

# Meta-Research

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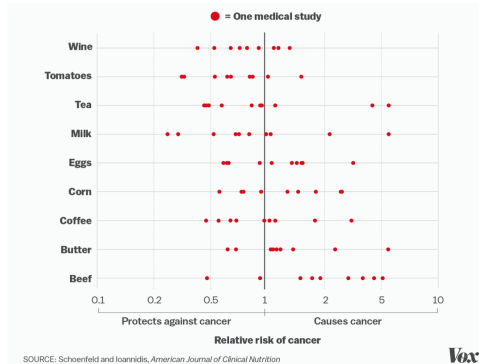
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# Road Map

- Why
- How to
- Fun things to do

# Why Are Meta-Analyses Important?

- Any one given study is not all that informative.



# Why Are Meta-Analyses Important?

- Research might be biased.

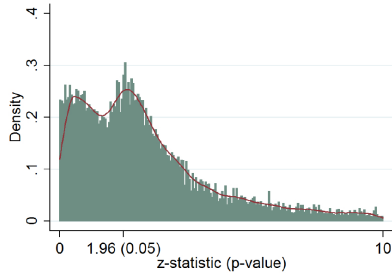


Figure: Brodeur *et al.*, 2016

# Why Are Meta-Analyses Important?

- Studies are often underpowered.

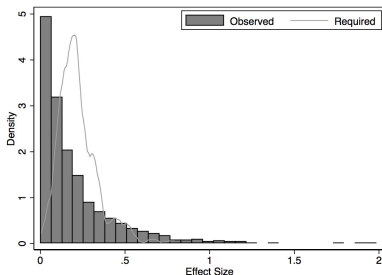


Figure: Vivalt, 2017

# Why Are Meta-Analyses Important?

- Studies are partially informative about *each other*.

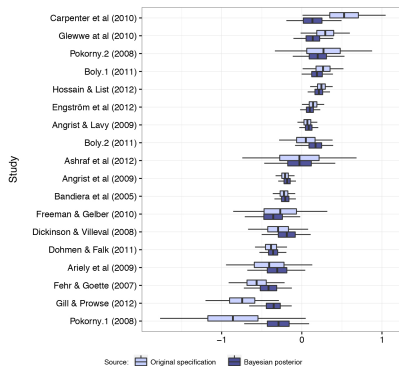


Figure: Fischer *et al.*, 2017

# How To Do A Meta-Analysis

- Start by making a lot of decisions about what you will do.
  - Topic
  - Search
  - Screening
  - Data extraction
  - Analysis

## Search and Screening: PICOS

- Population
- Intervention
- Comparison
- Outcome
- Study design

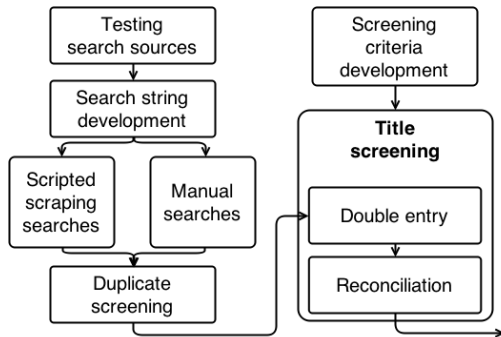


## Search and Screening: Quality Measures

- Was the study a randomized controlled trial?
- Was it blinded?
- Did it report attrition?
- Were the hypotheses pre-specified?
- Was it published in a good journal? (Bias?)
- *etc.*

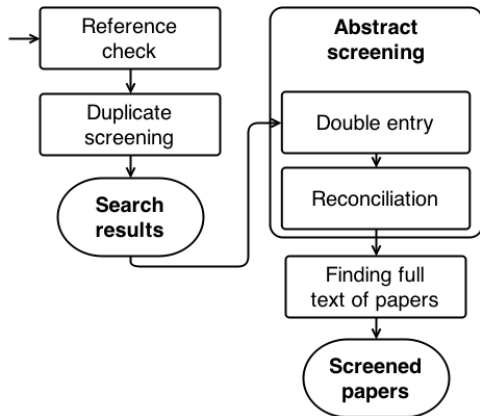
# Sample Process Diagram

Figure: Search and Screening, Part 1



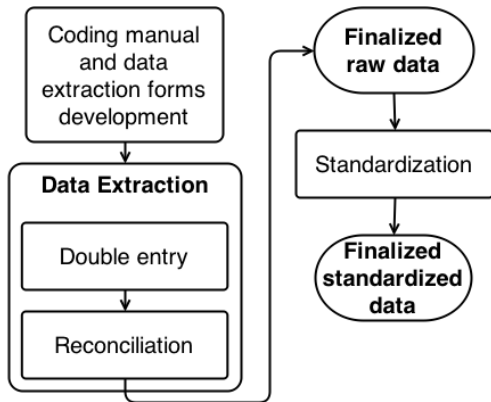
# Sample Process Diagram

Figure: Search and Screening, Part 2



# Sample Process Diagram

Figure: Data Extraction



## Conversion Between Data Types

- Studies report results in all sorts of different ways
- Some of them can be converted to each other, e.g. binary 2x2 tables can become risk ratios or odds ratios
- Refer to Cochrane handbook for different combinations

