
Meta analysis and the curse of publication bias

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The three stages of meta analysis:

- Literature search and systematic review of relevant studies
- Statistical summary of each study
 - Study estimates $\hat{\theta}_i$
 - Within-study variances σ_i^2
- Combining summary statistics into an overall inference
 - fixed effects model

$$\hat{\theta}_i \sim N(\theta, \sigma_i^2)$$

- MLE = $\tilde{\theta} = \frac{\sum w_i \hat{\theta}_i}{\sum w_i}$, $w_i = \frac{1}{\sigma_i^2}$

- $\text{Var}\{\tilde{\theta}\} = \frac{1}{\sum w_i}$

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An early example of a controlled trial in the treatment of juvenile offenders:

Klein *et al.* (*J Consult. Clin. Psychol.* 1977, **3**, 469-474)

	success	failure	%
structured counselling	34	12	74
control	29	27	52

$$\hat{\theta} = \pm \sqrt{\chi^2/n} = 0.227$$

$$\hat{\theta} \sim N(\theta, \sigma^2) , \sigma^2 = 1/n = 1/102$$

$$t = \hat{\theta}/\sigma = 2.29 , P = \Phi(-2.27) < 5\%$$

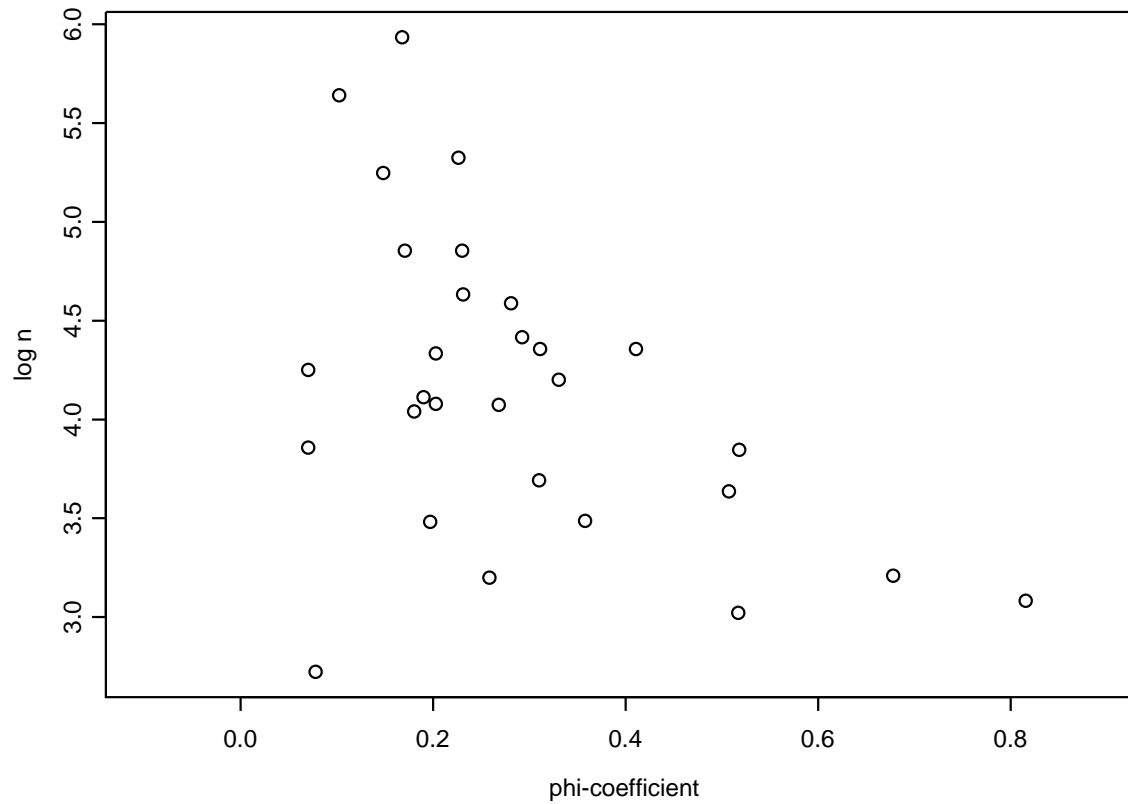
Conclusion : new treatment is effective in reducing recidivism

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One of the first systematic reviews of corrections studies:

Andrews et al. (Criminology, 1990, 28, 369-429)

Funnel plot for corrections meta analysis



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Corrections meta analysis:

