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Synthesizing Individual Participant Data Obtained From Complex Sampling Surveys: A Two-Stage IPD Meta-Analysis Approach

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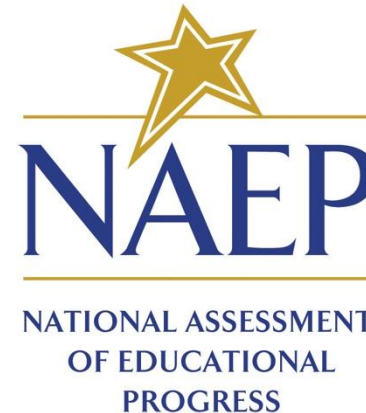
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What Are Complex Surveys?

Examples

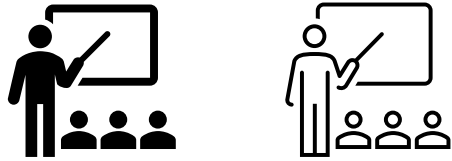
- Program for International Student Assessment (PISA)
- Trends in International Mathematics and Science Study (TIMSS)
- National Assessment of Educational Progress (NAEP)
- National Educational Panel Study (NEPS)



What Are Complex Surveys?

- Increasing the understanding of critical factors influencing teaching and learning
- Identifying key educational issues: educational inequalities
- Informing national strategies for monitor and improve the educational system

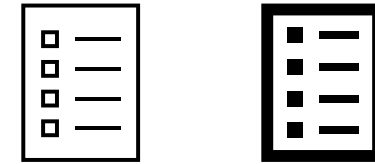
What Are Complex Surveys?



**Multistage
Sampling**

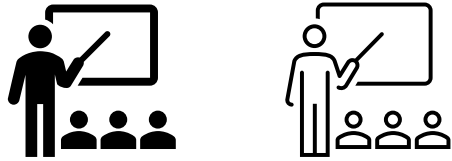


Survey Weights



**Rotated
Questionnaire
Design**

What Are Complex Surveys?

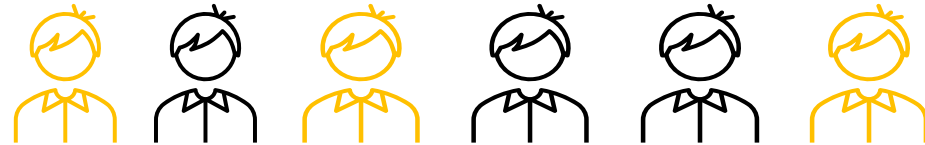


**Multistage
Sampling**

1st stage



2nd stage



What Are Complex Surveys?

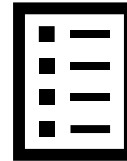
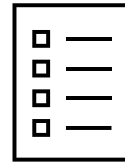
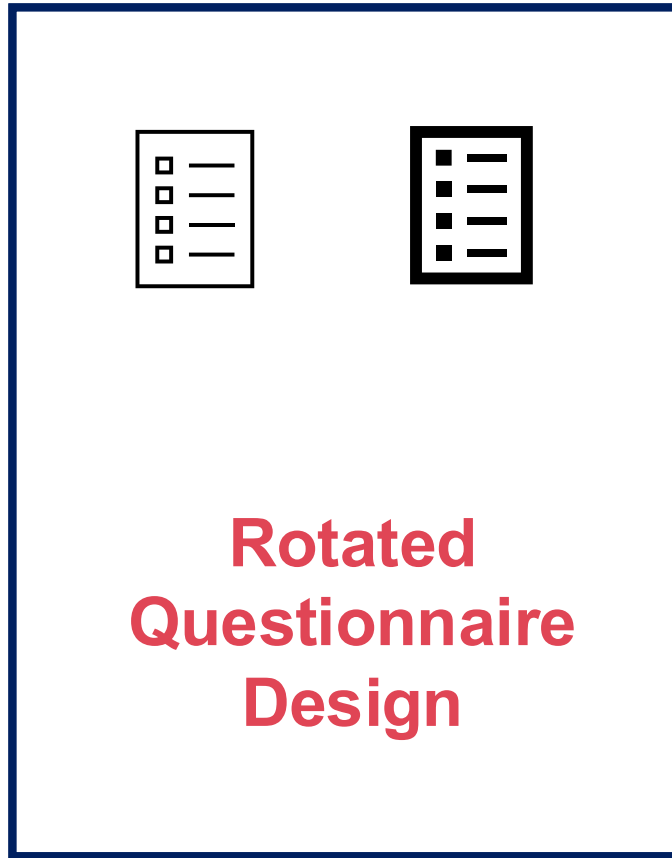


Survey Weights

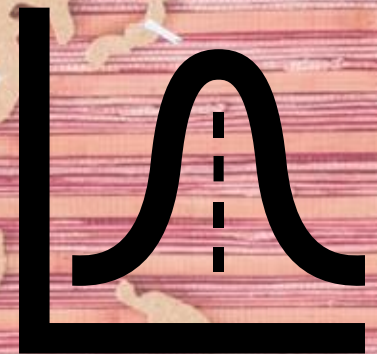
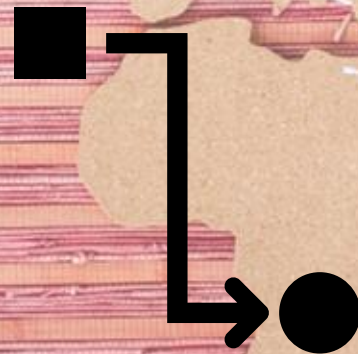
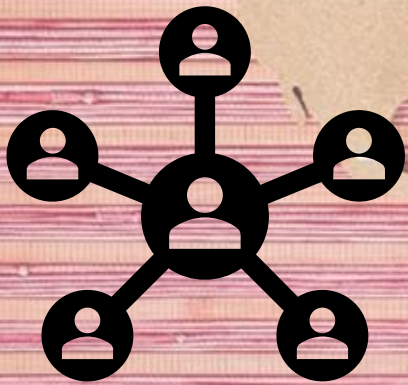
Total weights refer to the weight components that reflect the inclusion probability of a school and a student of being selected.

$$W_{ij} = \frac{1}{p_{ij}}$$

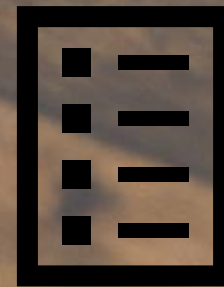
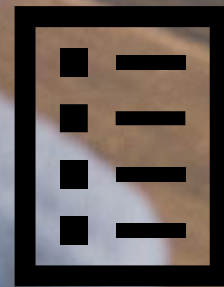
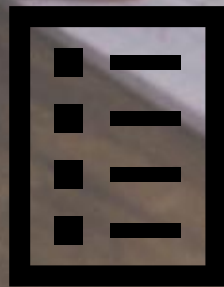
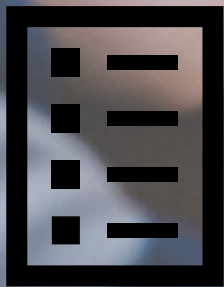
What Are Complex Surveys?



