

A re-appraisal of fixed effect(s) meta-analysis

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tl;dr

- Fixed-effect**S** meta-analysis answers a sensible question **re**-gardless of heterogeneity
- Other questions can also be sensible
- Fixed-effectS methods extend to useful measures of heterogeneity and meta-regression, small-sample corrections and Bayesian inference
- Knowing what question you're answering helps determine whether it's sensible

http://tinyurl.com/fixef

has these slides and more.

Generic example

Meta-analyzing trials* to estimate some overall effect;

		Zinc			Placebo	1	Zinc Pla	acebo		
Study	N	Mean	SD	N	Mean	SD		etter	Mea	n Difference, 95% CI
Douglas 1987	33	12.1	9.8	30	7.7	9.8	<u> </u>	<u>-</u>	—	4.40 [-0.45 , 9.25]
Petrus 1998	52	4.4	1.4	49	5.1	2.8	⊢		-	-0.70 [-1.57 , 0.17]
Prasad 2000	25	4.5	1.6	23	8.1	1.8	⊢∎→		_	3.60 [-4.57 , -2.63]
Prasad 2008	25	4	1.04	25	7.12	1.26	H ⊞ H		_	3.12 [-3.76 , -2.48]
Turner 2000b	68	6.89	3.35	71	7.55	3.96	⊢ ■:		-	-0.66 [-1.88 , 0.56]
Turner 2000c	72	7.9	4.25	71	7.55	3.96	⊢ ■			0.35 [-1.00 , 1.70]
Fixed effects, preci	sion-	-weighte	d averaç	ge (P\	NA) esti	mator	•		_	2.04 [-2.45 , -1.64]
Random effects, DerSimonian-Laird estimator						-			–1.21 [–2.69 , 0.28]	
							İ			
							-5.00 0.00	5.00	10.00	
							Mean Diffe	erence (days)		

• Generic Q: Which average? Why?

^{*} from Zinc for the Common Cold (2011) – Cochrane review of zinc acetate lozenges for reducing duration of cold symptoms (days)

Fixed effect (singular)

... based on the assumption that the results of each trial represents a statistical fluctuation around some common effect

Steve Goodman



Controlled Clinical Trials, 1989

In the *fixed effect* model for k studies we assume

and noise in σ_i is negligible. Obvious (and optimal) estimate is the *inverse variance-weighted* or *precision-weighted* average:

$$\widehat{\beta}_F = \sum_{i=1}^k \frac{\frac{1}{\sigma_i^2}}{\sum_{i=1}^k \frac{1}{\sigma_i^2}} \widehat{\beta}_i, \quad \text{with } \text{Var}[\widehat{\beta}_F] = \frac{1}{\sum_{i=1}^k \frac{1}{\sigma_i^2}}.$$

Fixed effectS (plural)

But assuming β_i exactly homogeneous is **silly** in (most) practice, as effects are **not identical**

- Environments & adherence differ (and much else)
- In my applied work, genetic ancestry differs

But but but note that if

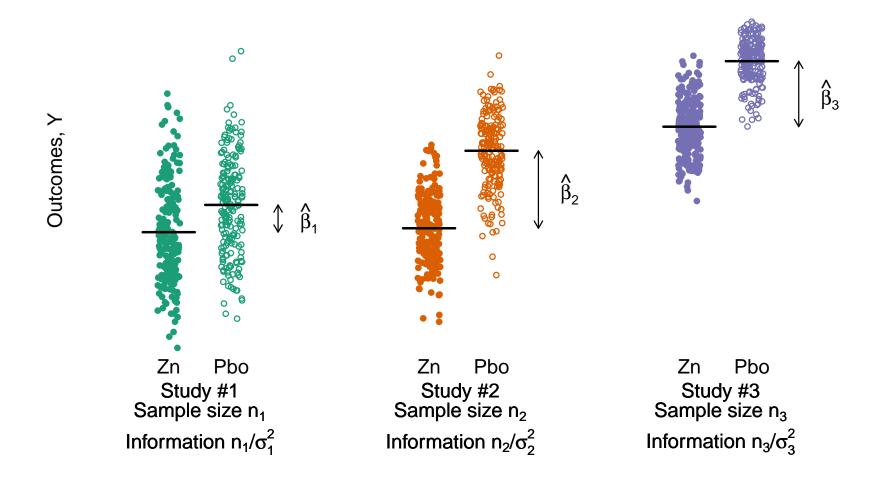
$$\widehat{eta}_i \sim N(eta_i, \sigma_i^2), \ 1 \leq i \leq k,$$
 by the CLT,

