tilde{Q} = \int_0^{\tau} Q(\tau^2) f(\tau^2) d\tau^2$$

- The χ^2 -distribution still holds because τ^2 is a pivot for $Q(\tau^2)$
- p_{Box} can be easily calculated through Monte Carlo simulation.

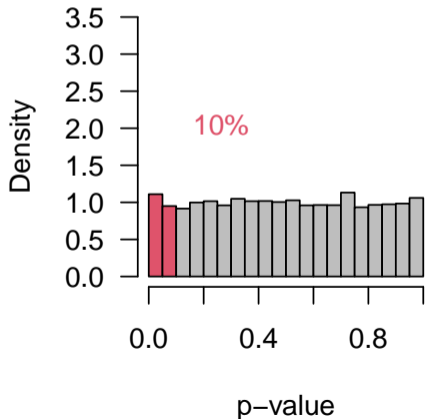
Example: Donepezil vs. Placebo



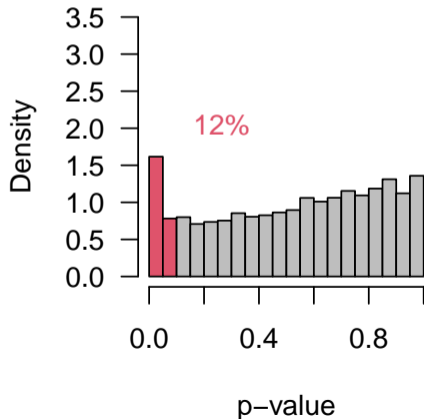
Type-I Error Assessment

$$k = 3, \tau_0 \sim f(\tau)$$

Normal distribution



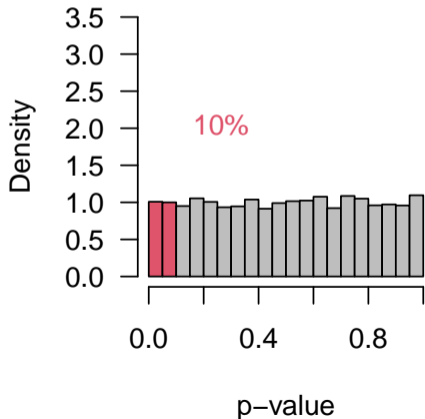
t(4) distribution



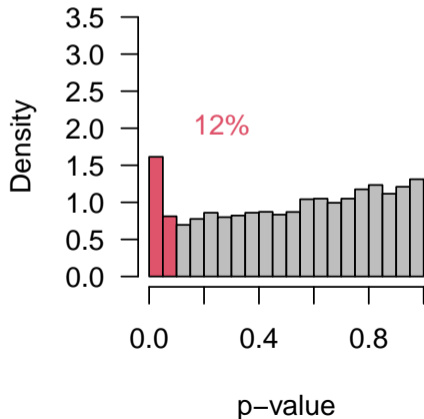
Power Assessment

$k = 3$, fixed $\tau_0 = 0.2 = E(\tau)$

Normal distribution



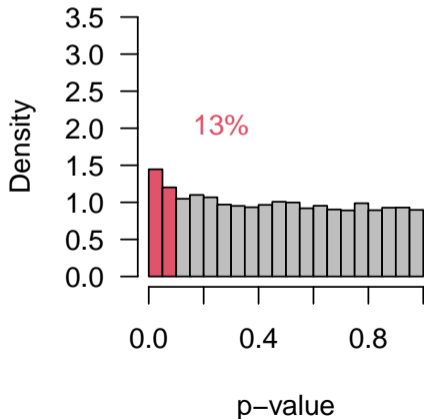
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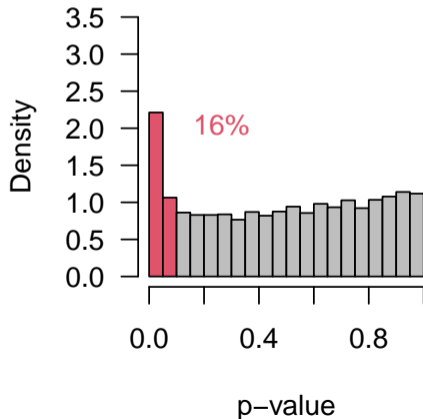
Power Assessment

