

Item Quality for Cognitive Assessments in Low-to-Middle Income Countries: Evidence from the Ethiopia Young Lives Data.

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INTRODUCTION

Across the globe, testing agencies have a common goal of creating items that maintain measurement invariance and are thus free of group characteristic biases (Khorramdel et al., 2020). Even though test agencies' efforts to consult with local talents to create cultural and gender relevant items in LMIC's, there is evidence that some of these items are biased and may be unfair to some students based on their economic status, gender, and location. We examine the cultural and gender relevant items included on the cognitive assessments for the Young Lives school effectiveness survey data from Ethiopia (2016/2017) (Boyden et al., 2016; Rossiter et al, 2017).

AIM

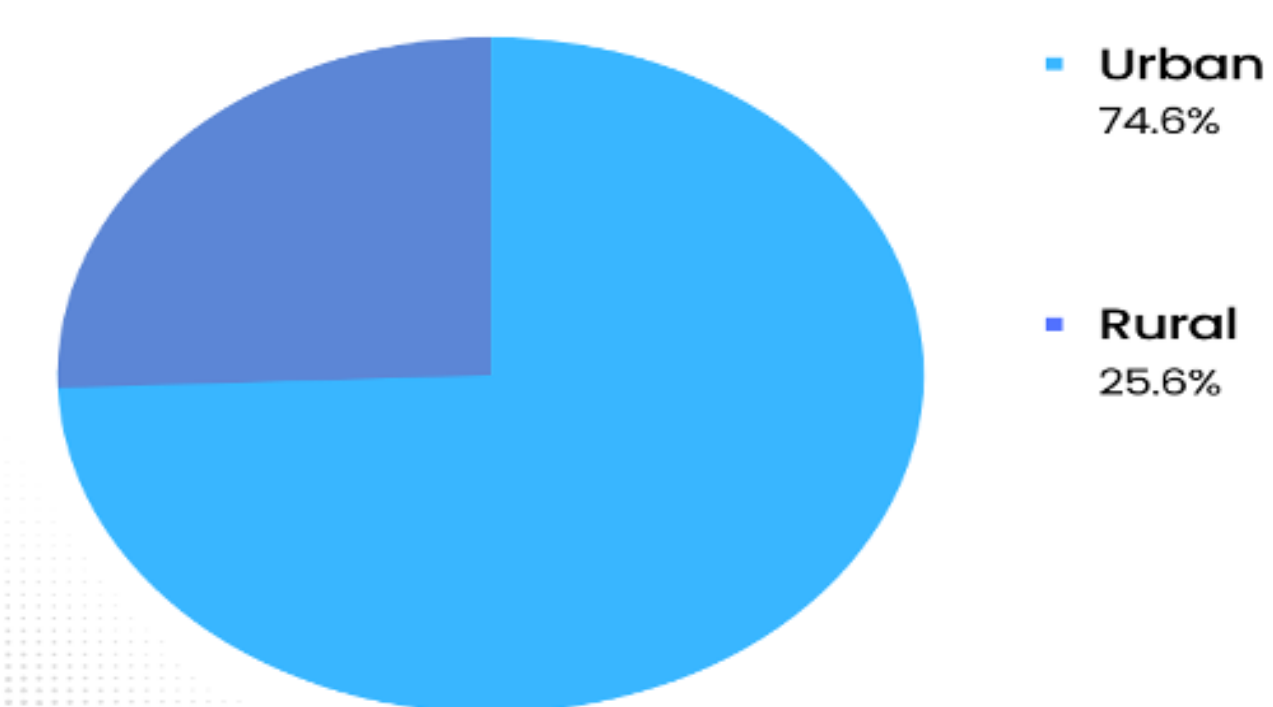
The purpose of this paper is to examine how gender and location (rural vs urban) may impact item quality for assessments administered in low-to-middle income countries (LMIC's). This study examines :

1. The item fit statistics and item targeting of the YL cognitive assessments for Ethiopia under the Rasch model.
2. Differential item functioning analysis (DIF) by gender and location for the cognitive assessment data.

Descriptive Statistics

- We use data from the first wave which was collected in 2016.
- The survey was completed by a total of $N = 12,182$ children in 7th and 8th grade (M age = 14.36) from 63 schools.
- About half the sample identified as boys (48.5 %) and the other half (48.9%) as girls, with 2.6% missing on this variable.