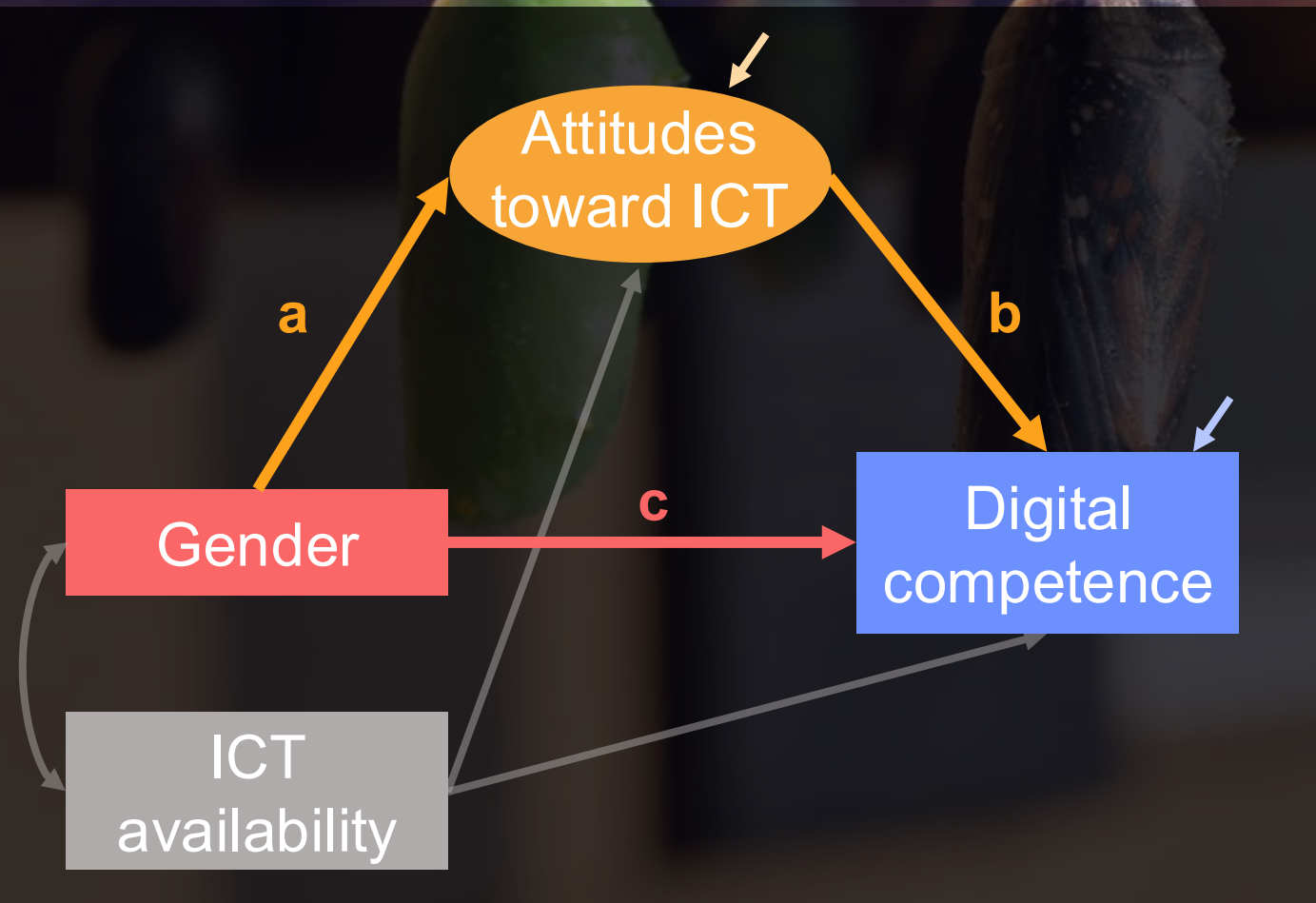
Large and Representative Samples



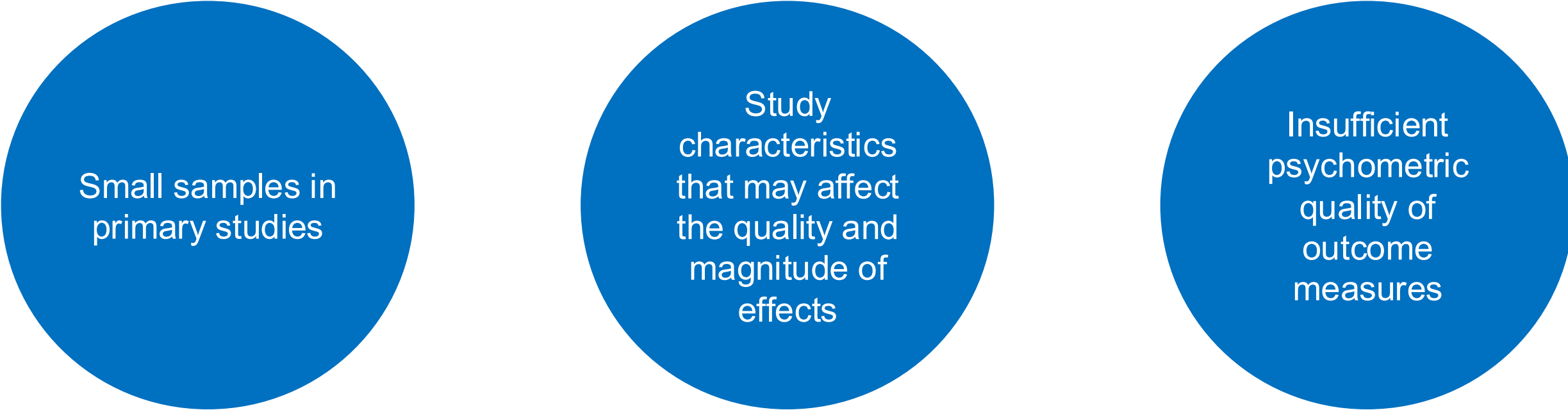
High Quality Measures



Raw Data



Challenges of Meta-Analysis in Educational Research

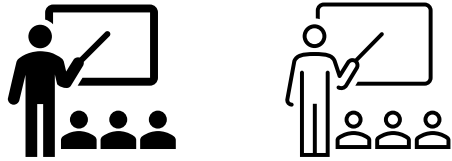


Small samples in
primary studies

Study
characteristics
that may affect
the quality and
magnitude of
effects

Insufficient
psychometric
quality of
outcome
measures

Challenges in the Synthesis of Complex Survey Data



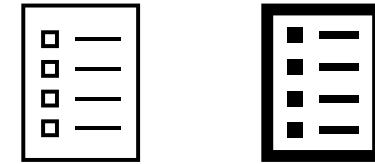
**Multistage
Sampling**

Stapleton et al., (2016)



Survey Weights

Rutkowski, Gonzalez, Joncas & von Davier (2010)

