

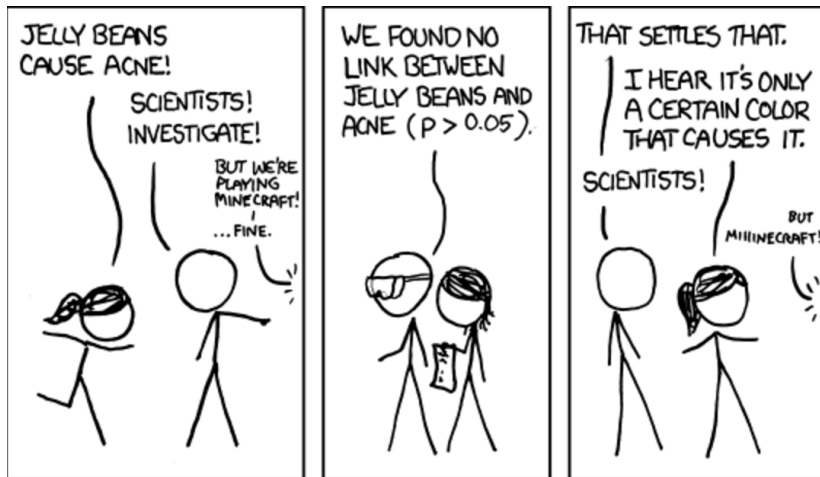
Lecture 7: Publication Bias

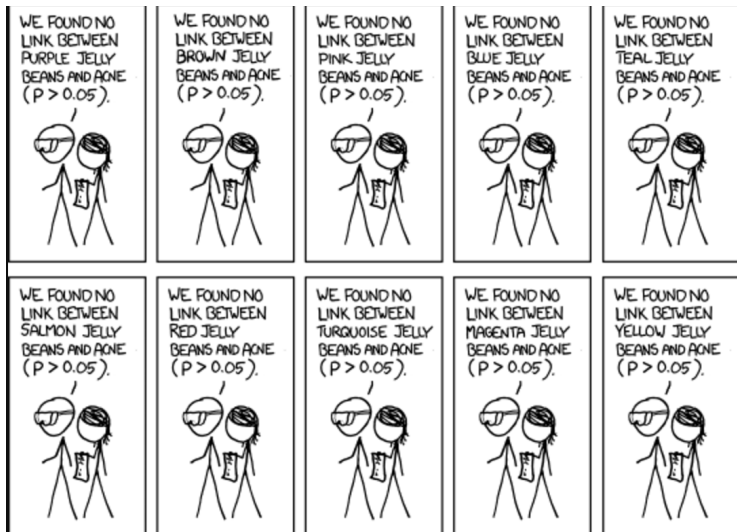
Tomas Havranek

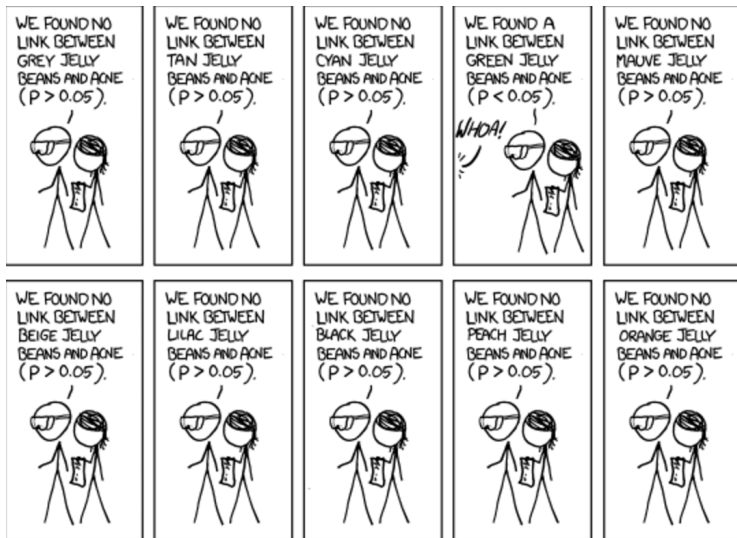
Charles University, CEPR, METRICS

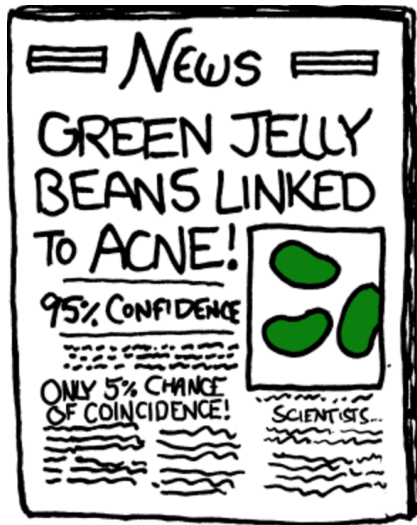
Research Synthesis in Economics and Finance,
University of Canterbury