# Analysis

Some references that might be useful:

- Borenstein *et al.* (2009). Introduction to Meta-Analysis
- Gelman *et al.* (2013). Bayesian Data Analysis
- Higgins and Green, *eds.* (2011).  
Cochrane Handbook for Systematic Reviews of Interventions

Coding references:

<http://stats.idre.ucla.edu/stata/seminars/introduction-to-meta-analysis-in-stata/>

[http://bookdown.org/MathiasHarrer/Doing\\_Meta\\_Analysis\\_in\\_R/](http://bookdown.org/MathiasHarrer/Doing_Meta_Analysis_in_R/)

# Analysis

- Fixed effect: there is one true effect
- Random-effects: true effects may vary by study
- Mixed model: true effects may vary by study, and we can explain some of that

# Fixed Effect Model

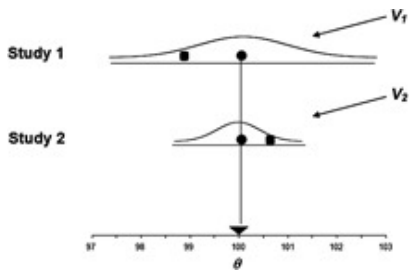


Figure: Borenstein *et al.*, 2009



## Example Fixed Effect Model

$$Y_i = \theta + \epsilon_i$$

$$\epsilon_i \sim N(0, \sigma_i^2)$$

$Y_i$  is the estimate of the effect in study  $i$

$\theta$  is the true effect

$\epsilon_i$  is the error, normally distributed with some sampling variance  $\sigma_i^2$

## Fixed Effect Model

- Meta-analysis is a kind of weighted average
- Need to pick a weighting rule

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$$w_i = \frac{1}{v_i}, M = \frac{\sum_i w_i Y_i}{\sum_i w_i}, v_M = \frac{1}{\sum_i w_i}$$

# Random-Effects Model

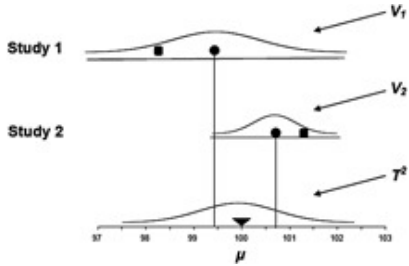


Figure: Borenstein *et al.*, 2009

## Example Random-Effects Model

$$Y_i = \theta_i + \epsilon_i$$

$$\epsilon_i \sim N(0, \sigma_i^2)$$

$$\theta_i \sim N(\mu, \tau^2)$$

$Y_i$  is the estimate of the effect in study  $i$

$\theta_i$  is the true effect in study  $i$

$\epsilon_i$  is the error, normally distributed with some sampling variance  $\sigma_i^2$

$\mu$  is the grand mean

$\tau^2$  is the inter-study variance

## Random-Effects Model

- Now the variance has two components:  $v_i = \sigma_i^2 + \tau^2$
- Most common way of estimating  $\tau^2$ : DerSimonian-Laird estimator
- Better: Maximum likelihood, Empirical Bayes, Sidik-Jonkman (Sidik and Jonkman, 2007), Full Bayes
- Sensitivity analyses recommended, especially for small number of studies

## Example Mixed Model

$$Y_i = \alpha + X_i\beta + \zeta_i + \epsilon_i$$

$$\theta_i = \alpha + X_i\beta + \zeta_i$$

$$\epsilon_i \sim N(0, \sigma_i^2)$$

$$\zeta_i \sim N(0, \tau^2)$$

$Y_i$  is the estimate of the effect in study  $i$

$\theta_i$  is the true effect in study  $i$  and is comprised of some component that can be explained ( $X_i\beta$ ) and some component that cannot ( $\zeta_i$ )

$\epsilon_i$  is the error, normally distributed with some sampling variance  $\sigma_i^2$

$\tau^2$  is the inter-study variance



# Estimating in Stata

*metan:*

- `metan TE lowerci upperci, fixedi`
- `metan TE lowerci upperci, randomi`
- `metan tdeath tnodeath cdeath cnodeath`
- `metan tsample tmean tsd csample cmean csd`
- `metan logor selogor, eform`

Mantel-Haenszel method will be used if “fixed” or “random” is specified rather than “fixedi” or “randomi”. “fixed” is the default

# Estimating in R

*meta* package:

- metagen for treatment effects:

```
m.se <- metagen(TE, se, data=mydata, comb.fixed=TRUE,  
comb.random=FALSE, sm="SMD")
```

- metacont for raw data:

```
m.raw <- metacont(NT, MT, ST, NC, MC, SC,  
data=mydata, comb.fixed=TRUE, comb.random=FALSE,  
sm="SMD")
```

Option `method.tau` governs how  $\tau^2$  is estimated. E.g.