and noise in σ_i is negligible, then can still define

$$\widehat{\beta}_F = \sum_{i=1}^k \frac{\frac{1}{\sigma_i^2}}{\sum_{i=1}^k \frac{1}{\sigma_i^2}} \widehat{\beta}_i, \quad \text{with } \text{Var}[\widehat{\beta}_F] = \frac{1}{\sum_{i=1}^k \frac{1}{\sigma_i^2}}.$$

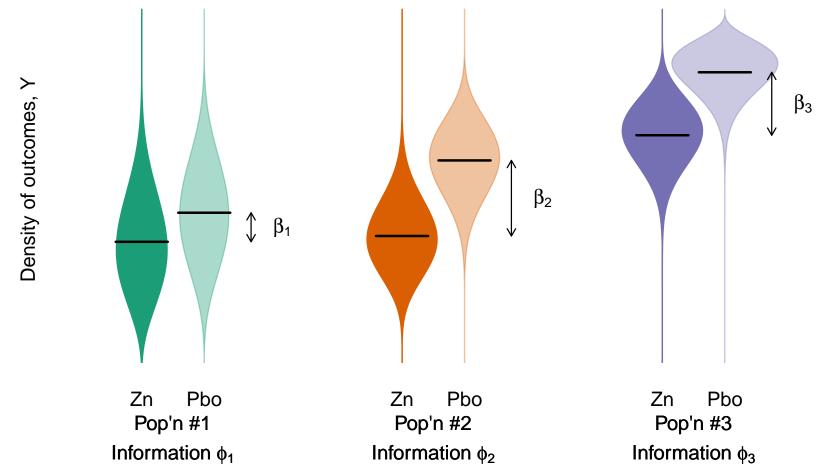
The fixed effects estimate provides valid statistical inference on an 'average' of the β_i , regardless of their homogeneity/heterogeneity

First, consider possible data from three studies;



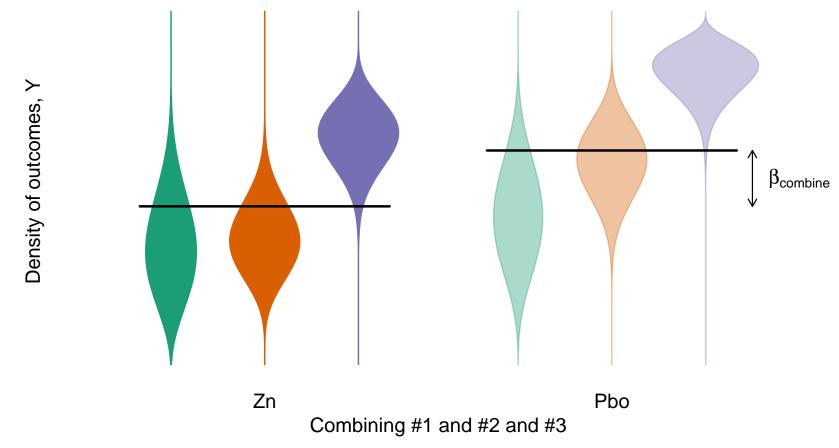
Each $n_i = 200$ here. We assume all σ_i^2 known, for simplicity.

Population parameters those 3 studies are estimating;