$k = 3$, fixed $\tau_0 = 0.4 = 2E(\tau)$

Normal distribution



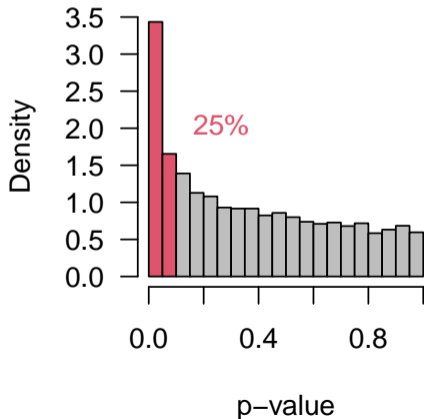
t(4) distribution



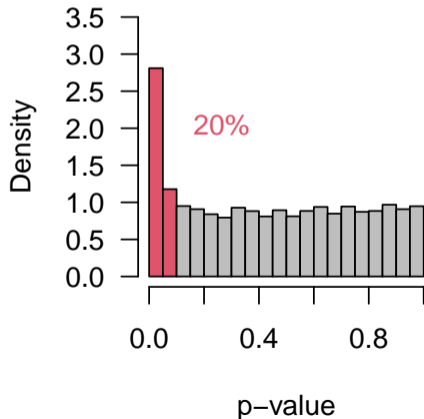
Power Assessment

$k = 3$, fixed $\tau_0 = 0.8 = 4E(\tau)$

Normal distribution



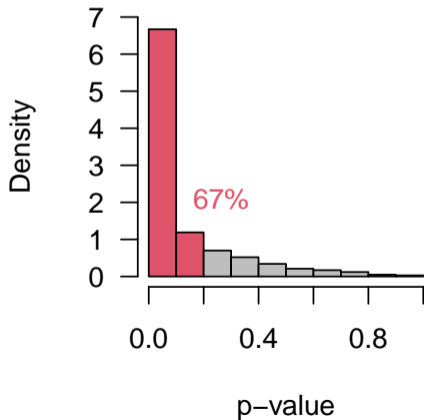
t(4) distribution



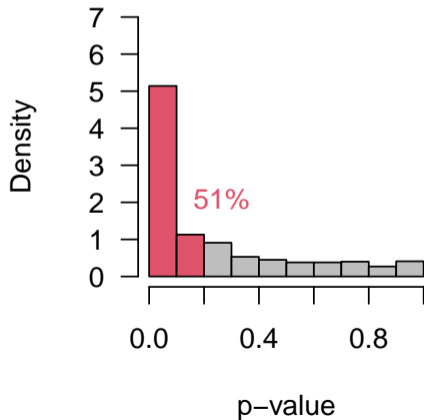
Aggregated Power Assessment for 10 Studies

$k = 3$, fixed $\tau_0 = 0.8 = 4E(\tau)$

Normal distribution



t(4) distribution



Summary

- Q-Test can be generalized to check **heterogeneity prior** and other model assumptions.
- Has **low power** for meta-analyses with very small k .
- Power can be increased by **summation** of Q-statistic across meta-analyses.

Summary and Discussion

- Q-Test can be generalized to check **heterogeneity prior** and other model assumptions.
- Has **low power** for meta-analyses with very small k .
- Power can be increased by **summation** of Q-statistic across meta-analyses.

- Heterogeneity prior is not the only critical assumption
- Normality assumption may also be wrong due to **publication bias** etc.

References

Box, G. E. P. (1980). Sampling and Bayes' inference in scientific modelling and robustness (with discussion). Journal of the Royal Statistical Society, Series A, 143:383–430.