

Heterogeneity in close and conceptual replications

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Context

- ▶ Replication crisis
- ▶ Replication projects
 - ▶ Close replications (Many Labs, Registered Replication Reports)
 - ▶ Keep design, materials, analyses, etc. as close to original as possible
- ▶ Meta-analyses
 - ▶ Conceptual replications
 - ▶ Different designs, materials, participants, analysis, etc.
- ▶ Heterogeneity – how much true effect sizes differ across studies

Why care about heterogeneity?

- ▶ Affects statistical power (McShane & Böckenholt, 2014; Shrout & Rodgers, 2018)
 - ▶ Heterogeneity decreases power
 - ▶ Power calculations should take this into account. But first we need a reliable estimate of heterogeneity, and what may be driving this.
 - ▶ Could this explain low success in 100 close replications from Open Science Collaboration (2015)?
- ▶ Design of practical applications
 - ▶ Heterogeneity can tell us something about the level of certainty around the result of the 'next study'
 - ▶ Successful translation of research into practice depends on consistency of findings

Aims

- ▶ Derive an estimate of heterogeneity in close replication studies
- ▶ Compare this to heterogeneity in a large sample of meta-analyses
- ▶ Investigate some possible causes of heterogeneity

Hypotheses

- ▶ Heterogeneity in close replications expected to be low
- ▶ Heterogeneity in conceptual replications
 - ▶ Higher in social than cognitive psychology
 - ▶ Higher replication success for cognitive than social psychology (Open Science Collaboration, 2015)
 - ▶ Higher in social than organisational psychology
 - ▶ Higher correlation between lab and field studies in organisational than social psychology (Mitchell, et al., 2012)

Methods

- ▶ 40 close replication studies (Many Labs and Registered Replication Reports)
- ▶ 147 meta-analyses sampled (cognitive, organisational, social psychology)
 - ▶ Cohen's d as measure of ES
 - ▶ τ as measure that quantifies heterogeneity
 - ▶ Generally assumed that population ES for a given phenomenon follow a normal distribution
 - ▶ τ is their standard deviation
- ▶ d and τ calculated by re-doing all meta-analyses

Methods – Two approaches to moderators

- ▶ 25 meta-analyses with $k \geq 60$, and with sufficient information to re-examine moderator analyses
 - ▶ Excluded 'broad' subsets (e.g. adults, children, mixed – mixed sample excluded)
- ▶ For meta-analysis as a whole, rated broadness/narrowness of inclusion criteria for studies on a 5-point scale
 - ▶ E.g. is the question addressed narrow/broad
 - ▶ Does manipulation of IV/DV follow standard protocol

How do meta-analyses address heterogeneity?

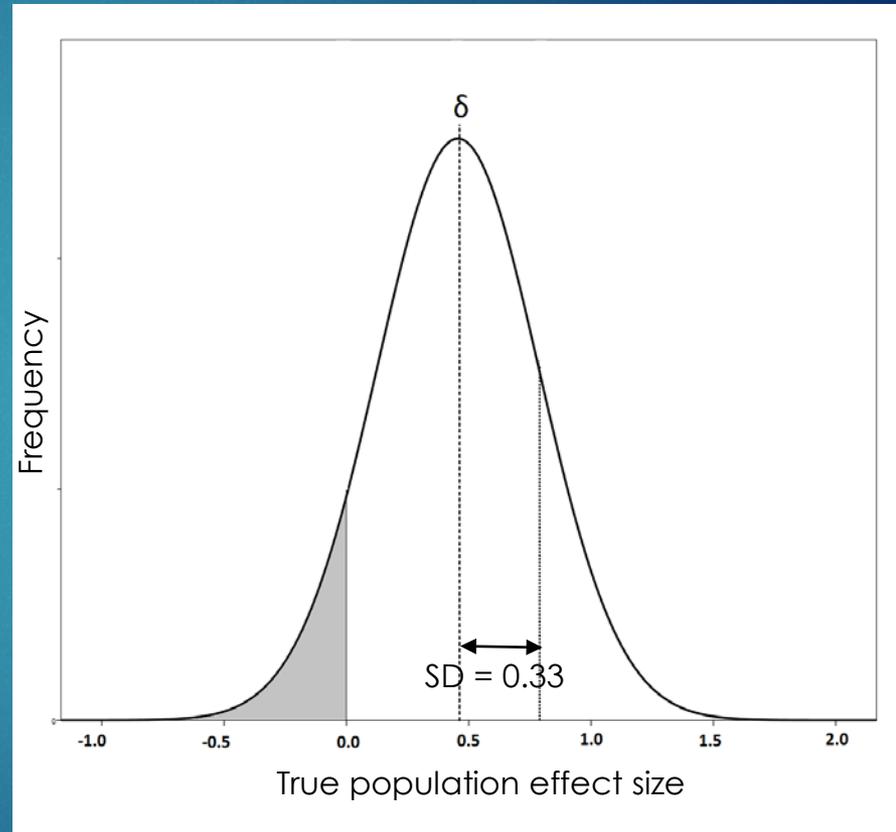
- ▶ Out of 147 meta-analyses...
 - ▶ 54% reported a measure of heterogeneity
 - ▶ Heterogeneity **quantified** in only 38 cases (26%)
- ▶ Post-PRISMA? (Preferred Reporting Items for Systematic Reviews and Meta-Analyses, 2009)
 - ▶ Heterogeneity only reported in 60% of cases