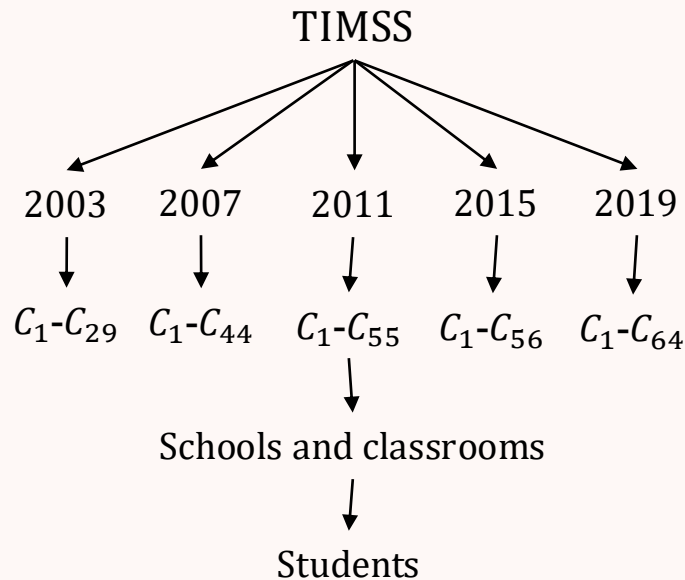


**Rotated
Questionnaire
Designs**

von Davier et al., (2009)

Two-Stage Individual Participant Data Meta-Analysis

Stage 1: Raw Data Analysis



- Hierarchical structure
- Sampling weights
- Plausible values
- Measurement Invariance

i: Students, *j*: Classrooms, *k*: Countries

Stage 2: Meta-Analysis

- Multivariate meta-analysis
- Multilevel meta-analysis
- Mixed-effects meta-regression

Two-Stage Individual Participant Data Meta-Analysis

Advantages

Level 1 (individual participants):

$$Y_{ijk} = \beta_{0jk} + \beta_{1jk}X_{ijk} + e_{ijk}$$

Level 2 (clusters):

$$\beta_{0jk} = \beta_{00k} + \beta_{01k}Z_{jk} + u_{0jk}$$

$$\beta_{1jk} = \beta_{10k} + u_{1jk}$$

Level 3 (primary studies):

$$\beta_{00k} = \gamma_{000} + v_{00k}$$

$$\beta_{01k} = \gamma_{010} + v_{01k}$$

$$\beta_{10k} = \gamma_{100} + v_{10k}$$

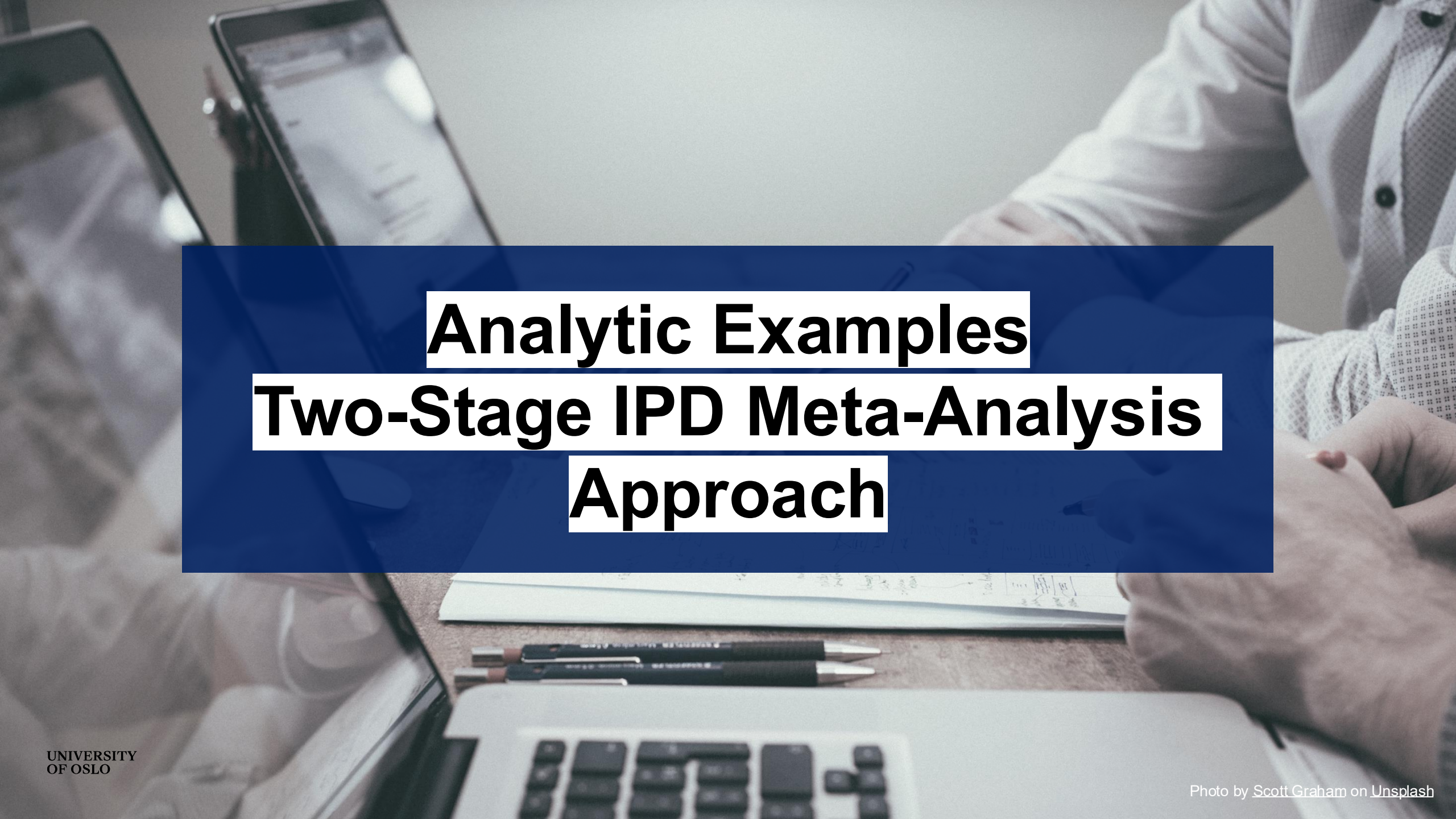
$$\beta_{01k} = \theta_k + r_k \text{ with } r_k \sim N(0, \sigma_{r_k}^2).$$

$$\theta_k = \mu + u_k \text{ with } u_k \sim N(0, \sigma_u^2).$$

Complex Survey
Designs

Measurement
Heterogeneity

Multilevel
Structures



Analytic Examples

Two-Stage IPD Meta-Analysis

Approach

Two-Stage Individual Participant Data Meta-Analysis

Stage 1

To what extent do girls and boys in secondary education differ in their digital skills?