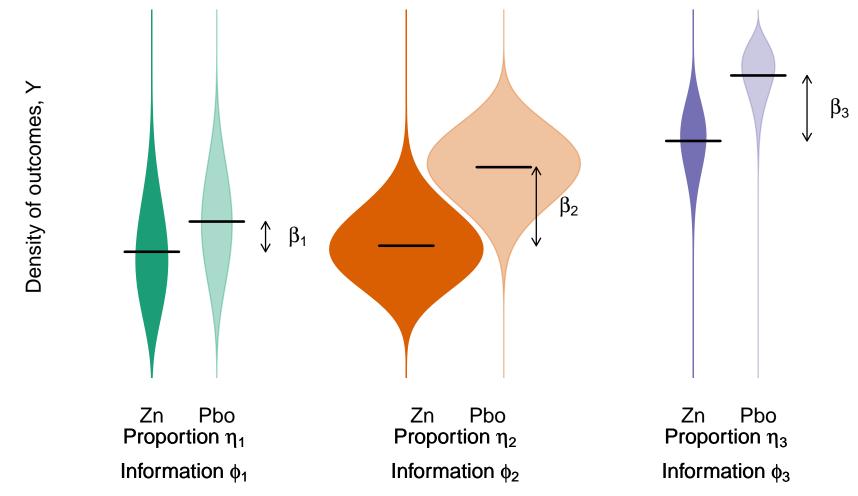
Parameters are differences in means (β_i) and information per observation (ϕ_i) .

One overall population we might learn about;



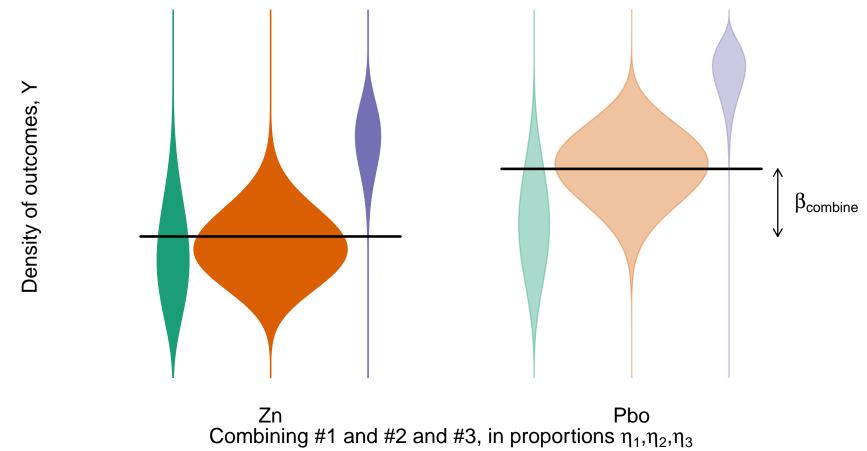
 β_{combine} is the mean difference (zinc vs placebo) with each subpopulation represented equally, i.e. weighted 1/1/1.

Another overall population we might learn about;



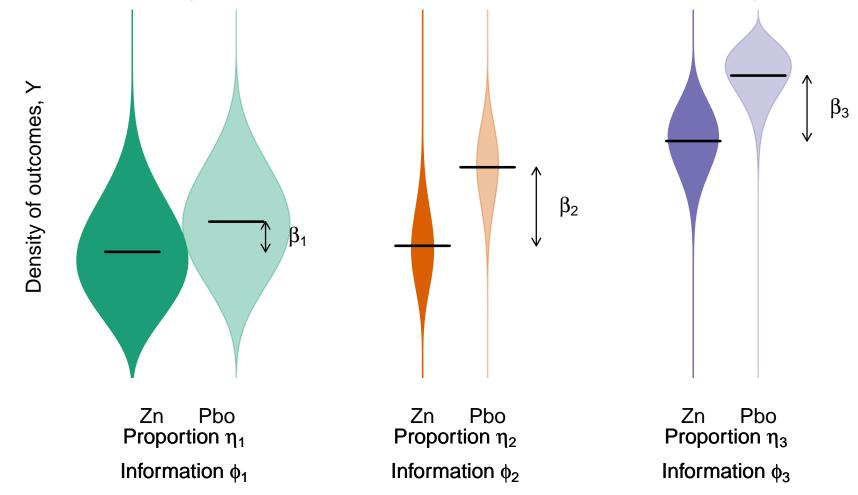
Weights here are 2/7/1, not 1/1/1 as before.