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- **Background:**

- Paxil (paroxetine): Antidepressant approved for adolescents, developed by GlaxoSmithKline (GSK).
- Study 329 (1998-2001): Clinical trial funded by GSK to test Paxil's safety and efficacy in adolescents.

- **What Happened?**

- Published study claimed Paxil was *"well-tolerated and effective."*
- Internal documents later revealed:
 - Negative results (e.g., increased suicidal ideation) were downplayed.
 - Positive outcomes were selectively reported.

- **Discovery of Bias:**

- Legal proceedings in 2012 forced disclosure of internal data.
- Independent reanalysis of Study 329 showed:
 - No significant benefit over placebo.
 - Serious safety concerns (e.g., suicidal ideation).

- **Key Lessons:**

- Publication bias can harm patients and public trust.
- Highlights the need for:
 - Transparency in research data.
 - Independent replication and reanalysis.

No bias, no correlation (?)

Researchers estimate

$$Earnings_j = \gamma Beauty_j + w_j.$$

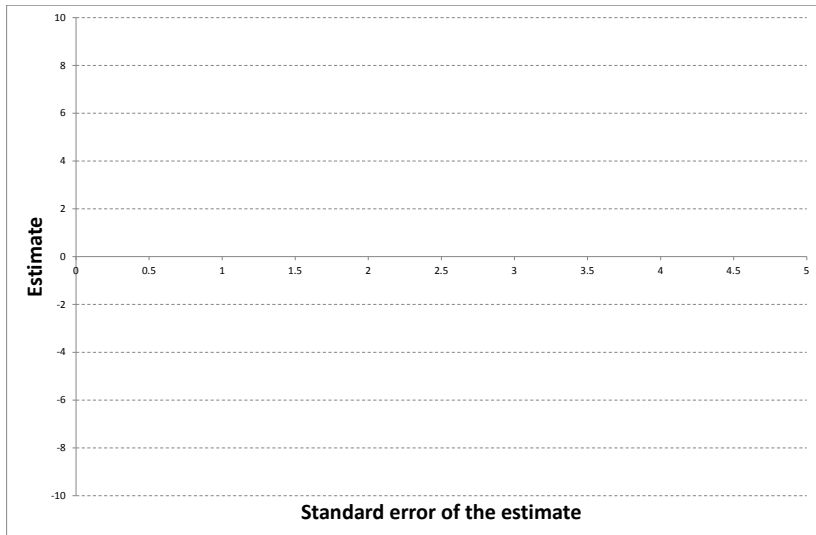
They assume that $\hat{\gamma}/SE_{\hat{\gamma}}$ has a t -distribution.

→ **estimates should not be correlated with standard errors**
(Card & Krueger, 1995)

Types of publication bias

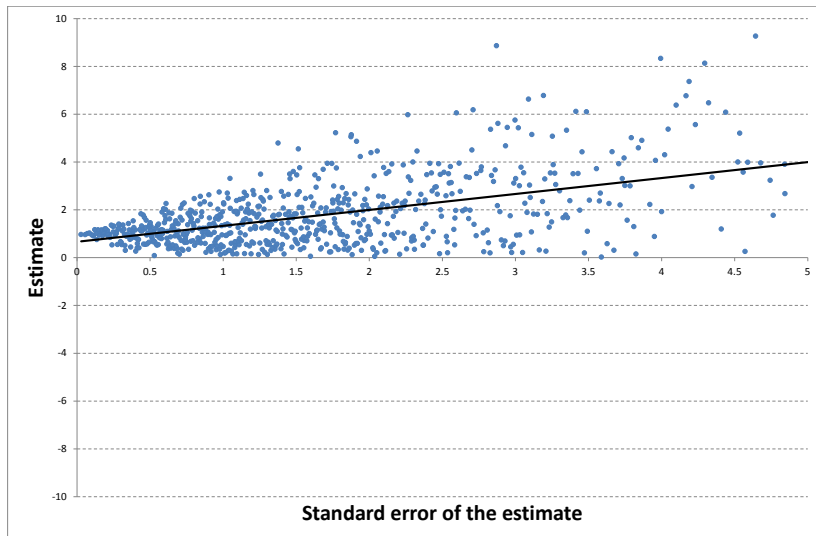
Type I: Selection for the “correct sign.”