ICILS 2013

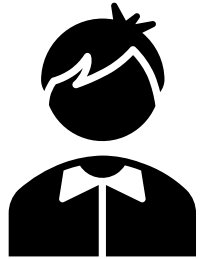
ICILS 2018

Constructs of interest	• Gender • CIL
Complex data structure	• Study ID • Country ID • School ID • Student ID • Student weights • Jackknife codes

Two-Stage Individual Participant Data Meta-Analysis

Stage 1

To what extent do girls and boys in secondary education differ in their digital skills?



$$d = \frac{\bar{X}_G - \bar{X}_B}{\sqrt{\frac{(n_G - 1)SD_G^2 + (n_B - 1)SD_B^2}{n_G + n_B - 2}}} \quad \text{with } v_d = \frac{n_G + n_B}{n_G n_B} + \frac{d^2}{2(n_G + n_B)}$$

$$g = \left(1 - \frac{3}{4(n_G + n_B - 2) - 1}\right) \cdot d \quad \text{with } v_g = \left(1 - \frac{3}{4(n_G + n_B - 2) - 1}\right)^2 \cdot v_d$$

Two-Stage Individual Participant Data Meta-Analysis

Stage 2

Three-level random-effects model with effect sizes nested in

countries

Level 1 (sampling variance): $\beta_{jk} = \theta_{jk} + r_{jk}$ $r_{jk} \sim N(0, \sigma_r^2)$, $q_{jk} \sim N(0, \sigma_q^2)$, $u_k \sim N(0, \sigma_u^2)$

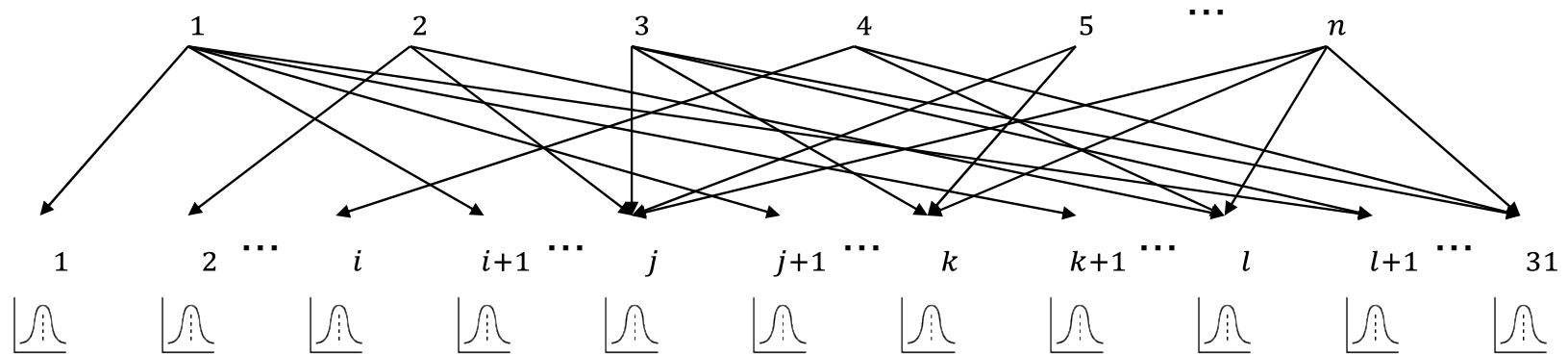
Level 2 (within countries): $\theta_{jk} = \lambda_k + q_{jk}$

Level 3 (between countries): $\lambda_k = \mu + u_k$

Countries ()

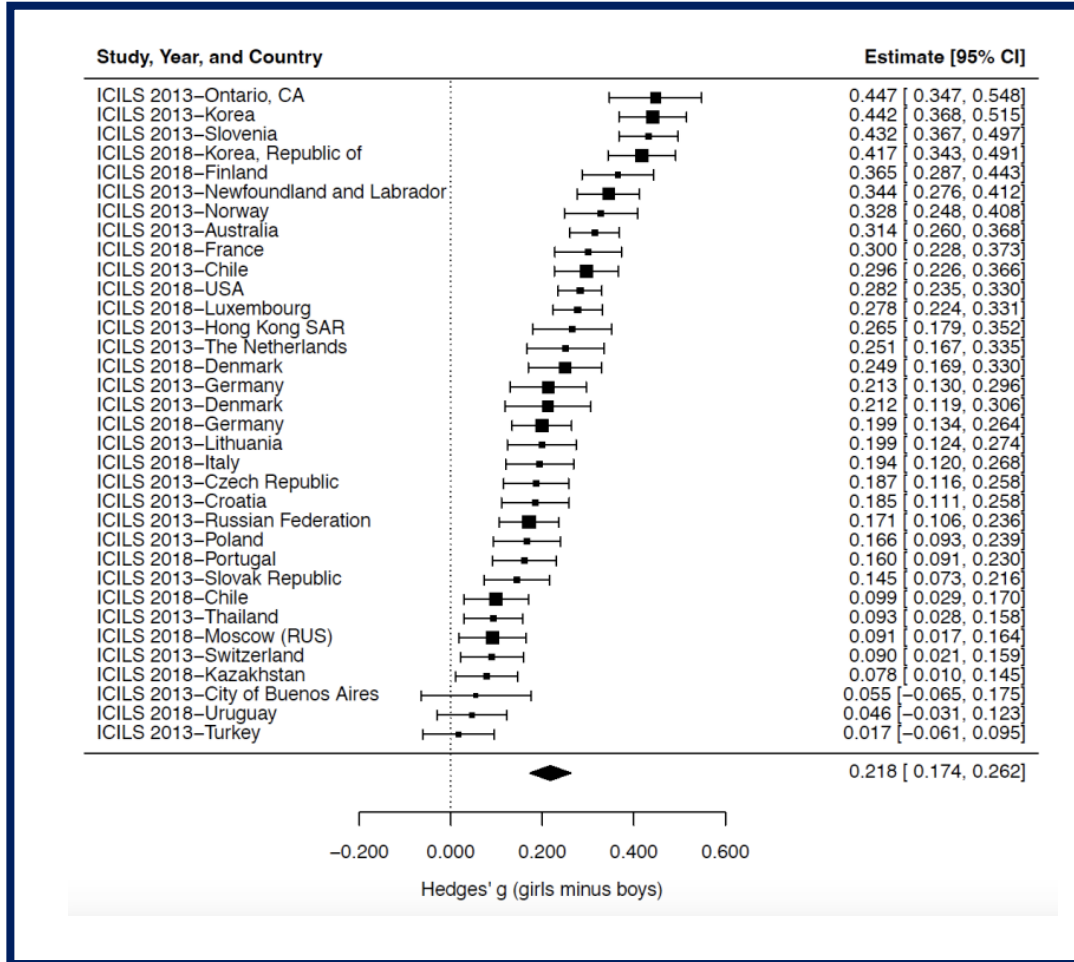
Effect sizes ()