Another overall population we might learn about;



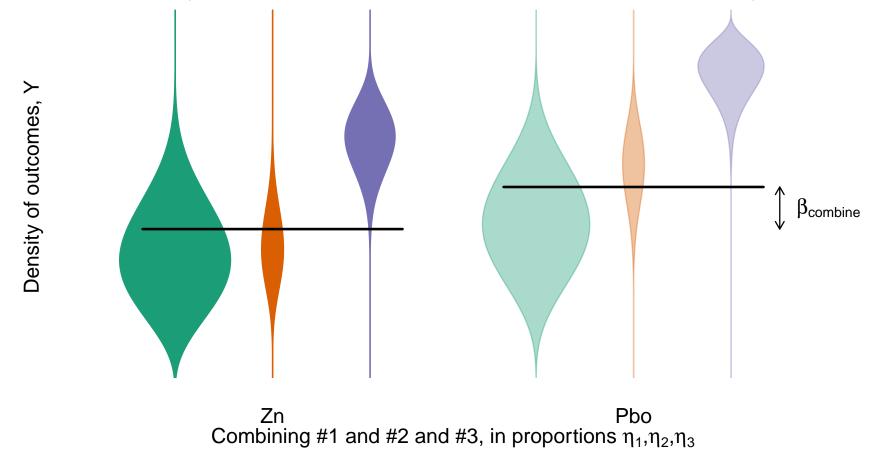
Still an average effect, but closer to β_2 than before.

And another; (obviously, there are unlimited possibilities)



Weights here are 7/1/2.

And another; (obviously, there are unlimited possibilities)



Weights here are 7/1/2 – smaller average effect, closer to β_1

Fixed effectS: general case

Upweighting studies which are larger **and** more informative about their corresponding β_i , we can estimate population parameter

$$\beta_F = \frac{\sum_{i=1}^k \eta_i \phi_i \beta_i}{\sum_{i=1}^k \eta_i \phi_i} = \frac{\sum_{i=1}^k \frac{1}{\sigma_i^2} \beta_i}{\sum_{i=1}^k \frac{1}{\sigma_i^2}},$$

by
$$\hat{\beta}_F = \frac{\sum_{i=1}^k \frac{1}{\sigma_i^2} \hat{\beta}_i}{\sum_{i=1}^k \frac{1}{\sigma_i^2}}$$
, with $\operatorname{Var}[\hat{\beta}_F] = \frac{1}{\sum_{i=1}^k \frac{1}{\sigma_i^2}}$.

- $\hat{\beta}_F$ is the precision-weighted average, a.k.a. inverse-variance weighted average a.k.a. fixed effect**S** estimator **note the plural!**
- $\widehat{\beta}_F$ is consistent for average effect β_F under regime where all $n_i \to \infty$ in fixed proportion
- Homogeneity, or tests for heterogeneity are **not required** to use $\hat{\beta}_F$ and its inference

Fixed effectS: general case

Homogeneity, or tests for heterogeneity are **not required** to use $\hat{\beta}_F$ and its inference

Users who have **only** seen the fixed effect (singular) motivation tend to view it as the **only** reason for ever using $\hat{\beta}_F$.

This is **specious**...



If I see you buy eggs, should I think the **only** reason is that you're making an omelet?

Fixed effectS: general case

The basic ideas here are **not new**:

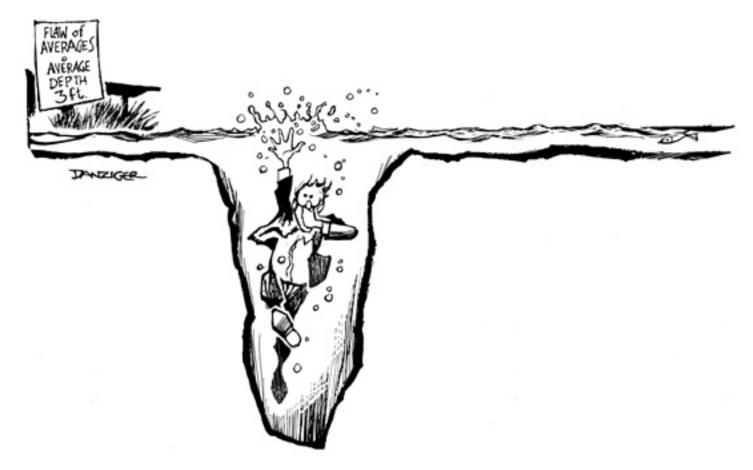
- Same average-effect argument already supports e.g. the Mantel-Haenszel estimate
- Fixed effect**S** arguments presented by e.g. Peto (1987), Fleiss (1993) and Hedges (various, e.g. Handbook of Research Synthesis), all noting the validity of β_F and inference using $\widehat{\beta}_F$ under heterogeneity

Also

- Can still motivate $\widehat{\beta}_F$ when σ_i are estimated, though $Var[\widehat{\beta}_F]$ requires more care (Domínguez-Islas & Rice 2018)
- Can use them in Bayesian work, with exchangeable priors (Domínguez-Islas & Rice, under review) – much less sensitive than default methods

But what about heterogeneity?