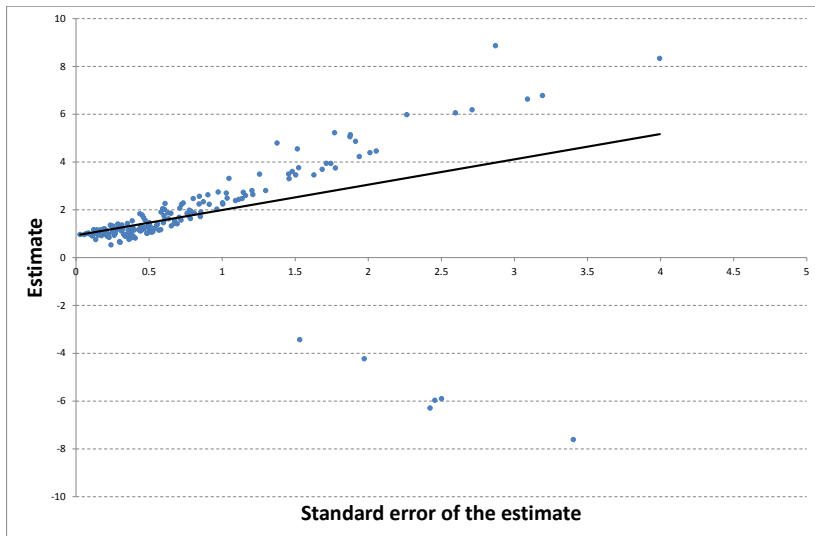
Type II: Selection for statistical significance.

