

# Metric Misspecification due to Test Multidimensionality and Consequences for the Measurement of Growth

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Ideas In Testing Research Seminar  
November 4, 2022





- Educational research outcomes frequently rely on an assumption that measurement metrics have interval-level properties.
- Education measurement scales, including the latent scales derived from item response theory (IRT) models, may lack interval scale properties that permit comparisons of score gains (Ballou, 2009; Betebenner, 2011; Michell, 2009).
- While most investigators know enough to be suspicious of interval-level claims, and in some cases even question their findings in light of such suspicions, what is absent is an understanding of the measurement conditions that create metric distortions.



- ECLS-K (Early Childhood Longitudinal Study) Reading Assessment
  - ▶ possible dimensionality issue in the test items
    - e.g. items on sub-domains including basic skills, initial understanding, developing interpretation, and critical stance
  - ▶ dimensionality is related to different item types
    - e.g. "name letter" (easier) versus "decoding" (more difficult) items
  - ▶ dimensions are highly correlated
  - ▶ Unidimensional Item Response Theory (UIRT) model is used to scale the test scores
- We seek to simulate multidimensionality of the form on ECLS-K and examine metric distortion when 2PL is applied.



- Two-dimensional response data
  - ▶ highly correlated dimensions

$$\begin{pmatrix} \theta_1 \\ \theta_2 \end{pmatrix} \sim N \left( \begin{pmatrix} -1 \\ 1 \end{pmatrix}, \begin{pmatrix} 1 & 0.8 \\ 0.8 & 1 \end{pmatrix} \right)$$

- ▶ Between-item dimensionality
  - ▶ easy items measuring  $\theta_1$ , difficult items measuring  $\theta_2$
- Model fit

	AIC	BIC	logLik
UIRT	197711.2	198232.6	-98775.62
MIRT	195555.9	196083.8	-97696.95

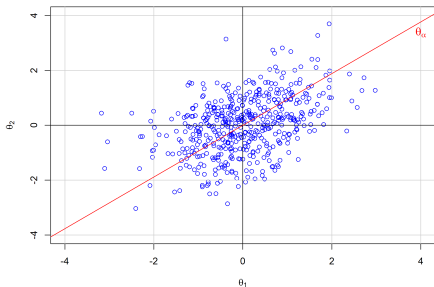
- Item-fit statistics can hardly detect any misfit when fitting UIRT model to multidimensional data with highly correlated dimensions.

# Fitting UIRT to Multidimensional Data



A long-standing conjecture: the fitted UIRT to multidimensional data represents a linear composite of the dimensions present in a test.

$$\theta_\alpha = w_1\theta_1 + w_2\theta_2$$



**Figure 1:** Illustration of a latent bivariate distribution for  $(\theta_1, \theta_2)$  with a corresponding linear composite direction denoted by  $\theta_\alpha$ , Strachan et al. (2022)



- Two-dimensions where dimensionality is related to item types

$$\begin{pmatrix} \theta_1 \\ \theta_2 \end{pmatrix} \sim N \left( \begin{pmatrix} -1 \\ 1 \end{pmatrix}, \begin{pmatrix} 1 & 0.8 \\ 0.8 & 1 \end{pmatrix} \right)$$

- ▶  $\theta_1$  on easy items:  $a \sim N(1.3, 0.2), b \sim N(-1, 1)$
- ▶  $\theta_2$  on difficult and discriminative items:  $a \sim N(2, 0.2), b \sim N(1, 1)$
- Calibrate the response data with UIRT model
- Estimate  $w_1$  and  $w_2$  by ability groups from separate latent regressions

