

# 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
  - ▶  $\tau$  as measure that quantifies heterogeneity
    - ▶ Generally assumed that population ES for a given phenomenon follow a normal distribution
    - ▶  $\tau$  is their standard deviation
- ▶  $d$  and  $\tau$  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?

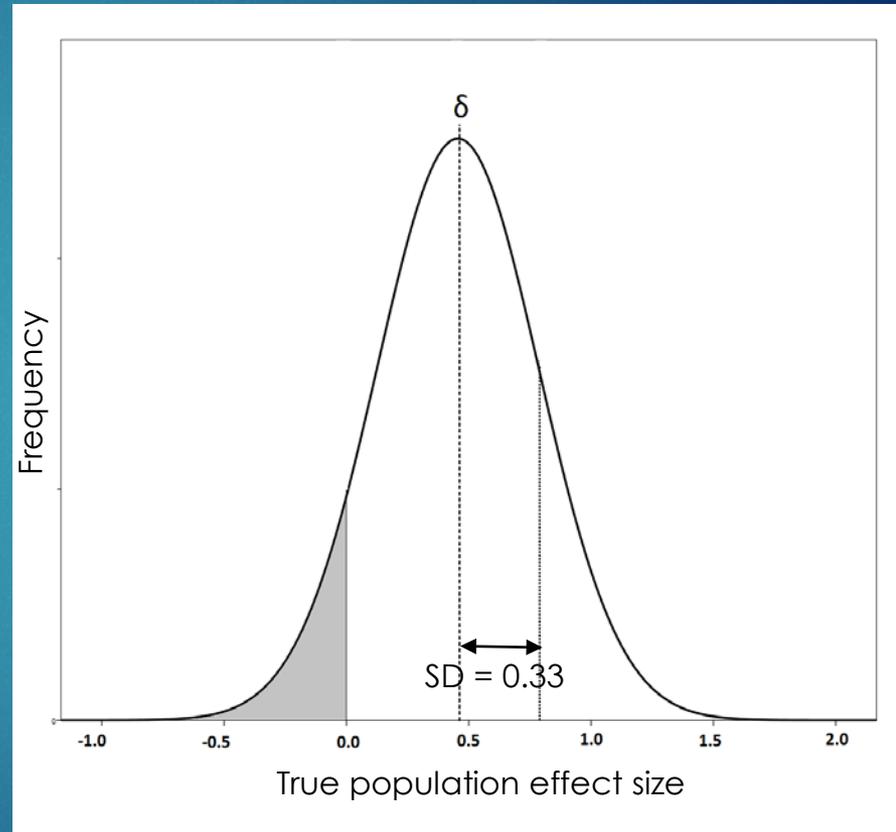
- ▶ Out of 147 meta-analyses...
  - ▶ 54% reported a measure of heterogeneity
  - ▶ Heterogeneity **quantified** in only 38 cases (26%)
- ▶ Post-PRISMA? (Preferred Reporting Items for Systematic Reviews and Meta-Analyses, 2009)
  - ▶ Heterogeneity only reported in 60% of cases

# Findings in our sample

- ▶ Average  $\tau$  was low ( $M = 0.08$ ) in **close** replications
- ▶ Average  $\tau$  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