We all know the 'flaw of averages';



- ullet Average effect eta_F answers one question
- This does not mean other questions aren't interesting!

But what about heterogeneity?

A weighted variance of effects:

$$\zeta^{2} = \frac{1}{\sum_{i=1}^{k} \eta_{i} \phi_{i}} \sum_{i=1}^{k} \eta_{i} \phi_{i} (\beta_{i} - \beta_{F})^{2}.$$

And an empirical estimate of it:

$$\hat{\zeta}^2 = \frac{\sum_{i=1}^k \sigma_i^{-2} (\hat{\beta}_i - \hat{\beta}_F)^2 - (k-1)}{\sum_{i=1}^k \sigma_i^{-2}} = \frac{Q - (k-1)}{\sum_{i=1}^k \sigma_i^{-2}}$$

where Q is Cochran's Q and $I^2 = 1 - (k-1)/Q$ (truncated at zero) are standard statistics for assessing homogeneity.

- (Weighted) standard deviation ζ measure on the β scale is easier to interpret than Q or I^2
- Inference on ζ far more stable than mean of (hypothetical) random effects distributions

But what about heterogeneity?

Meta-regression — essentially weighted linear regression of the $\hat{\beta}_i$ on known study-specific covariates x_i — also tells us about differences from zero, beyond the overall effect $\hat{\beta}_F$.

Using extensions of the arguments for $\widehat{\beta}_F$, the standard linear meta-regression 'slope' estimate can be written

$$\widehat{\beta}_{MR} = \frac{\sum_{i=1}^{k} w_i (x_i - \widehat{x}_F)^2 \frac{\widehat{\beta}_i - \widehat{\beta}_F}{x_i - \widehat{x}_F}}{\sum_{i'=1}^{k} w_{i'} (x_{i'} - \widehat{x}_F)^2}, \text{ where } \widehat{x}_F = \frac{\sum_{i=1}^{k} w_i x_i}{\sum_{i'=1}^{k} w_{i'}} \text{ and } w_i = \frac{1}{\sigma_i^2},$$

which with no further assumptions estimates

$$\beta_{MR} = \frac{\sum_{i=1}^{k} \eta_i \phi_i (x_i - x_F)^2 \frac{\beta_i - \beta}{x_i - x_F}}{\sum_{i'=1}^{k} \eta_{i'} \phi_{i'} (x_{i'} - x)^2}, \text{ where } x_F = \frac{\sum_{i=1}^{k} \eta_i \phi_i x_i}{\sum_{i'=1}^{k} \eta_{i'} \phi_{i'}}.$$

- $Var[\hat{\beta}_{MR}]$ also available
- ANOVA/ANCOVA breakdowns of total 'signal:noise' available to accompany ζ^2 and $\widehat{\beta}_{MR}$ analysis

Are you going to stop now?

Summary, under standard conditions;

Name:	Common effect	Fixed effect S
Assumptions:	Effect size $ \text{All } \beta_i = \beta_0 $	$\frac{\textbf{Effect size}}{\beta_i \text{ unrestricted}}$
Plausible?	Rarely	Often!
i idasibic.	ixarciy	Often
\widehat{eta}_F estimates:	Single β_0	Sensible average, eta_F
\widehat{eta}_F estimates: Valid estimate?		
\widehat{eta}_F estimates:	Single β_0	Sensible average, eta_F
\widehat{eta}_F estimates: Valid estimate?	Single β_0 Yes	Sensible average, β_F Yes

^{* ...} if we can ignore uncertainty about the σ_i^2

Acknowledgements

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Clara Dominguez-Islas



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Julian Higgins

Reminder: http://tinyurl.com/fixef for slides & more.