

# EVAL 6970: Meta-Analysis Heterogeneity and Prediction Intervals

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# Agenda

- Identifying and quantifying heterogeneity
  - Review questions
  - In-class activity
- Prediction intervals
  - Review questions
  - In-class activity

# Quantifying Heterogeneity

- Although we are usually concerned with the dispersion in true effect sizes, observed dispersion includes both true variance and random error
- The mechanism used to isolate true variance is to compare the observed dispersion with the amount expected if all studies shared a common effect size
  - The excess is assumed to reflect real differences among studies
  - This portion of the variance is used to create indices of heterogeneity

# Indices of Heterogeneity

- The  $Q$  statistic (the weighted sum of squares;  $WSS$ ) and its  $p$ -value serve as a statistical test of significance (compared to the expected  $WSS$ ) and address the viability of the null hypothesis that the true dispersion is exactly zero, and not the excess amount of dispersion
  - Sensitive to the number of studies and is metric-free (i.e., the effect size index)

# Indices of Heterogeneity

- The estimate of  $\tau^2$  (i.e.,  $T^2$ ) serves as the between-study variance and  $\tau$  (i.e.,  $T$ ) serves as the standard deviation of the true effects and both describe the distribution of true effects
  - They are sensitive to the effect size metric, but not the number of studies
  - 95% LL and UL can be computed for both  $T^2$  and  $T$

# Indices of Heterogeneity

- $T$  is useful for understanding the range of effects and is on the same metric as the effects
  - If a summary effect were 0.50 and  $T$  was 0.10, most effects fall in the range of 0.30 to 0.70
  - If a summary effect were 0.50 and  $T$  was 0.20, then most effects fall in the range of 0.10 and 0.90
  - This index is useful regarding interpretation (e.g., utility of the intervention)

# Indices of Heterogeneity

- $I^2$  provides information regarding what proportion of observed variance is real and is a ratio with a range from 0% to 100%
  - An  $I^2$  near zero indicates that the dispersion is attributable to random error, and any attempt to explain the variance is an attempt to explain something that is random
  - As  $I^2$  increases, some of the variance is real and can potentially be explained by subgroup analysis or meta-regression
  - 95% LL and UL (uncertainty intervals) can be computed for  $I^2$

# Factors Affecting Heterogeneity Statistics

	Range of possible values	Depends on number of studies	Depends on scale
$Q$	$0 \leq Q$	✓	
$p$	$0 \leq p \leq 1$	✓	
$T^2$	$0 \leq T^2$		✓
$T$	$0 \leq T$		✓
$I^2$	$0\% \leq I^2 < 100\%$		

- All indices are based on  $Q$  (in relation to  $df$ )
- Each is useful for a specific purpose



# The $Q$ Statistic

- The first step in partitioning study-to-study variation is to compute  $Q$

$$Q = \sum_{i=1}^k W_i (Y_i - M)^2$$

- or

$$Q = \sum_{i=1}^k \left( \frac{Y_i - M}{S_i} \right)^2$$

# The Expected Value of $Q$

- Next, determine the expected value of  $Q$  (on the assumption that all studies share a common effect size)
- Because  $Q$  is a standardized measure the expected value does not depend on the metric of the effect size, the expected value ( $WSS$ ) is

$$df = k - 1$$

## Using $Q$ to Test Homogeneity of Effects

- To test the null hypothesis that all studies share a common effect size use  $Q$  and  $df$  from previous slides
- $Q$  will follow a central chi-squared distribution with degrees of freedom equal to  $k - 1$ , and a  $p$ -value for any observed value of  $Q$  can be obtained

$$= \text{CHIDIST}(Q, df)$$

# Estimating $\tau^2$

- $\tau^2$  is the variance of the true effect sizes
- Since  $\tau^2$  cannot be directly computed, it is estimated with  $T^2$

$$T^2 = \frac{Q - df}{C}$$

- Where

$$C = \sum W_i - \frac{\sum W_i^2}{\sum W_i}$$

# Estimating $\tau$

- $\tau$  is the standard deviation of the true effect sizes and the estimate of this parameter is  $T$

$$T = \sqrt{T^2}$$

- $T$  is on the same scale as the effect size and can be used to describe the distribution of effect sizes

# Estimating $I^2$

- $I^2$  represent the proportion of observed variance that reflect real differences in effect sizes and is the ratio of excess dispersion to total dispersion

$$I^2 = \left( \frac{Q - df}{Q} \right) \times 100\%$$

# Estimating $I^2$

- $I^2$  reflects the overlap of the study confidence intervals, which is not dependent on the actual location or dispersion of the true effects
- Therefore,  $I^2$  is a measure of inconsistency across the findings of the studies, and not real variation across the underlying true effects

# Review Questions

1. What question does  $Q$  address?
2. What information does  $T^2$  and  $T$  provide?
3. What information does  $I^2$  provide and how is it interpreted?



# Today's First In-Class Activity

- Using “Data Sets for Heterogeneity Analysis XLSX” from the course Website
  - Calculate the heterogeneity statistics for Step 5 and Step 9 (except for the prediction intervals) for each data set
  - In some cases, you will need to refer to the book (formulas and page numbers are provided)
  - Verify your results using Meta-Analysis 2.0
  - Interpret the heterogeneity statistics

# Prediction Intervals

- The precision of the summary effect is addressed by the LL and UL confidence intervals, which reflects only error ( $V_{M^*}$ )
- The distribution of true effect sizes is addressed by the prediction interval which incorporates true dispersion ( $T^2$ ) as well as error ( $V_{M^*}$ )

# Prediction Intervals

- Based on sample values

$$LL_{pred} = M^* - t_{df}^{\alpha} \sqrt{T^2 + V_{M^*}}$$

- And

$$UL_{pred} = M^* + t_{df}^{\alpha} \sqrt{T^2 + V_{M^*}}$$

- With

$$df = k - 2$$

$$= \text{TINV}(\alpha, df)$$

# Prediction Intervals

- Confidence intervals reflect the accuracy of the mean (in 95% of cases the mean effect size falls in the summary effect)
- The prediction intervals address the dispersion of effects (in 95% of cases the true effect in a new study will fall within the prediction intervals)

# Review Questions

1. What information about heterogeneity do prediction intervals provide?
2. How are prediction intervals interpreted?
3. How does this interpretation differ from the interpretation of  $M^*$ ?

# Today's Second In-Class Activity

- Using “Data Sets for Heterogeneity Analysis XLSX” from the course Website
  - Calculate the prediction intervals (Step 9) of  $M^*$  for each data set
  - Interpret the prediction intervals in contrast to  $M^*$