k studies with data $(\hat{\theta}_i, \sigma_i)$ for $i = 1, 2, \dots, k$

Conventional (fixed effects) meta analysis estimate is

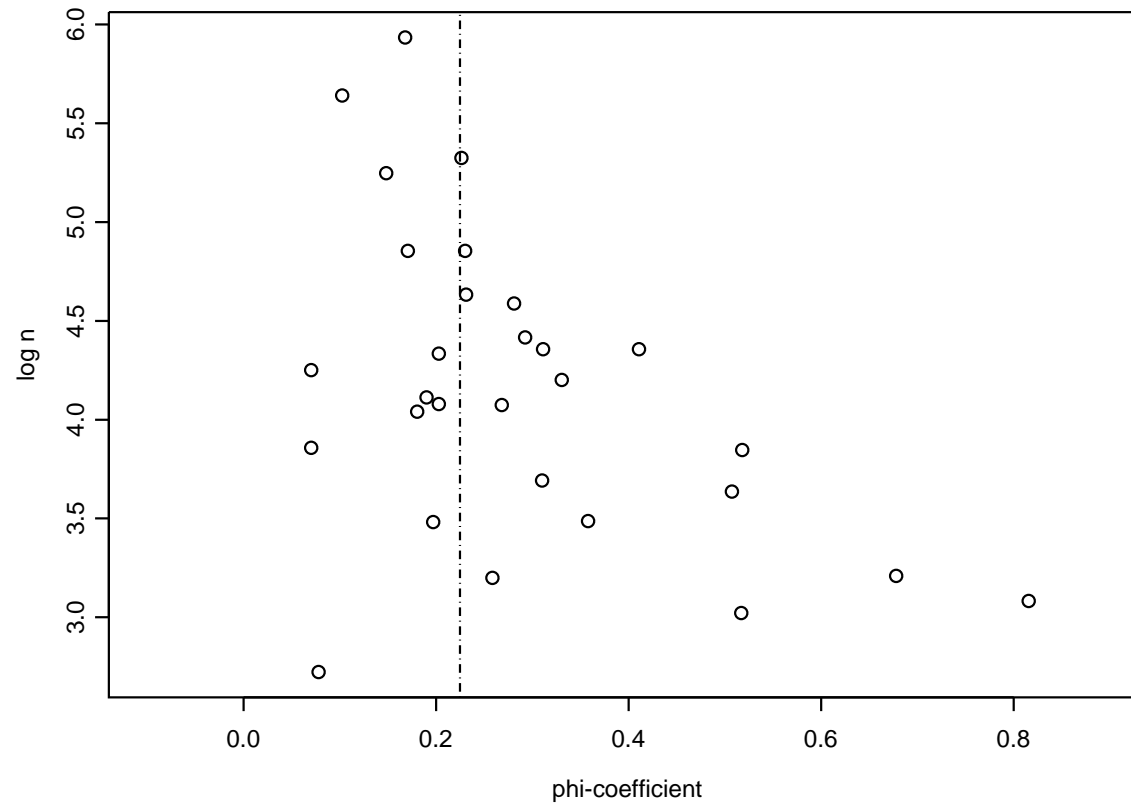
$$\hat{\theta} = \frac{\sum w_i \hat{\theta}_i}{\sum w_i}, \text{Var}(\hat{\theta}) = \frac{1}{\sum w_i}, w_i = \frac{1}{\sigma_i^2}$$

$$k = 29, \hat{\theta} = 0.225, SD = 0.0198$$

cf. the Klein study: $\hat{\theta} = 0.227, SD = 0.0990$

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Funnel plot for corrections meta analysis



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Second example:

What is the effect of passive smoking on the risk of lung cancer?

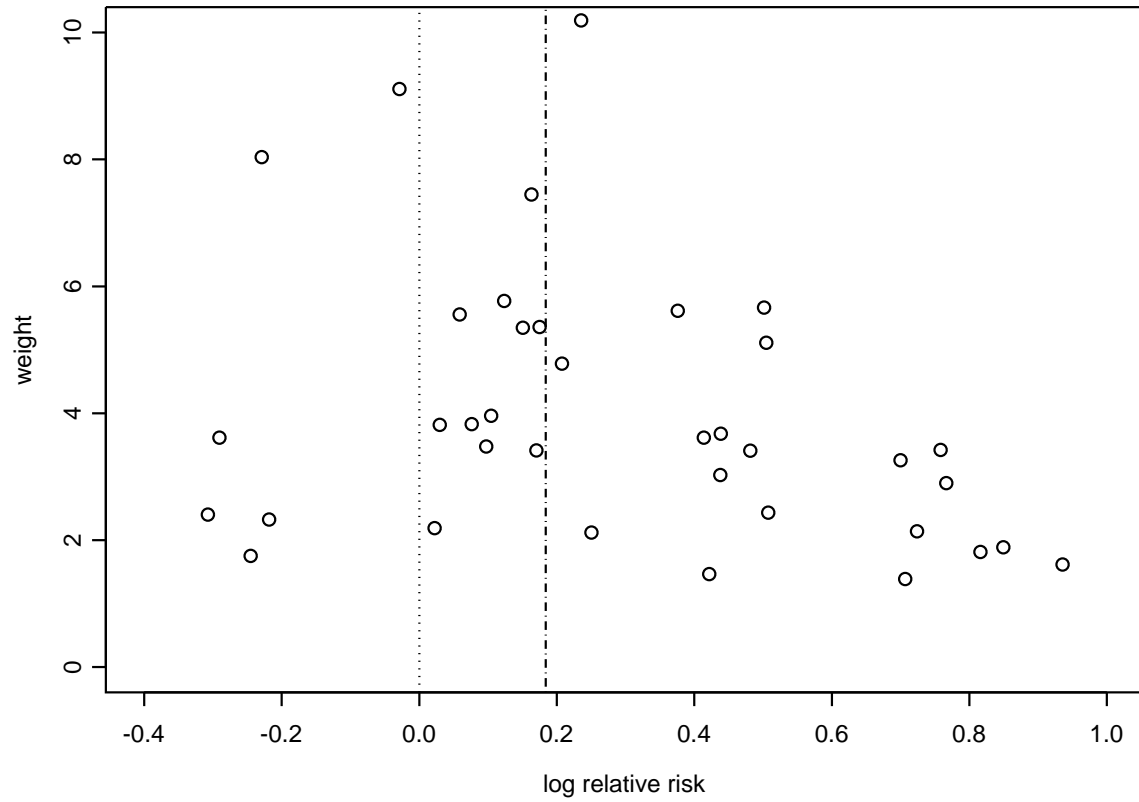
Meta analysis of 37 case-control studies quoted by *SCOTH* (Hackshaw *et al.*, *BMJ* 1997, **315**, 980-988)

$\hat{\theta}$ = estimated log relative risk = 0.184 , $SD = 0.037$

\Rightarrow risk increase $\in (12\%, 30\%)$

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Funnel plot for smoking meta analysis



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Third example:

Yusuf *et al.* (1993)

Meta analysis of 15 clinical trials on the effectiveness of intravenous magnesium in acute myocardial infarction

$$\theta = \log \frac{P(\text{death} \mid \text{treatment})}{P(\text{death} \mid \text{control})}$$

$$\text{Relative risk} = \exp\{\tilde{\theta}\} = .58(.46, .73)$$

$$\text{P-value} \approx 2 \times 10^{-6}$$

Published conclusion: "magnesium is an effective, safe, simple and inexpensive intervention that should be introduced into clinical practice without delay"