Sampling variances



Two-Stage Individual Participant Data Meta-Analysis

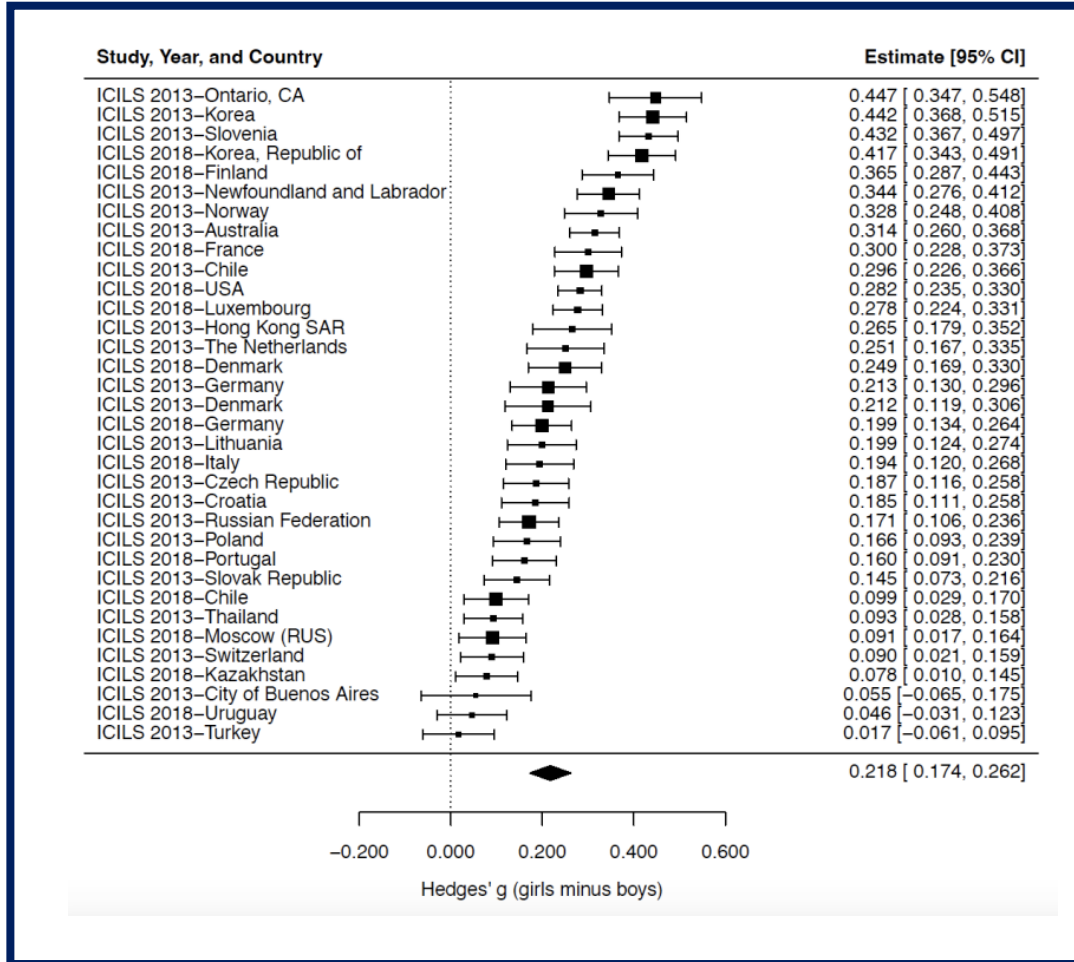
Individual Participant Data



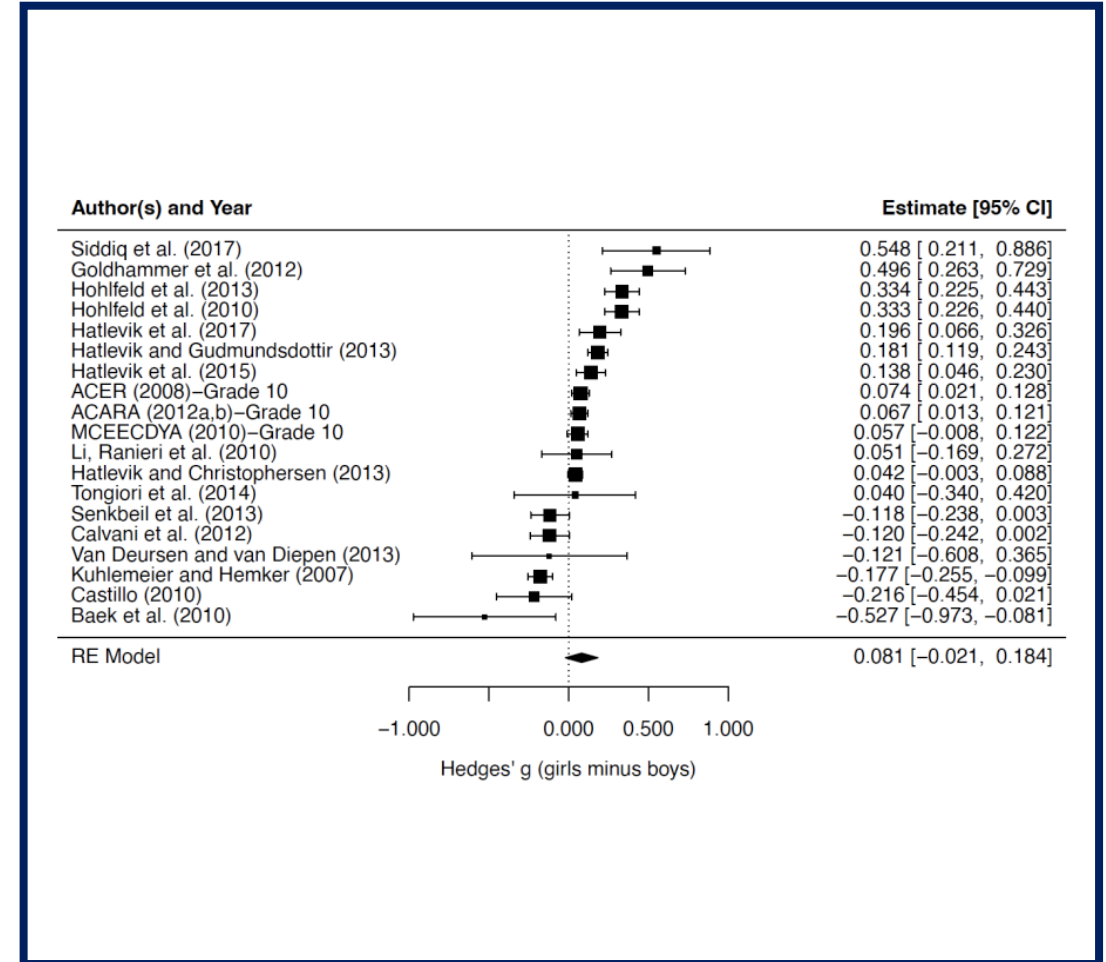
$$\bar{g}_{IPD} = 0.218 \text{ (95\% CI [0.174, 0.262])}$$

Two-Stage Individual Participant Data Meta-Analysis

Individual Participant Data



Aggregate Data



Two-Stage Individual Participant Data Meta-Analysis

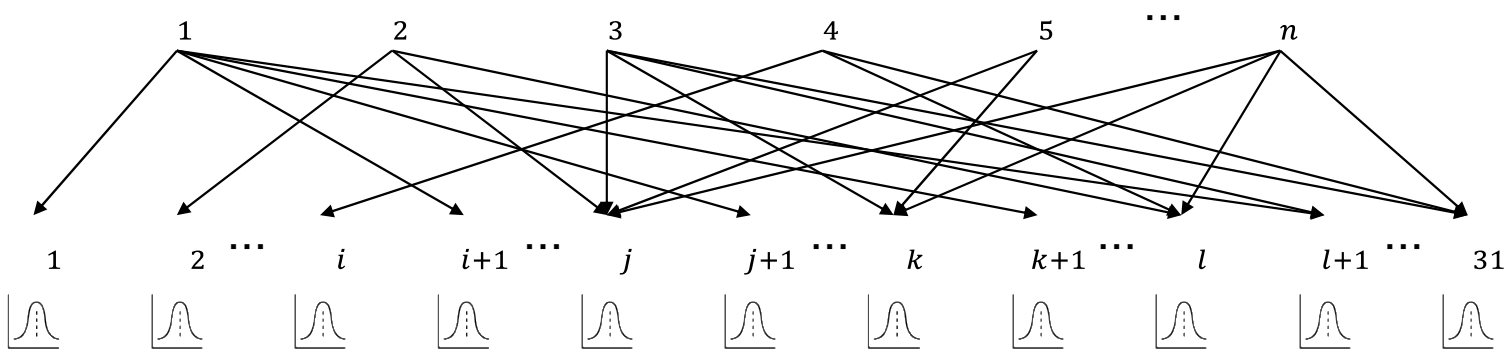
Combining AD and IPD

Three-level random-effects model with effect sizes nested in countries/studies

Studies ()