Findings in our sample

- ▶ Average τ was low ($M = 0.08$) in **close** replications
- ▶ Average τ was much higher ($M = 0.33$) in **conceptual** replications
- ▶ Overall ES for average close replication was $d = 0.24$; for meta-analysis this was $d = 0.45$
- ▶ Heterogeneity in conceptual replications
 - ▶ No significant differences between the 3 sub-disciplines (cognitive, social, organisational)
 - ▶ The distinctive success rates of these sub-disciplines in terms of replication is not reflected in heterogeneity levels

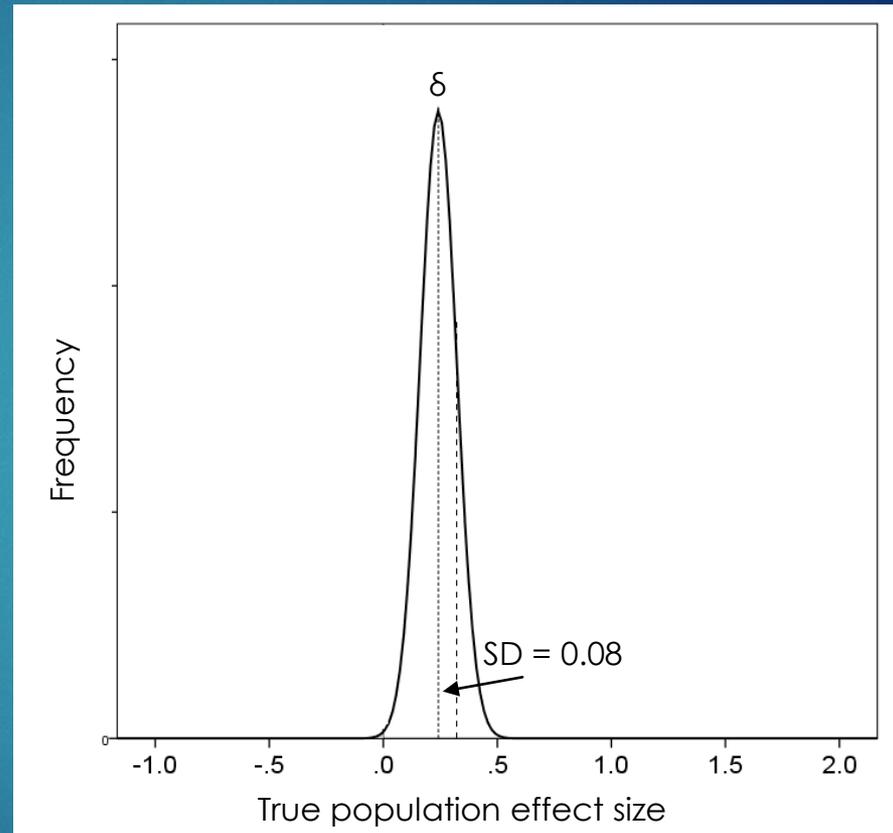
What does this level of heterogeneity mean?

- Average $\tau = 0.33$
- Cohen's d of 0.2/0.5/0.8 are often used as benchmarks for small/medium/large effects
 - All of these occur frequently in the distribution of true effect sizes
- Can expect replications to find results in the opposite direction



Close replications

- ▶ Average $\tau = 0.08$
- ▶ High consistency in results



Variability in heterogeneity – why?

- ▶ Mixing apples and oranges
broad versus narrow inclusion criteria
- ▶ Looked at moderators in a sub-set of 25 large meta-analyses
 - ▶ Looked at one moderator in each case
- ▶ No significant difference in heterogeneity between overall meta-analyses ($M = 0.34$) and subset based on moderators ($M = 0.36$)
- ▶ Broad/narrow inclusion criteria?
 - ▶ Narrow sub-sample, heterogeneity still high ($M = 0.29$)

Variability in heterogeneity – why?