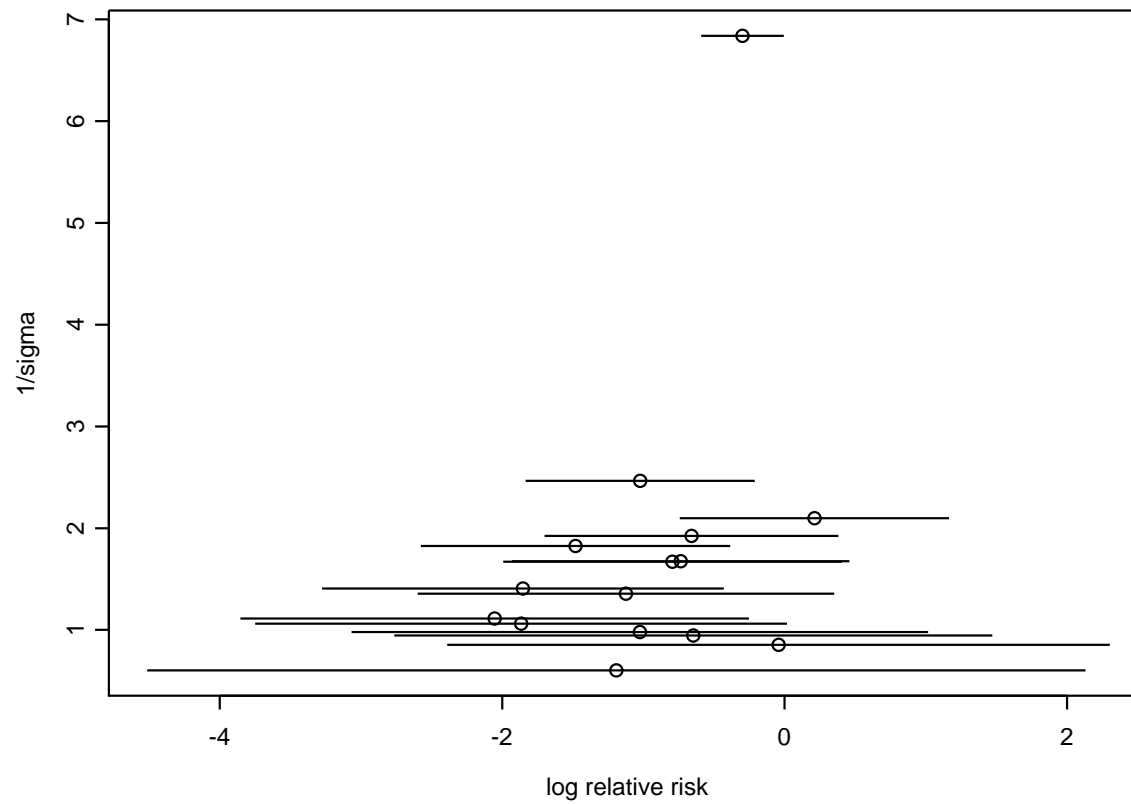
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This was soon contradicted by ISIS-4 (1995), a very large multi-centre randomized clinical trial which gave mortality rates

- 2216/29011 (magnesium)
- 2103/29039 (control)
- Relative risk = 1.06(0.99, 1.13)
- P-value ≈ 0.09

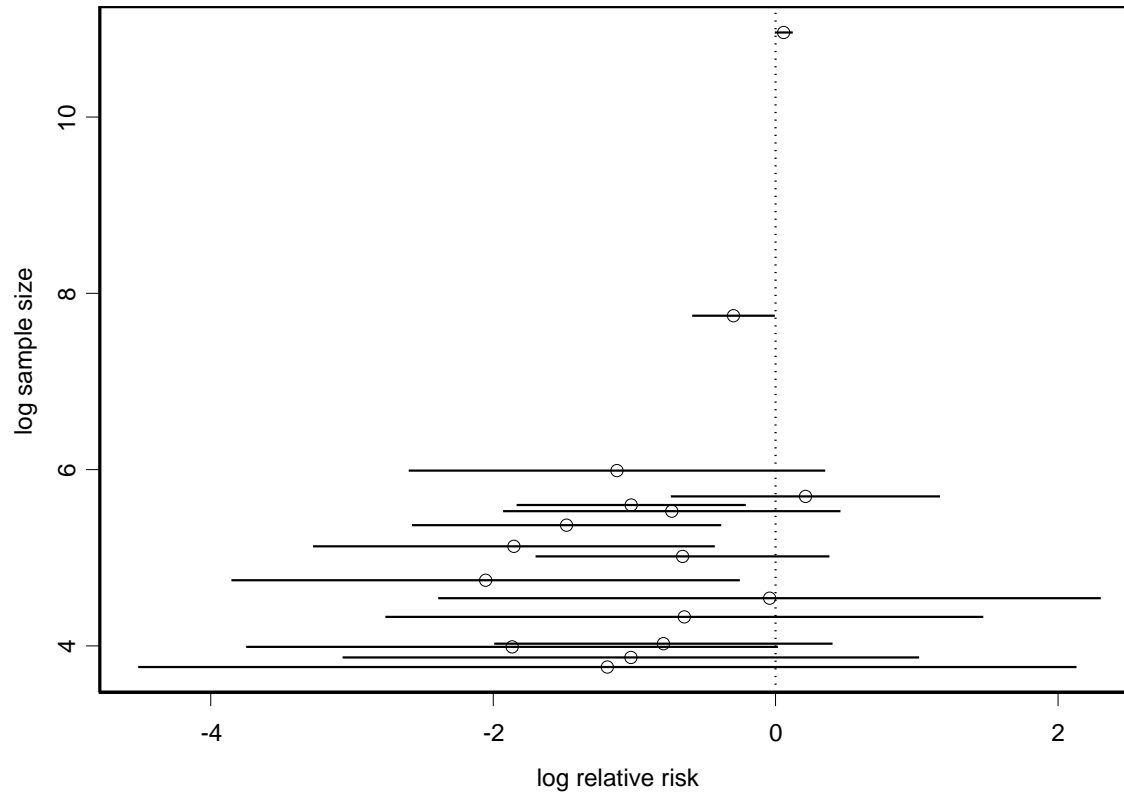
Conclusion: there is **no significant difference**, magnesium may in fact be harmful.