Effect sizes ()

Sampling variances



IPD ? AD

Two-Stage Individual Participant Data Meta-Analysis

Stage 2 - Moderation Analysis AD vs. IPD

Baseline model	$\hat{\mu}$ [95 % CI]	B [95 % CI]	τ^2_{IPD} [95 % CI]	τ^2_{AD} [95 % CI]	γ^2_{IPD}	γ^2_{AD}
Model 1: Data-specific effect sizes, data-specific between-country, and data-specific between-sample residual heterogeneity	.074 [-.003, .177]	.144 [.032, .256]	.003 [.001,.013]	.035 [.011,.093]	.009 [.001, .020]	.002 [.000, .073]

Two-Stage Individual Participant Data Meta-Analysis

Stage 2 - Moderation Analysis AD vs. IPD

Baseline model	$\hat{\mu}$ [95 % <i>CI</i>]	B [95 % <i>CI</i>]	τ^2_{IPD} [95 % <i>CI</i>]	τ^2_{AD} [95 % <i>CI</i>]	γ^2_{IPD}	γ^2_{AD}			
Model 1: Data-specific effect sizes, data-specific between-country, and data-specific between-sample residual heterogeneity	.074 [-.003, .177]	.144 [.032, .256]	.003 [.001,.013]	.035 [.011,.093]	.009 [.001, .020]	.002 [.000, .073]			
Model 2: Data-specific effect sizes, overall between-country heterogeneity, and data-specific between-sample residual heterogeneity	.031 [-.079 .140]	.187 [.080, .295]	.003 [.000 .016]	.035 [.013, .095]	.008 [.000, .020]				
	df	AIC	BIC	AICc	logLik	LRT	pval	QE	tau^2
Full	6	-40.5818	-28.9909	-38.6728	26.2909			479.8761	NA
Reduced	5	-41.3649	-31.7057	-40.0315	25.6824	1.2170	0.2700	479.8761	NA

Two-Stage Individual Participant Data Meta-Analysis

Stage 2 - Moderation Analysis AD vs. IPD

Baseline model	$\hat{\mu}$ [95 % CI]	B [95 % CI]	τ_{IPD}^2 [95 % CI]	τ_{AD}^2 [95 % CI]	γ_{IPD}^2	γ_{AD}^2
Model 1: Data-specific effect sizes, data-specific between-country, and data-specific between-sample residual heterogeneity	.074 [-.003, .177]	.144 [.032, .256]	.003 [.001,.013]	.035 [.011,.093]	.009 [.001, .020]	.002 [.000, .073]
Model 2: Data-specific effect sizes, overall between-country heterogeneity, and data-specific between-sample residual heterogeneity	.031 [-.079 .140]	.187 [.080, .295]	.003 [.000 .016]	.035 [.013, .095]	.008 [.000, .020]	
Model 3: Data-specific effect sizes, overall between-country heterogeneity, and overall between-sample residual heterogeneity	.065 [-.012, .142]	.158 [.070, .245]		.002 [.000, .013]	.002 [.000, .013]	

Two-Stage Individual Participant Data Meta-Analysis

Example 2

To what extent is class-average student achievement in mathematics related to individual students' mathematics self-concept after controlling for students' individual mathematics performance in primary school (BFLPE)?



Constructs of interest

- Mathematics achievement
- Self-concept

Complex data structure

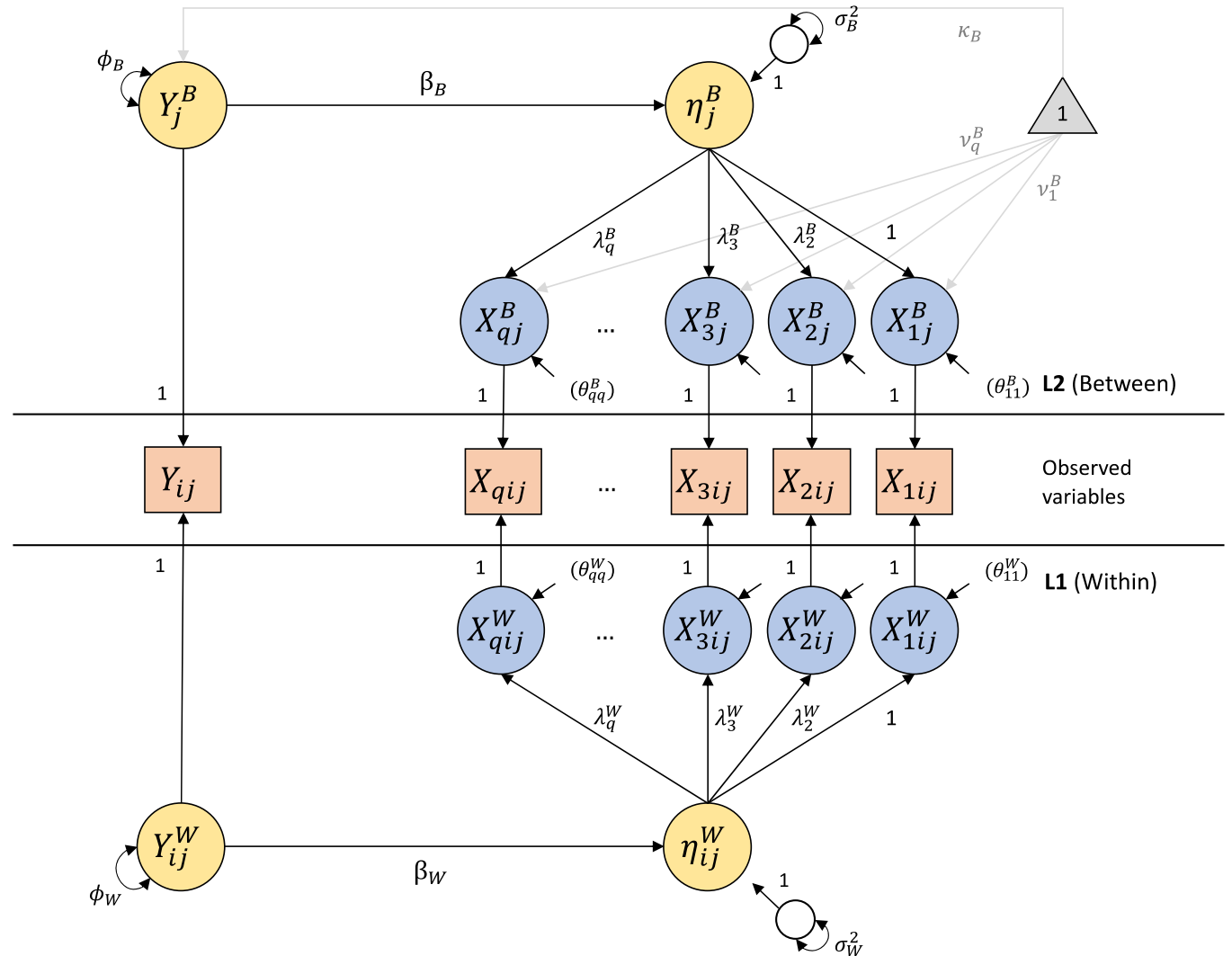
- Study ID
- Country ID
- Classroom ID
- Student ID
- Student weights
- Classroom weights