Exploratory analyses

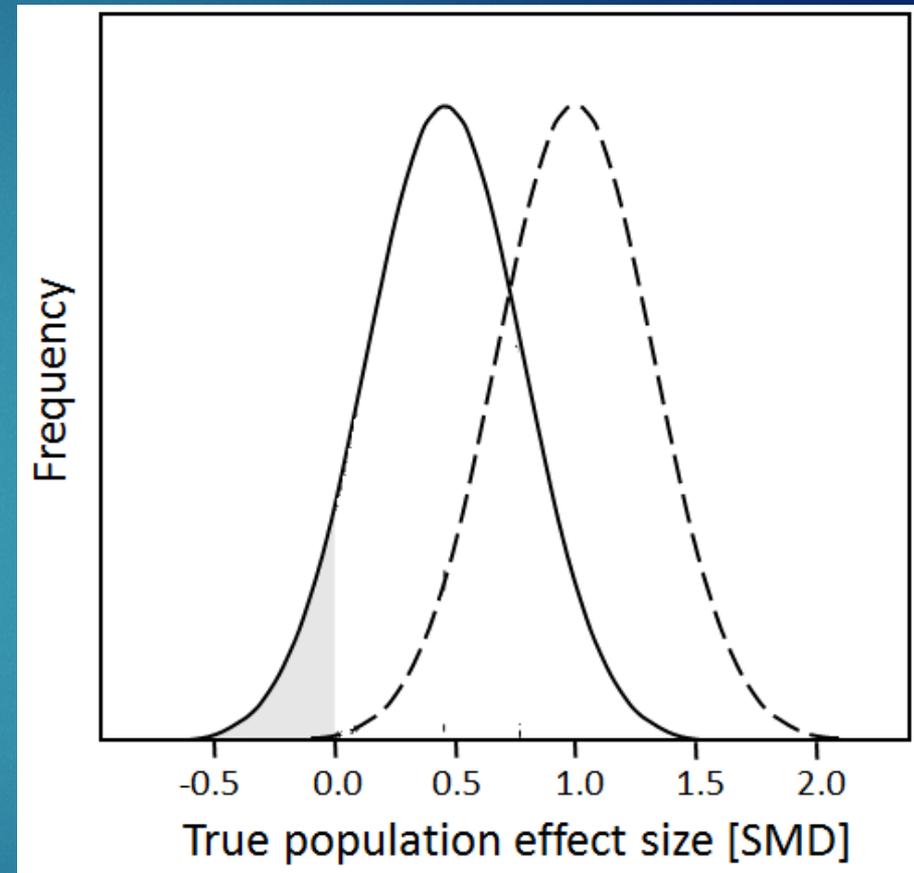
- ▶ Research areas with larger ES have greater heterogeneity (Kenny & Judd)
 - ▶ Strong relationship between mean d and τ
 - ▶ For close replications ($r = .70, p < .001$)
 - ▶ For conceptual replications ($r = .45, p < .001$)
- ▶ Maturity of a research field, or broader inclusion criteria?
 - ▶ Relationship between k and τ ($r = 0.30, p < .001$)
 - ▶ Establishing an effect -> exploring boundaries
 - ▶ Used a median date split
 - ▶ No significant difference in τ between the earlier and later dates

Conclusions

- ▶ Heterogeneity in close replications proved low – producing reliable results is possible
- ▶ Heterogeneity in close replications reduces power only marginally - for sample sizes that generate 80% power at zero heterogeneity...
 - ▶ For $\tau = 0.08$, power
 - ▶ Does not reduce for large effects
 - ▶ Drops to 78% for medium effects
 - ▶ Drops to 71% for small effects
- ▶ For Open Science Collaboration (2015), mean effect size was large ($d = 0.87$)
 - ▶ Power therefore not affected
 - ▶ 'Hidden moderators' typically of no concern

Conclusions

- ▶ Heterogeneity in meta-analyses is large (and not strongly affected by bias)
 - ▶ Mean ES reported in MA with large heterogeneity have limited use
- ▶ Research Planning
 - ▶ Difficult to estimate efficacy of an intervention (effect could be in opposite direction)
 - ▶ Heterogeneity and effect size determine how predictable the result of the 'next study' is



Conclusions

- ▶ $\tau = 0.33$ has a more dramatic effect on power
 - ▶ Drops to 71% for large effects
 - ▶ Drops to 66% for medium effects
 - ▶ Drops to 57% for small effects

Implications

- ▶ Cumulative knowledge
 - ▶ Science = quest to explain apparent complexity in observations through simpler fundamental principles
 - ▶ (Unexplained) heterogeneity is a measure of how much this quest fails

- ▶ Falsifiability of theories
 - ▶ Say test of theory X requires induction of good mood. We use mood induction procedure Y
 - ▶ When effectiveness of Y is debatable (large heterogeneity), failed test of theory X becomes meaningless
 - ▶ Weak tools undermine falsification and thereby good theoretic progress

 - ▶ When knowledge Y is used as a tool, we need to replicate as closely as possible

Limitations

- ▶ Difference in effect size between close replications ($d = 0.24$) and conceptual replications ($d = 0.45$)
 - ▶ How does low heterogeneity in close replications generalise to psychological research findings?
- ▶ We used Hunter-Schmidt meta-analysis model
 - ▶ Similar results for Hedges, and DerSimonian-Laird models
 - ▶ HS estimates of heterogeneity were slightly more conservative

Thank you

▶ Questions...