$$\hat{\theta} = \hat{w}_1\theta_1 + \hat{w}_2\theta_2$$

# UIRT Approximation: Two-dimensional Example, Cont'd

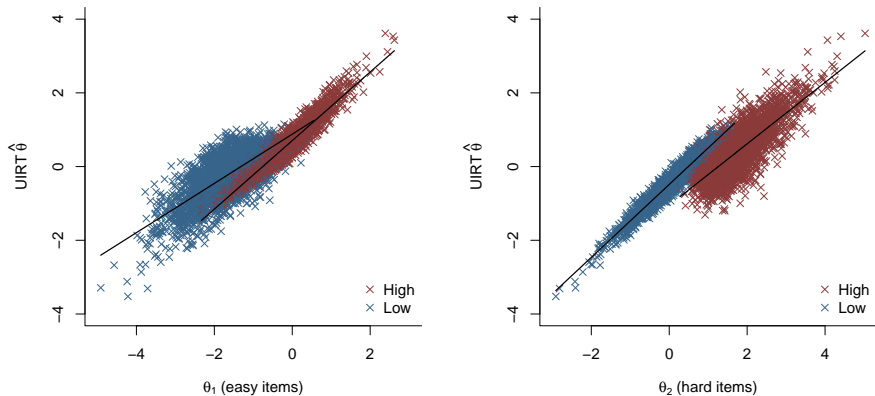


Figure 2: Illustration of the UIRT Approximation by Dimension, Two Group

# UIRT Approximation: Two-dimensional Example, Cont'd

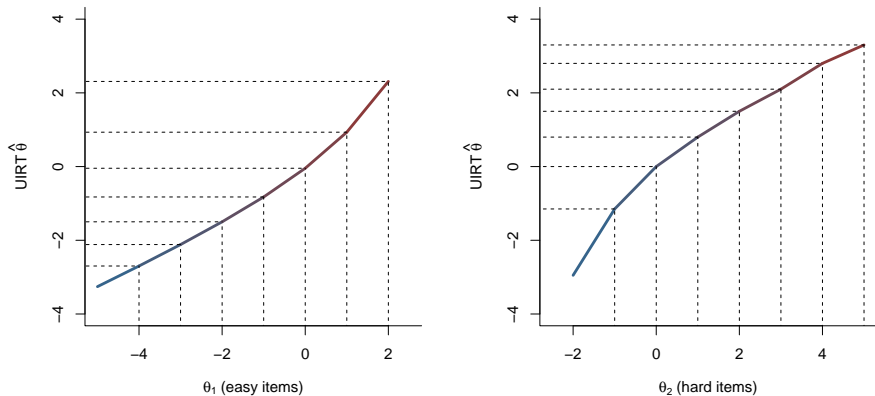
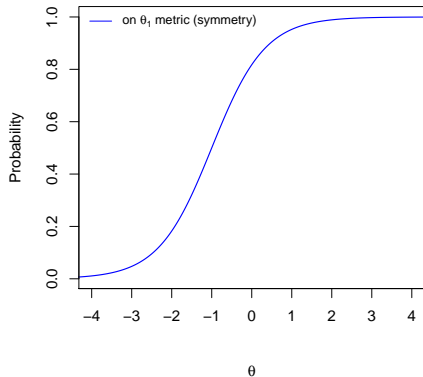


Figure 3: Illustration of the UIRT Approximation by Dimension



# UIRT Approximation: Two-dimensional Example, Cont'd

Example Easy Item



Example Hard Item

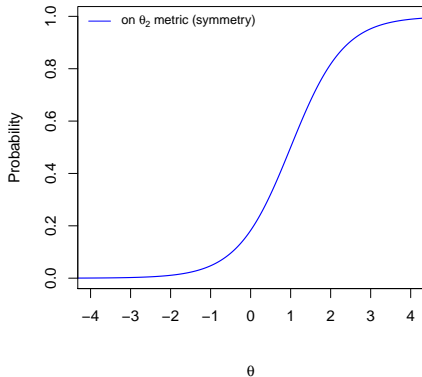
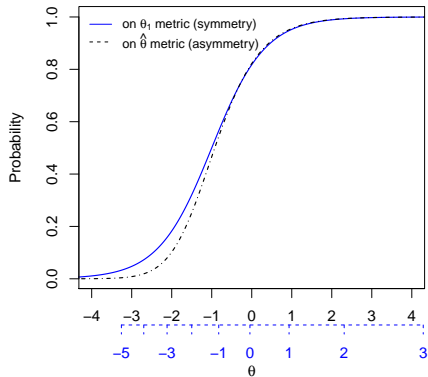


Figure 4: Illustration of the UIRT Metric Distortion in ICCs

# UIRT Approximation: Two-dimensional Example, Cont'd

Example Easy Item



Example Hard Item

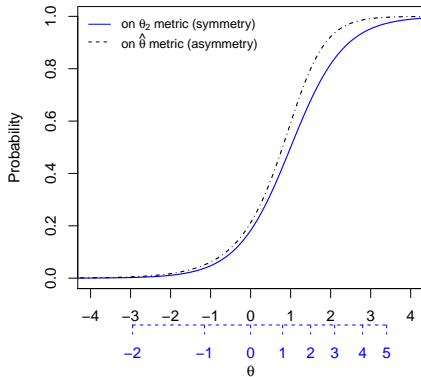


Figure 5: Illustration of the UIRT Metric Distortion in ICCs

# Consequences of UIRT Approximation: Matthew Effect

- Students who start lower on the metric may tend to be credited with lesser gains than students that start higher even if they grow equivalent amounts.