Funnel plot for magnesium studies



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Magnesium Meta Analysis



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To appear in a meta analysis a study has to be

- written up
- submitted
- accepted for publication
- found by the reviewer

Conjecture

Studies reporting a significant result are more likely to survive this selection process

⇒ the meta analysis will be biased

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Probit selection model

$$\hat{\theta}|\sigma \sim N(\theta, \sigma^2)$$

$$P(\text{select}|\hat{\theta}, \sigma) = \Phi(\alpha + \beta\hat{\theta}/\sigma)$$

$$P(\text{select}|\sigma) = \Phi \left\{ \frac{\alpha + \beta\theta/\sigma}{(1 + \beta^2)^{\frac{1}{2}}} \right\}$$

$$P(\sigma|\text{selection}) = \frac{P(\text{select}|\sigma)P(\sigma)}{P(\text{select})}$$

$$P(\sigma) = \frac{P(\text{select})P(\sigma|\text{select})}{P(\text{select}|\sigma)}$$

$$\int P(\sigma)d\sigma = P(\text{select}) \int \frac{1}{P(\text{select}|\sigma)} P(\sigma|\text{select})d\sigma$$

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$$\begin{aligned} P(\text{select}) &= [\mathbf{E}_\sigma \{ P(\text{select}|\sigma)^{-1} | \text{select} \}]^{-1} \\ &\approx \left(\frac{1}{n} \sum \left[\Phi \left\{ \frac{\alpha + \beta\theta/\sigma_i}{(1 + \beta^2)^{\frac{1}{2}}} \right\} \right]^{-1} \right)^{-1} \\ &= p(\alpha, \beta, \theta) \end{aligned}$$

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Log likelihood function is

$$L(\alpha, \beta, \theta) = -\frac{1}{2} \sum \frac{(\hat{\theta}_i - \theta)^2}{\sigma_i^2} + \sum \log \Phi(\alpha + \beta \hat{\theta}_i / \sigma_i) - \sum \log \Phi \left\{ \frac{\alpha + \beta \theta / \sigma_i}{(1 + \beta^2)^{\frac{1}{2}}} \right\}$$

For a sensitivity analysis we can fix

$$p(\alpha, \beta, \theta) = p$$

and find the conditional likelihood function over the two parameters
 (β, θ)

Sensitivity analysis

For any given value of p we can get the MLE $\hat{\theta}_p$ and the associated confidence limits $\{\theta_p^{(L)}, \theta_p^{(U)}\}$

Then

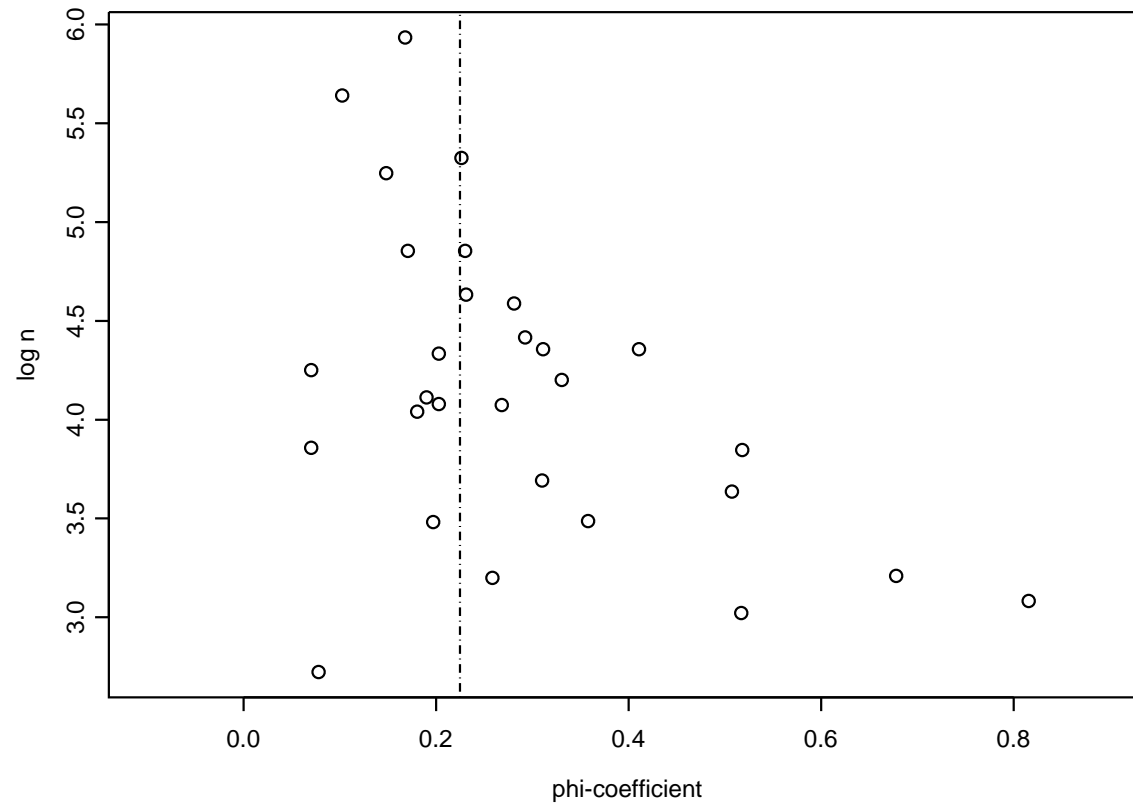
- $p = 1 \Rightarrow$ standard analysis (no bias)
- $p < 1 \Rightarrow$ allowing for selection bias