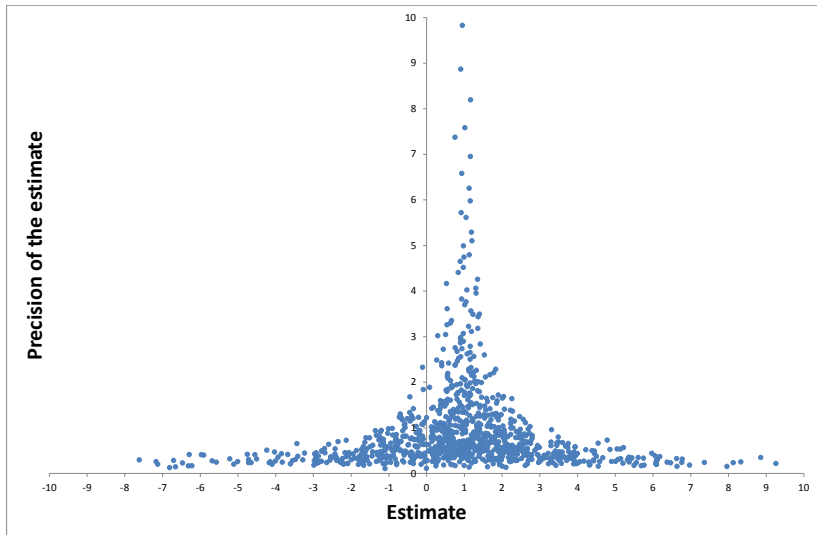


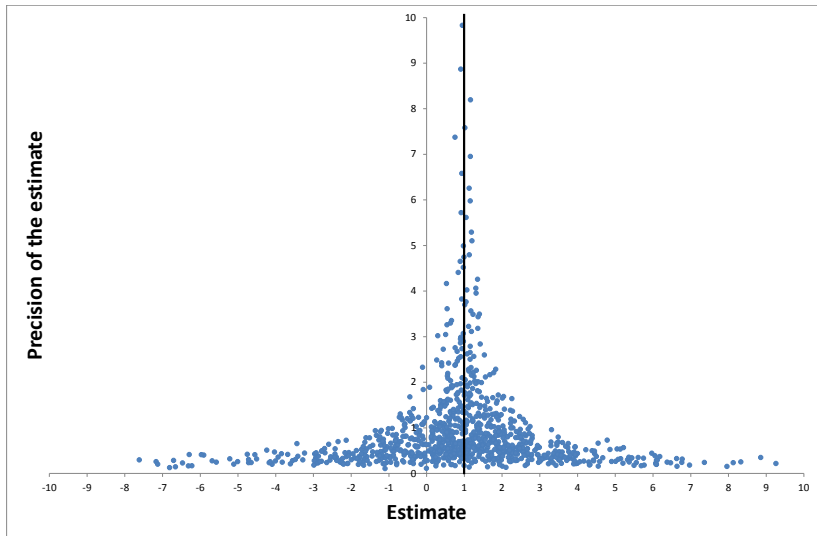


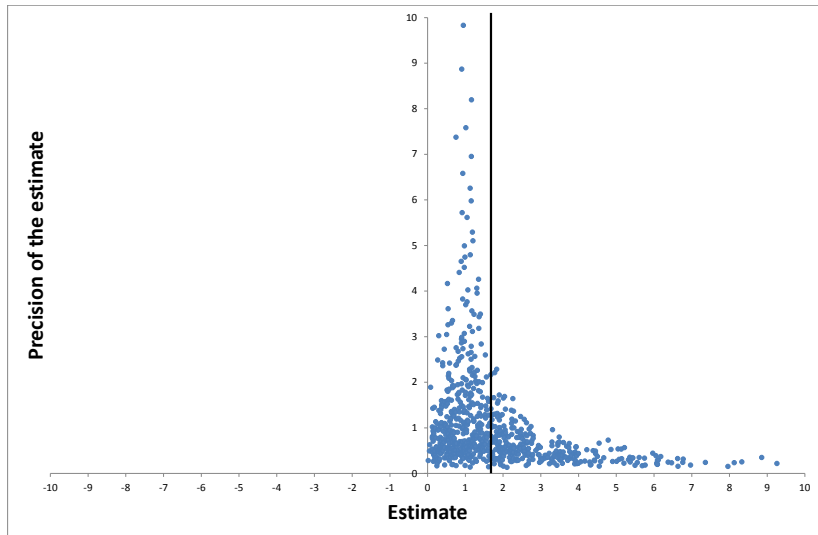


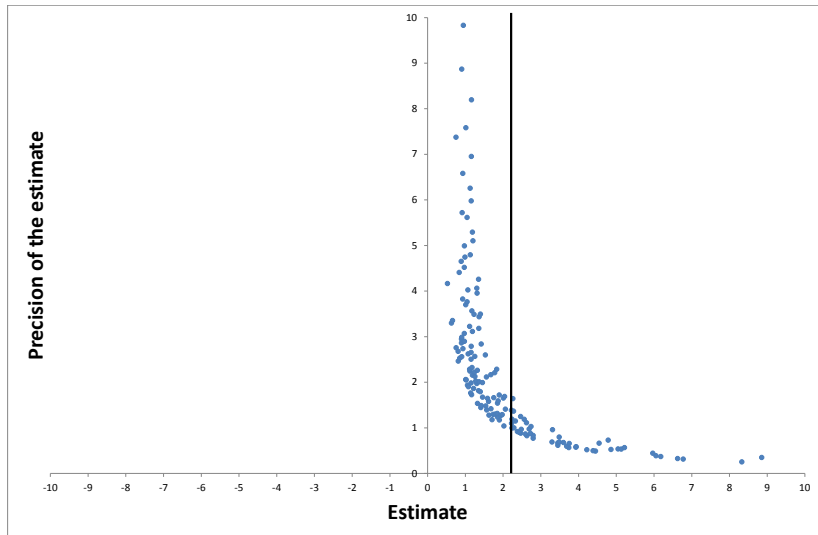












Funnel asymmetry test

In the absence of publication bias the funnel should be symmetrical:

$$\text{estimate}_i = \underbrace{\beta}_{\text{true effect}} + \underbrace{\beta_0 SE_i}_{\text{publication bias}} + \mu_i.$$

The equation is heteroscedastic. Weighted least squares yield

$$t_i = \beta_0 + \beta \left(\frac{1}{SE_i} \right) + \vartheta_i.$$

This is call the funnel asymmetry test—precision effect test (FAT-PET).