Two-Stage Individual Participant Data Meta-Analysis

Stage 1

Multilevel SEM to estimate contextual effects

$$ES_{BFLPE} = (\beta_B - \beta_W)$$



Two-Stage Individual Participant Data Meta-Analysis

Stage 2

Three-level random-effects model with effect sizes nested in countries

Level 1 (sampling variance): $\beta_{jk} = \theta_{jk} + r_{jk} \quad r_{jk} \sim N(0, \sigma_r^2), q_{jk} \sim N(0, \sigma_q^2), u_k \sim N(0, \sigma_u^2)$

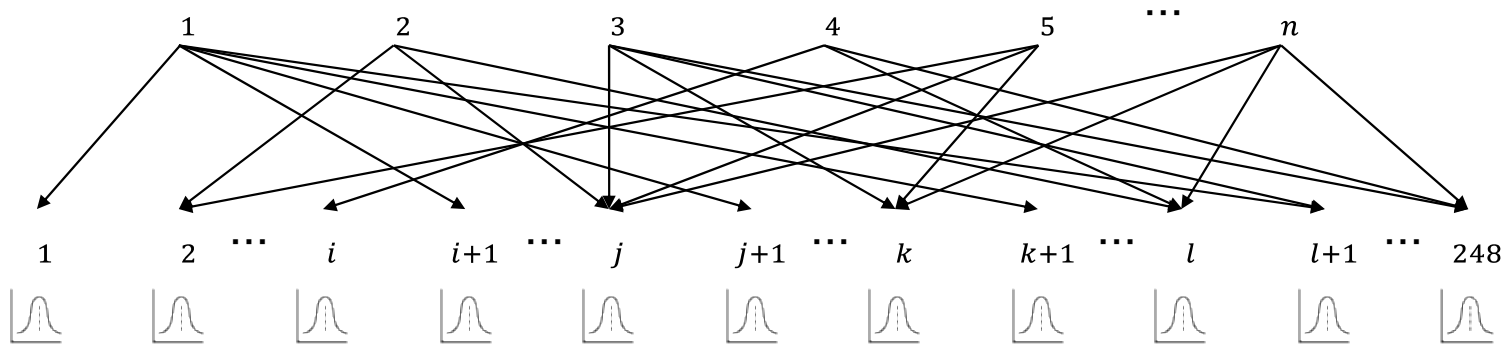
Level 2 (within countries): $\theta_{jk} = \lambda_k + q_{jk}$

Level 3 (between countries): $\lambda_k = \mu + u_k$

Countries ()

Effect sizes ()

Sampling variances



Two-Stage Individual Participant Data Meta-Analysis

Stage 2

Three-level random-effects model with effect sizes nested in cycles

Level 1 (sampling variance): $\beta_{jk} = \theta_{jk} + r_{jk}$ $r_{jk} \sim N(0, \sigma_r^2)$, $q_{jk} \sim N(0, \sigma_q^2)$, $u_k \sim N(0, \sigma_u^2)$

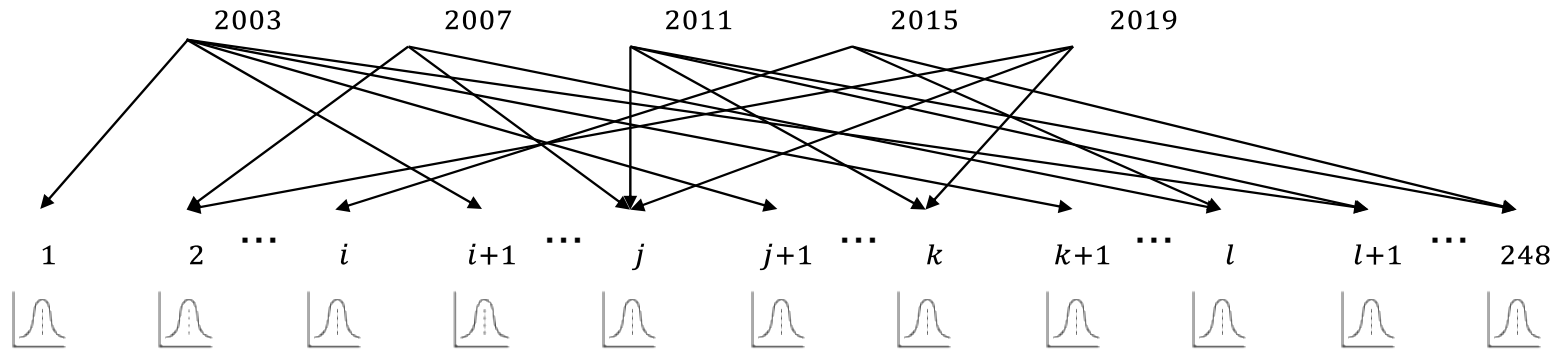
Level 2 (within cycles): $\theta_{jk} = \lambda_k + q_{jk}$

Level 3 (between cycles): $\lambda_k = \mu + u_k$

ILSA cycles

Effect sizes ()

Sampling variances



Two-Stage Individual Participant Data Meta-Analysis

Stage 2

Four-level cross-classified random-effects model with **countries** and **cycles**

Level 1: $\beta_{j(kl)} = \theta_{j(kl)} + r_{j(kl)}$ $r_{j(kl)} \sim N(0, \sigma_r^2), q_{j(kl)} \sim N(0, \sigma_q^2),$

Level 2: $\theta_{j(kl)} = \lambda_{(kl)} + q_{j(kl)}$ $u_k \sim N(0, \sigma_u^2), p_l \sim N(0, \sigma_p^2)$