As p decreases from 1 we are adjusting for increasing bias

Note

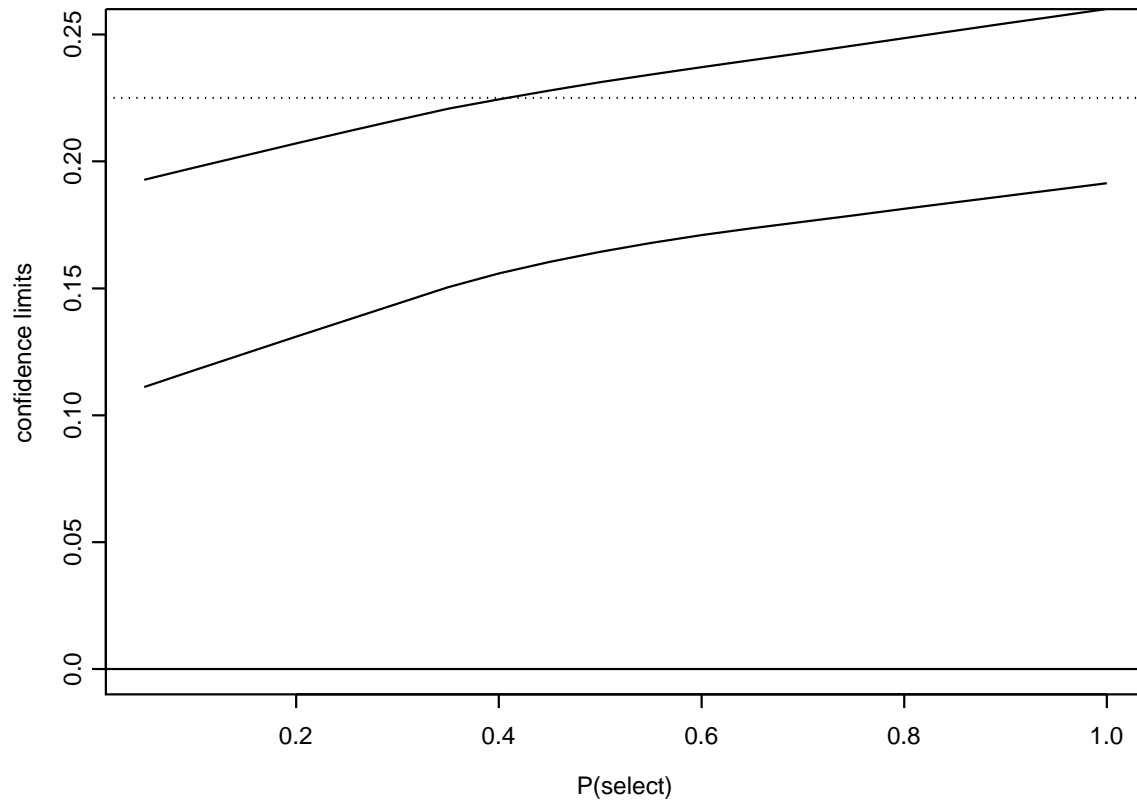
In practice we find that the profile likelihood for p is surprisingly flat (and non-robust), so it is virtually impossible to estimate p (i.e. to estimate the number of unpublished studies).

Funnel plot for corrections meta analysis



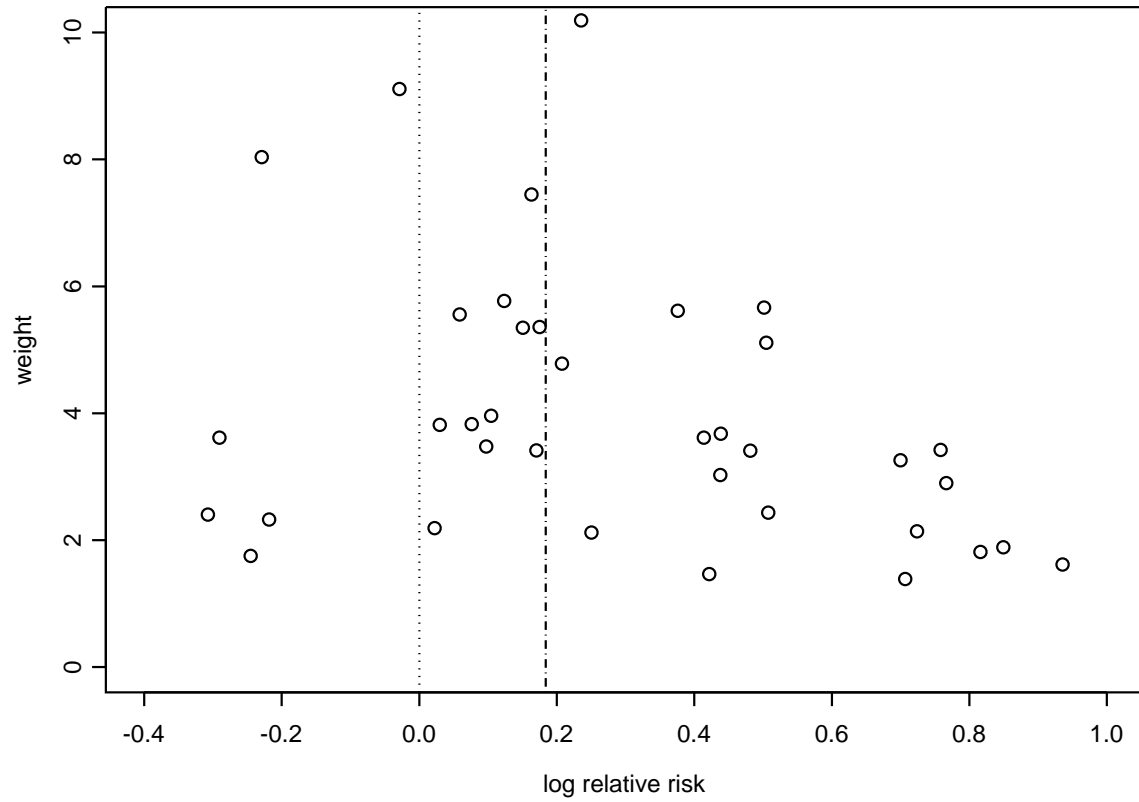
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Corrections Meta Analysis : confidence intervals for phi-coefficient