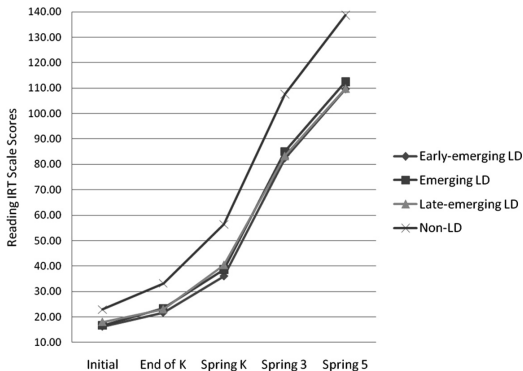


Figure 6: Group Differences in Reading Growth and Achievement over the First 6 Years of School, ECLS-K data, from Judge & Bell (2010)

# Consequences of UIRT Approximation: Vertical Linking

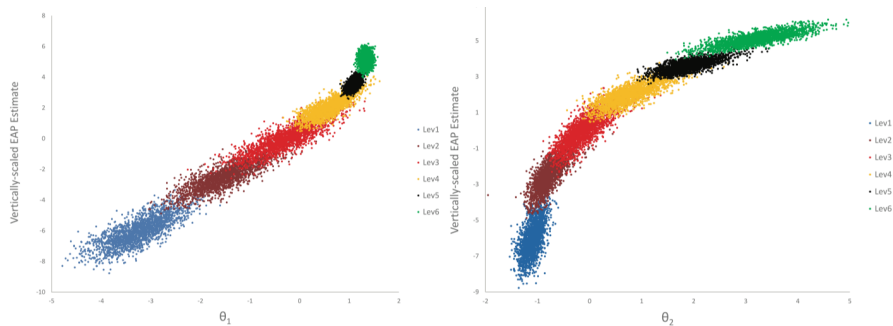


Figure 7: Relationships between Vertically Scaled EAP Estimates and  $\theta$ s, from Carlson (2017)



- UIRT  $\theta$  as a curvilinear approximation when dimensionality is related to item difficulty
- Interpretation of the UIRT  $\theta$ 

“The IRT scale scores may be used as longitudinal measures of overall growth. However, gains made at different points on the scale have qualitatively different interpretations. [...] Comparison of gain in scale score points is most meaningful for groups that started with similar initial status.” (Pollack et al., 2005)
- Selecting anchor items in vertical linking



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# Thank you!

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- Samejima's (2000) logistic positive exponent (LPE) model

$$P_{ij}(X_{ij} = 1|\theta_i; a_j, b_j, \xi_j) = \left( \frac{\exp(a_j(\theta_i - b_j))}{1 + \exp(a_j(\theta_i - b_j))} \right)^{\xi_j}$$

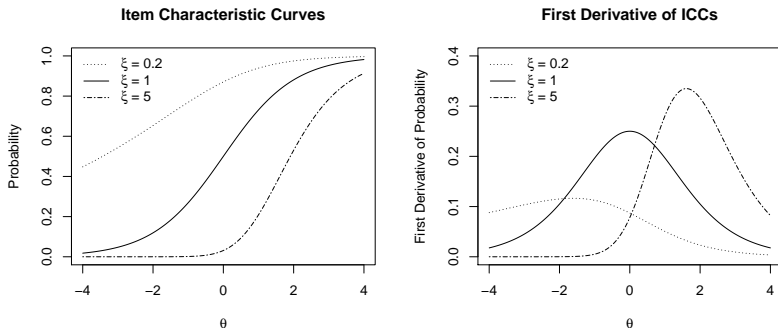


Figure 8: Example Item Characteristic Curves (ICCs) and their First Derivatives of LPE Items ( $a = 1, b = 0$  for all items).