Precision effect test with standard error (PEESE):

$$\text{estimate}_i = \underbrace{\beta}_{\text{true effect}} + \underbrace{\beta_0 SE_i^2}_{\text{publication bias}} + \mu_i.$$

The equation is heteroscedastic. Weighted least squares yield

$$t_i = \beta_0 SE_i + \beta \left(\frac{1}{SE_i} \right) + \vartheta_i.$$

WAAP (Ioannidis et al., 2017)

Weighted **A**verage of **A**dequately **P**owered (WAAP) estimates.

- Focus on studies with adequate statistical power to reduce the impact of selective reporting.
- Combines results only from studies with at least 80% power.

Issues:

- Simple and robust.
- Need to estimate the true effect first and then compute power.

Stem-based correction technique (Furukawa, 2019)

Stem-based method:

- A non-parametric estimator that exploits the variance-bias trade-off.
- Focuses on the most precise estimates, referred to as the “stem” of the funnel plot.

Key features:

- Robust to various publication selection processes.
- Does not rely on specific distributional assumptions.

Application:

- Filters studies with the highest precision (lowest standard errors) to estimate the true effect.
- Reduces influence from noisier, potentially biased studies in the broader funnel plot.

Endogenous kink model (Bom & Rachinger, 2019)

Overview:

- Fits a piecewise linear model with a kink determined by the data.
- Identifies where publication selection is less likely to occur.

Key features:

- Endogenously estimates the cutoff based on standard errors.
- Often reduces bias and improves efficiency compared to traditional methods.

Application:

- Highlights genuine effects by focusing on less selective segments of the data.
- Effective in meta-analyses with substantial heterogeneity in study precision.

Selection model (Andrews & Kasy, 2019)

Overview:

- Models the selection process explicitly to address publication bias.
- Assumes researchers are more likely to publish statistically significant results.

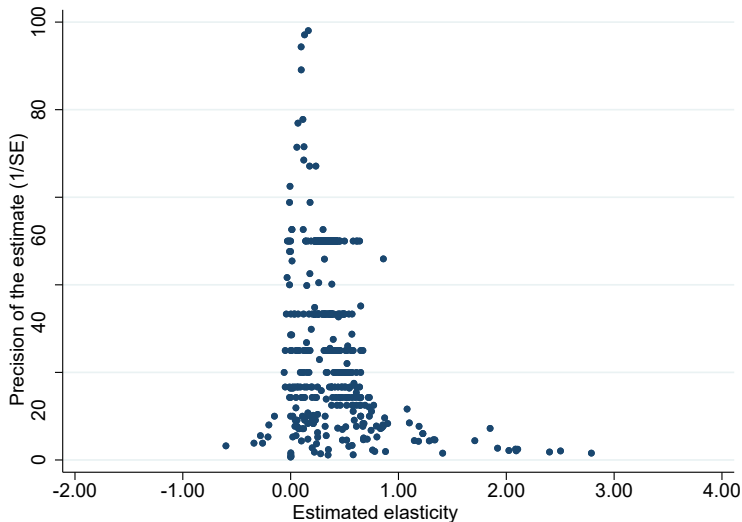
Key features:

- Uses the likelihood of publication as a function of statistical significance.
- Adjusts estimates to reflect the missing studies.

Comparison to funnel-based models:

- Selection models are more rigorous but require stronger assumptions about the publication process.
- Funnel methods are simpler and more robust to model misspecification.

Example: labor supply elasticity



Results vary: RoBMA?

Panel A: Linear tests

	OLS	FE	Precision	Study	IV
Standard error (<i>publication bias</i>)	1.595*** (0.262) [0.93, 2.20]	0.788*** (0.283) -	2.222*** (0.484) [1.23, 3.31]	2.106*** (0.258) [1.53, 2.56]	2.706 (2.103) -
Constant (<i>mean beyond bias</i>)	0.301*** (0.044) [0.12, 0.42]	0.370*** (0.026) -	0.247*** (0.065) [0.11, 0.31]	0.252*** (0.109) [0.13, 0.38]	0.130* (0.066) -
Observations	723	723	723	723	522
Studies	36	36	36	36	23

Panel B: Nonlinear tests

	Ioannidis <i>et al.</i> (2017)	Andrews & Kasy (2019)	Bom & Rachinger (2019)	Furukawa (2021)	van Aert & van Assen (2021)
Effect beyond bias	0.260*** (0.041)	0.356*** (0.061)	0.207** (0.094)	0.187* (0.112)	0.374*** (0.020)
Observations	723	723	723	723	723
Studies	36	36	36	36	36

Bias caused by preference for positive estimates