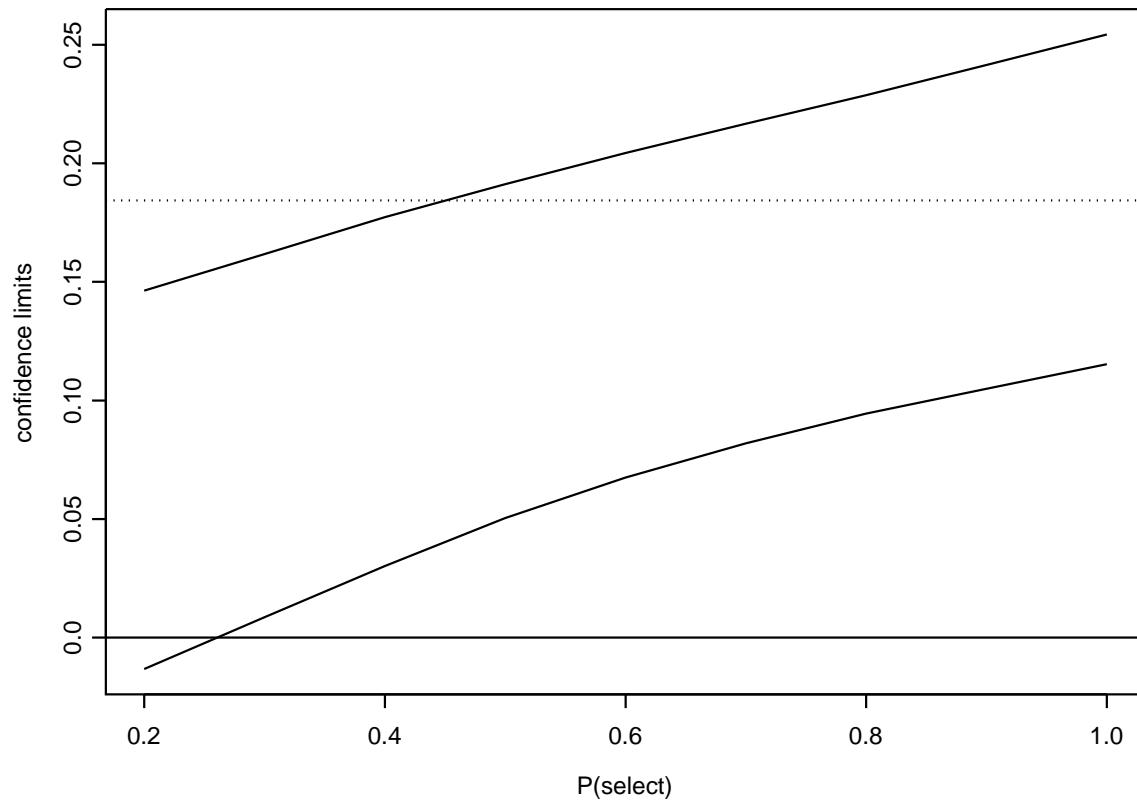
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Funnel plot for smoking meta analysis



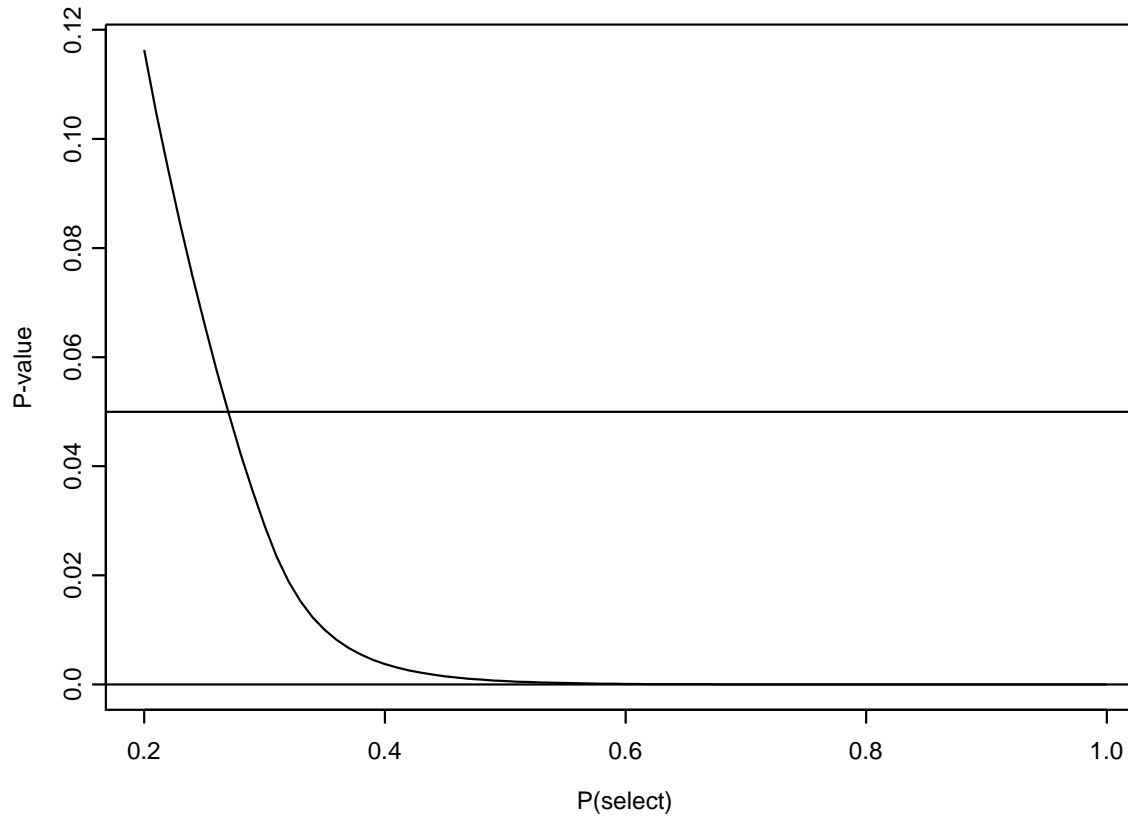
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Smoking Meta Analysis : confidence intervals for log relative risk