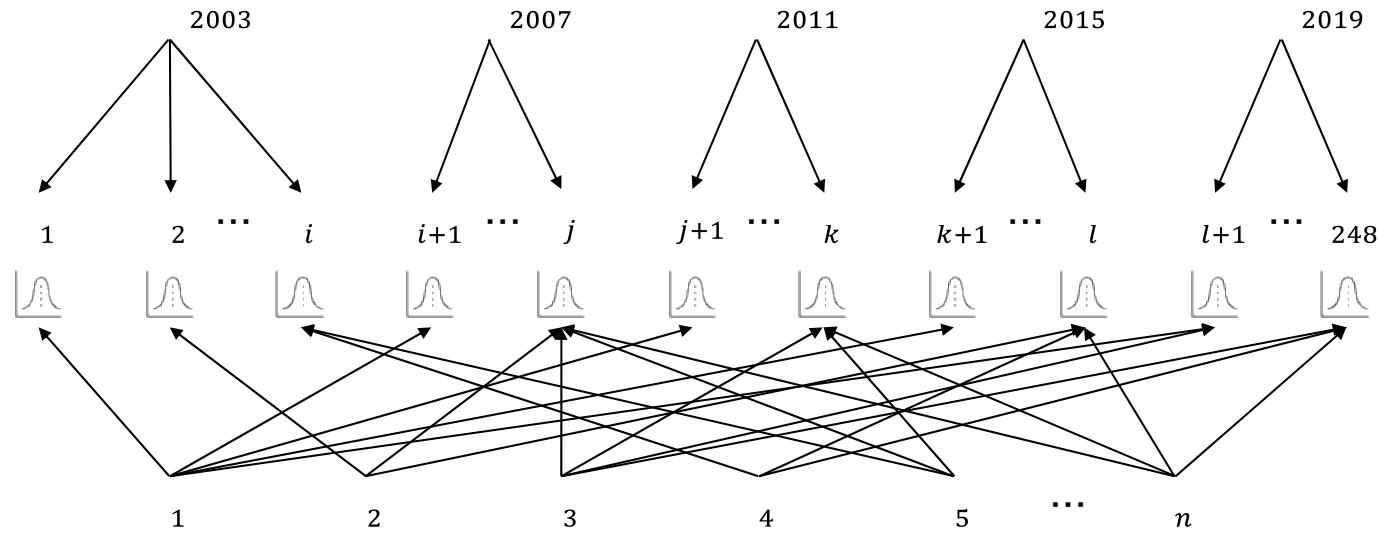
Levels 3 and 4: $\lambda_{(kl)} = \mu + u_k + p_l$

ILSA cycles

Effect sizes ()

Sampling variances

Countries ()



Two-Stage Individual Participant Data Meta-Analysis

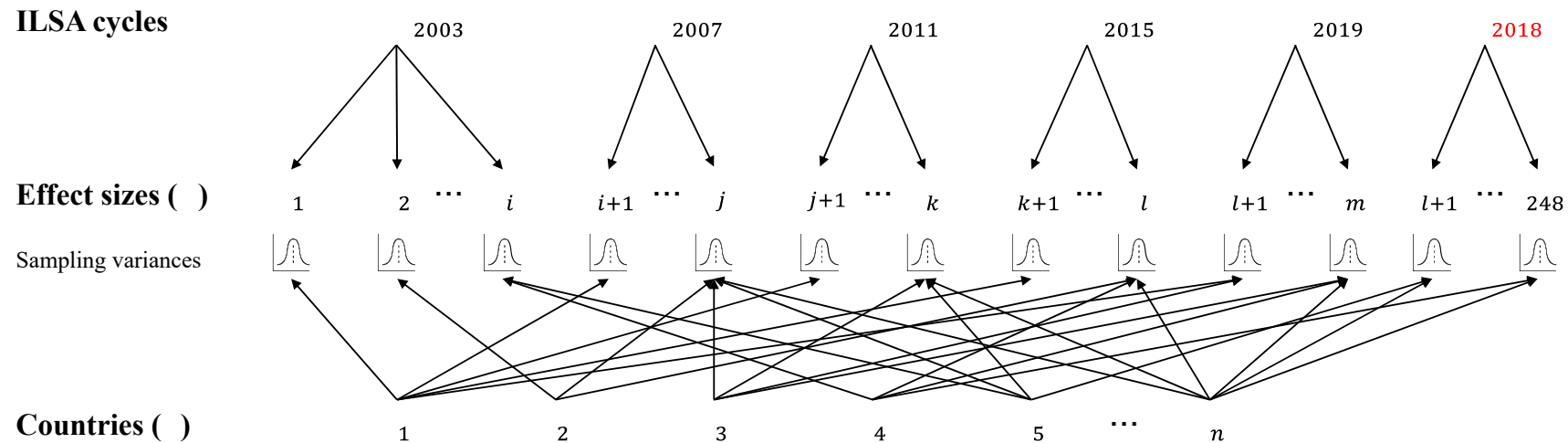
Stage 2

Four-level cross-classified random-effects model with **countries** and **cycles**

Level 1: $\beta_{j(kl)} = \theta_{j(kl)} + r_{j(kl)}$ $r_{j(kl)} \sim N(0, \sigma_r^2), q_{j(kl)} \sim N(0, \sigma_q^2),$

Level 2: $\theta_{j(kl)} = \lambda_{(kl)} + q_{j(kl)}$ $u_k \sim N(0, \sigma_u^2), p_l \sim N(0, \sigma_p^2)$

Levels 3 and 4: $\lambda_{(kl)} = \mu + u_k + p_l$



Two-Stage Individual Participant Data Meta-Analysis

Stage 2

Baseline model	$\hat{\mu}$ [95 % CI]	τ^2_{ES} [95 % CI]	σ^2_C [95 % CI]	σ^2_S [95 % CI]	I^2_{ES}	I^2_C	I^2_S
Model 1: Standard random-effects model	-.458 [-.481, -.436]	.025 [.020, .030]	-	-	82.4 %	-	-
Model 2: Three-level random-effects model with effect sizes nested in countries	-.451 [-.486, -.416]	.004 [.002, .006]	.021 [.014, .031]	-	12.1 %	70.0 %	-
Model 3: Three-level random-effects model with effect sizes nested in TIMSS cycles	-.458 [-.484, -.432]	.025 [.020, .031]	-	.000 [.000, .002]	82.4 %	-	0.0 %
Model 5: Four-level cross-classified random-effects model	-.452 [-.489, -.415]	.003 [.001, .006]	.021 [.014, .031]	.000 [.000, .002]	10.9 %	70.6 %	0.7 %

Two-Stage Individual Participant Data Meta-Analysis

Stage 2

Moderator	Three-level mixed-effects meta-regression			
	Model 2a		Model 2b	
	<i>B</i>	<i>SE</i>	<i>B</i>	<i>SE</i>
Intercept	-0.137	0.174	-0.812	0.225
<i>Cultural dimensions</i>				
PDI	-0.002	0.002	-	-
IDV	-0.002	0.001	-	-
MAS	-0.001	0.001	-	-
UAI	-0.001	0.001	-	-
LTO	0.000	0.001	-	-
IVR	0.000	0.002	-	-
<i>Economic development</i>				
HDI	-	-	0.426*	0.258
<i>Moderator test</i>				
$Q_M(df)$	8.6 (6), $p = .20$		5.5 (1), $p = .02$	
<i>Variance explanation</i>				
R_{ES}^2	0.0%		2.5%	
R_C^2	5.5%		12.3%	

Potential of Two-Stage IPD Meta-Analysis

1. Enlarge the generalizability of meta-analytic conclusions
2. Two-stage IPD meta-analysis enables researchers to synthesize information from complex surveys studies
3. Standardized analyses across studies
4. Direct and model-based generation of the effect sizes of interest
5. Appropriate handling of statistical dependencies in meta-analytic data sets from complex sampling surveys



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