`method.tau="DL"`, `method.tau="ML"`, `method.tau="EB"`, etc.

## Estimating in R

*meta* package [cont]:

- `metabin` for raw binary data:  
`m.bin <- metabin(Et, Nt, Ec, Nc, data=mydata,  
comb.fixed=FALSE, comb.random=TRUE, sm="RR")` (risk ratios)

*metafor* package:

- `m <- rma(yi = TE, sei = se, method = "FE", data = mydata)`

# Estimating in R

Bayesian hierarchical models:

- Priors over  $\mu$  and  $\tau$ , update using data, draw posterior distribution
- A long, hard way: coding up a Bayesian hierarchical model
- The short way: Rachael's new package using Stan
- Check: do you have the right functional form?
- Check: how sensitive are your results to the priors you selected?

## Fun Things to Do

- Model and characterize heterogeneity
- Re-interpret study results in light of other studies
- Estimate biases
- Other ways to leverage Bayesian model

# Characterizing Heterogeneity

- Some possible measures:
  - $\text{var}(Y_i)$  or  $\tau^2$
  - coefficient of variation
  - $I^2 = \frac{\tau^2}{\tau^2 + \sigma^2}$
- Each possible measure of heterogeneity is flawed

# Studies are partially informative about *each other*

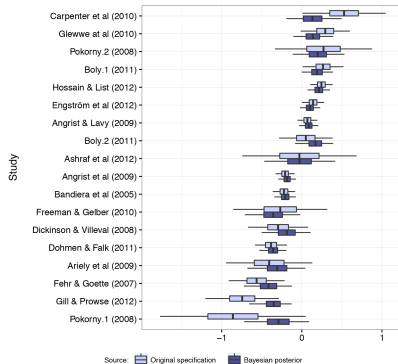


Figure: Fischer *et al.*, 2017

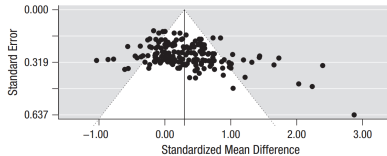
# Model Heterogeneity

- Generally not enough data for meta-regression....



# Estimate Biases

- Funnel plots:



**Fig. 2.** Funnel plot of the meta-analysis of published studies. Each plotted point represents the standard error and standardized mean difference between bilinguals and monolinguals for a single study. The white triangle represents the region where 95% of the data points would lie in the absence of a publication bias. The vertical line represents the average standardized mean difference of 0.30 found in the meta-analysis.

Figure: de Bruin *et al.*, 2015

## Examples in Stata

*metafunnel:*

- Plots funnel plots

*metabias:*

- Various tests for funnel-plot asymmetry

*metatrim:*

- “Trim and fill” method of adjusting meta-analysis to account for biases

These methods require heroic assumptions....

## Examples in R

*forest*: part of the *meta* package

- `forest(m.raw, raw)` after `m.raw`

Why might a funnel plot not necessarily be a good way to gauge publication bias?

# Estimate Biases

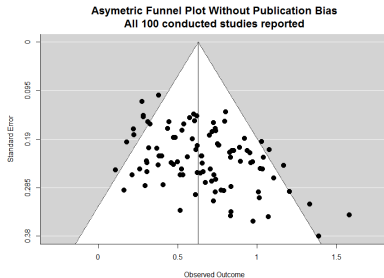


Figure: Simonsohn, Data Colada blog, 2017

# Estimate Biases

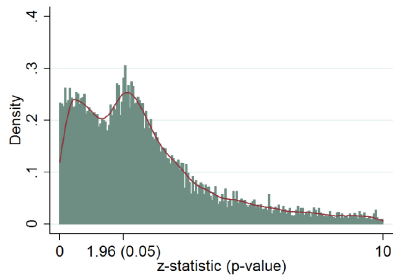


Figure: Brodeur *et al.*, 2016

## Estimate Biases

- When can you use a caliper test?
- How should you use a caliper test?

## Estimate Biases

- False positive and false negative report probability (FPRP/FNRP), “modified” FPRP/FNRP, Bayesian False Discovery Probability, or the conditional error probability (Wacholder *et al.*, 2004; Ioannidis, 2005; Wakefield, 2007; Wakefield, 2008; Lucke, 2009)
  - Need some kind of prior
  - Intuition: results that seem unlikely are either really novel and valuable or false positives

## Estimate Biases

- In econ: Ioannidis, Stanley and Doucouliagos analyze power (2017). Use fixed effect meta-analysis result as true effect to try to avoid bias (but heterogeneity in treatment effects?)
- Type S and Type M errors (Gelman and Carlin, 2014)



## Learn about Learning

- Can estimate how much we learn from an impact evaluation, e.g. given estimates of  $\tau$ ,  $\sigma$
- Value of information

# Cautions

- Like any study, meta-analyses are only as good as the methods used (and studies included)
- Meta-analyses can be biased
- Same best practices apply:
  - Pre-specify (register with Cochrane)
  - Document what is done in sufficient detail that it could be replicated
  - Version data