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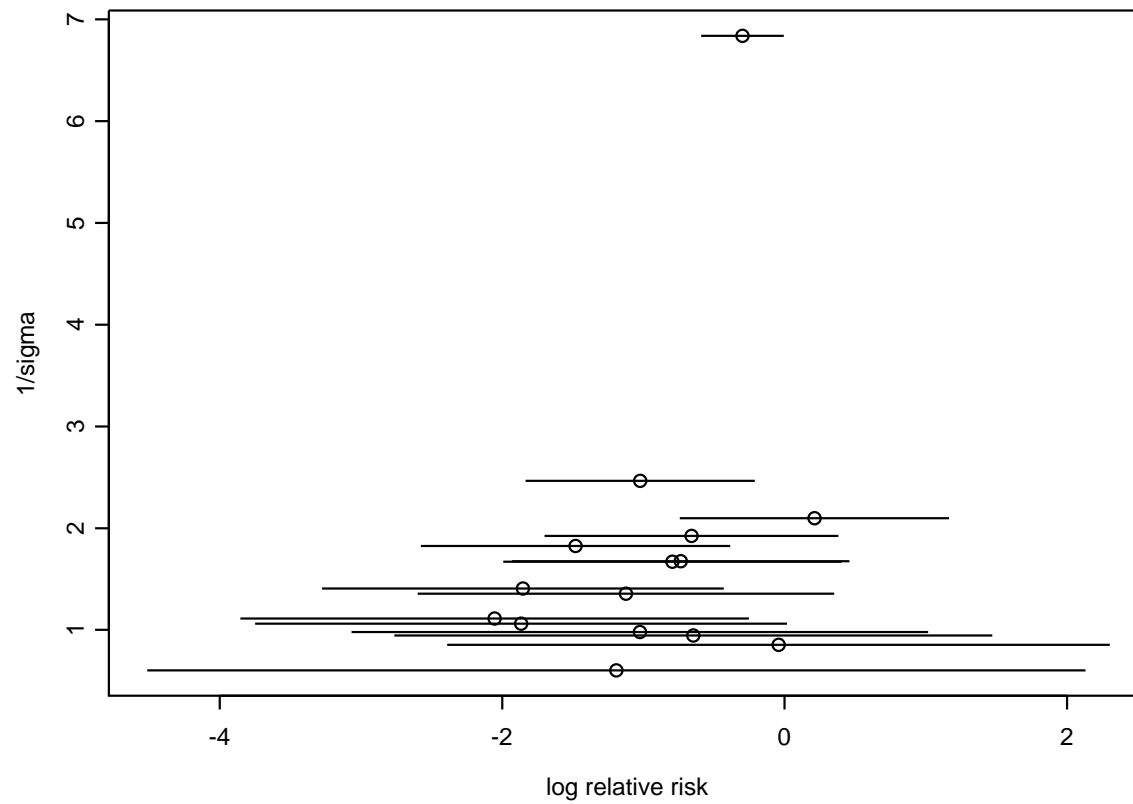
Smoking Meta Analysis : P-values



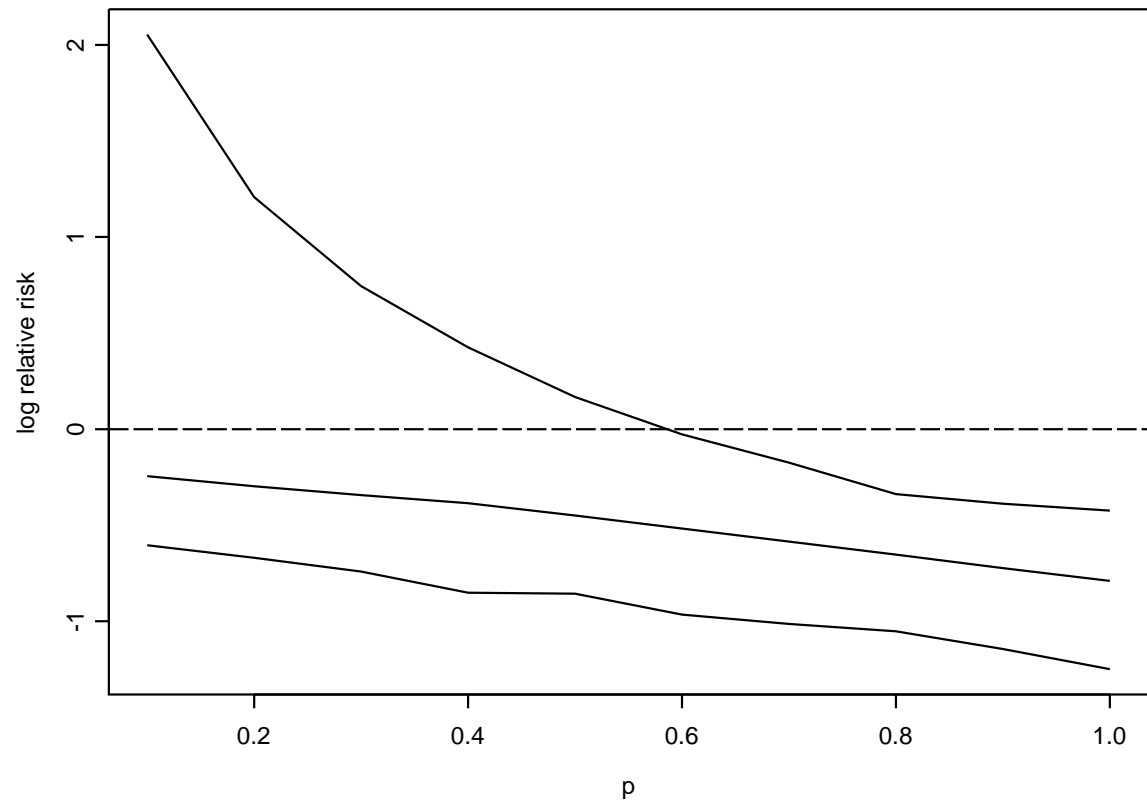
Conclusion: risk may be overestimated, but still significant unless there are more than about 75 "unpublished" studies