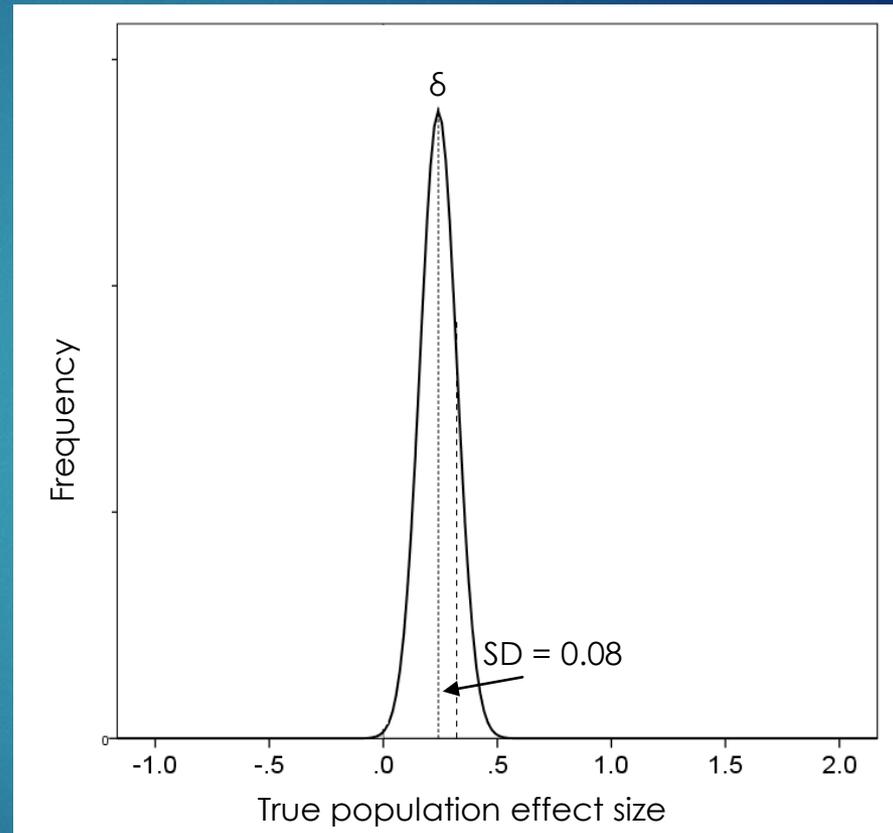
# What does this level of heterogeneity mean?

- 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

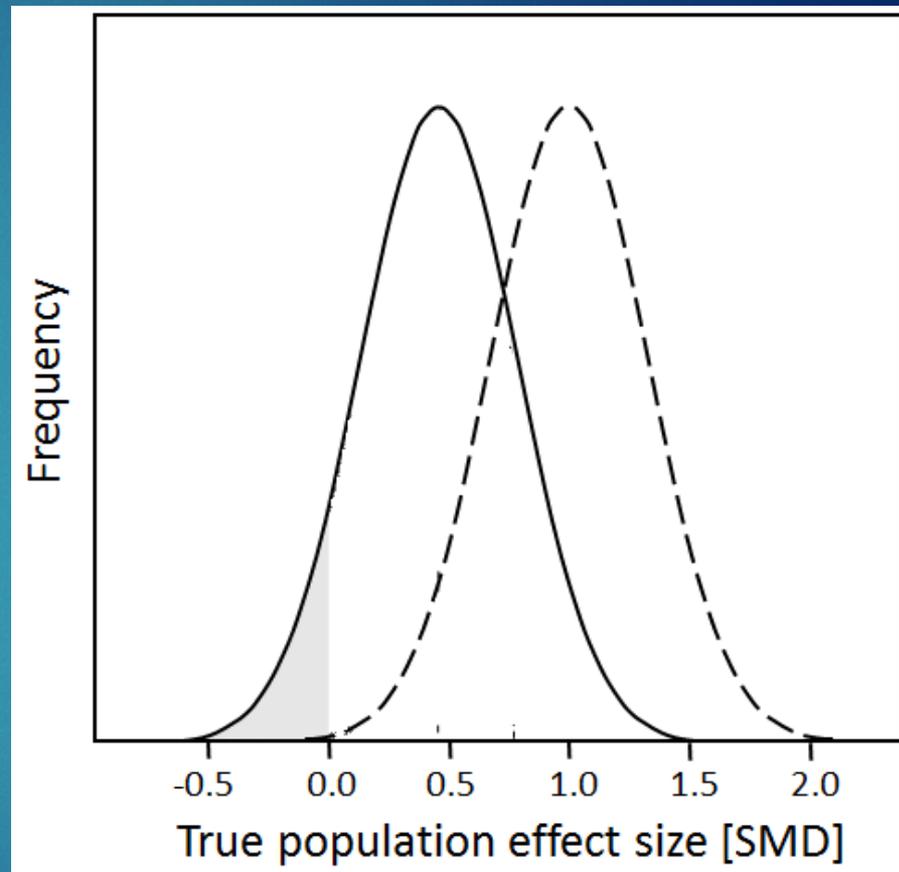
- ▶ Research areas with larger ES have greater heterogeneity (Kenny & Judd)
  - ▶ Strong relationship between mean  $d$  and  $\tau$
  - ▶ For close replications ( $r = .70, p < .001$ )
  - ▶ For conceptual replications ( $r = .45, p < .001$ )
  
- ▶ Maturity of a research field, or broader inclusion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