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Funnel plot for magnesium studies

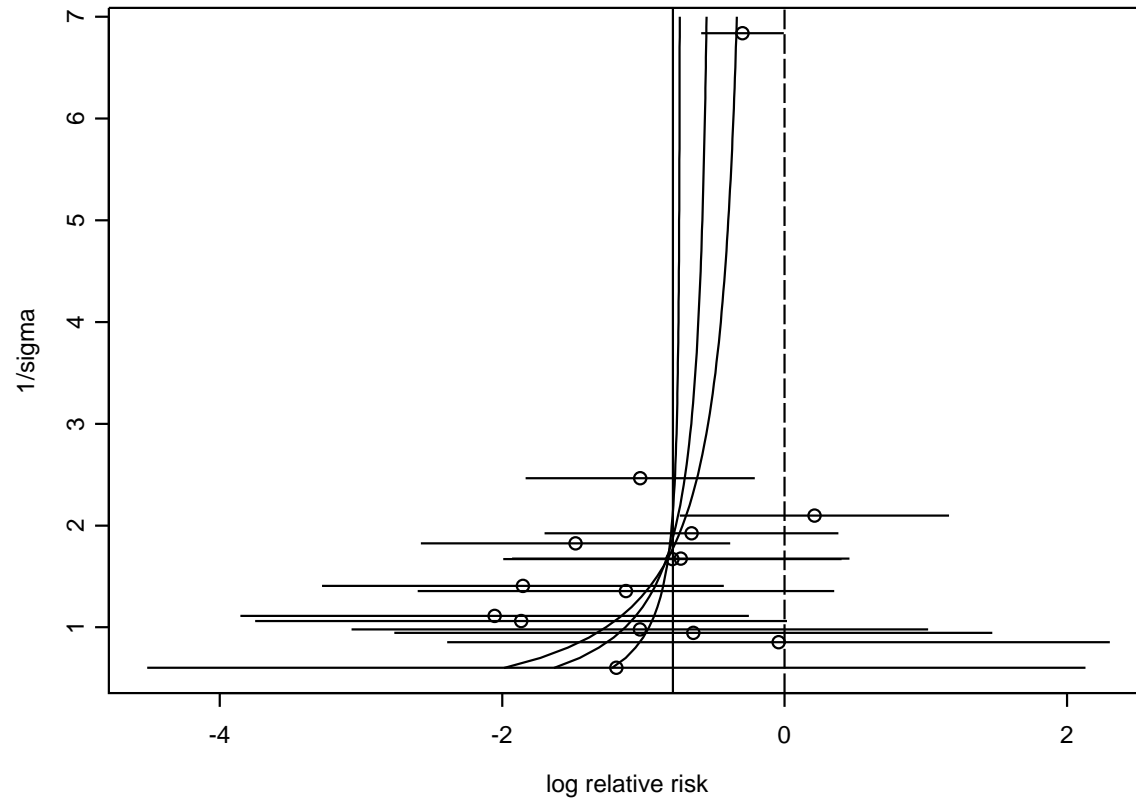


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confidence intervals for magnesium studies

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MLE's of $E(\hat{\theta} | \text{select}, p)$ for $p = 1, 0.9, 0.5, 0.1$

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General comments

- many meta analyses suffer from publication bias, but this is almost always ignored
- publication bias usually means that treatment effects are exaggerated and that significance is exaggerated
- skewness in the funnel plot is a tell-tale sign of publication bias — 40% of all meta analyses show this?
- it is impossible to adjust for publication bias unless we make some unverifiable assumptions about the selection mechanism
- eg the 'selection by significance' approach of Copas (*Applied Stats*, 2013)
- There is no fully convincing way of correcting for publication bias — sensitivity analysis and not bias correction in the usual sense

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- Fixed effects meta analysis may be OK, but usually we use random effects models by adding a random effects term to the likelihood

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References

Andrews DA *et al.* (1990). Does correctional treatment work?: a clinically relevant and psychologically informed meta-analysis. *Criminology*, **28**, 369-429.

Copas JB (1999) What works?: selectivity models and meta-analysis. *J Roy Stats Soc, A*, **162**, 95-109.

Copas JB and Shi JQ (2000) Reanalysis of epidemiological evidence on lung cancer and passive smoking. *British Medical Journal*, **7232**, 417-418.

Copas JB (2013) A likelihood-based sensitivity analysis for publication bias in meta-analysis. *Applied Stats*, bf 62. 47-66.

Hackshaw, AK *et al.* (1997) The accumulated evidence on lung cancer and environmental tobacco smoke. *British Medical Journal*, **315**, 980-8.

ISIS-4 Collaborative Group (1995) A randomized factorial trial assessing

early oral captopril, oral mononitrate and intravenous magnesium sulphate in 58,050 patients with suspected myocardial infarction.

Lancet, **345**, 669-685.

Thompson, S and Pocock, SJ (1991) Can meta analysis be trusted?

Brit Med J, **303**, 1499-1503.

Yusuf S, Koon T, Woods K (1993) Intravenous magnesium in acute myocardial infarction: an effective, safe, simple and inexpensive intervention. *Circulation*, **87**, 2043-2046.