A PIE CHART SHOWING % URBAN VS RURAL



METHODS

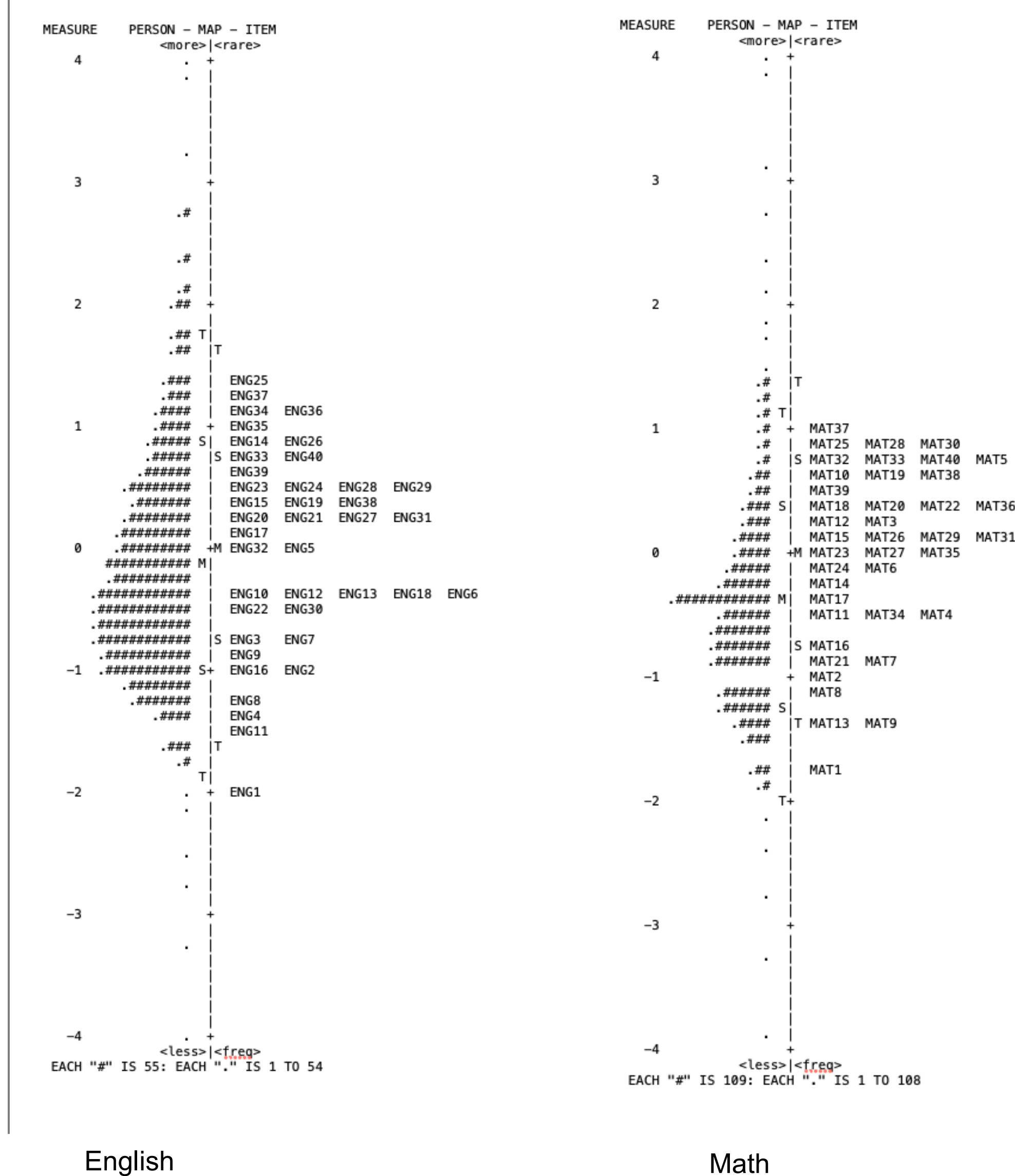
- ❖ Item responses to the cognitive assessments were analyzed using the **Rasch model**. This is consistent with the hypothesized unidimensional nature of this data (Rossiter et al., 2017; Linacre, 2022).
- ❖ Item fit statistics were examined to see how well the data fit the Rasch model, and Wright maps were utilized to clarify item targeting.
- ❖ We conducted differential item functioning (**DIF**) analysis to explore the possibility of item-level performance differences (Walker, 2011) based on gender and location.

RESULTS

Test-Item Targeting

The person-distributions for both the Math & English assessments showed the ability of students on the same measurement scale as the item difficulty; this pattern was acceptable but suggests that more difficult items should be added to the assessment for highly proficient students at the upper part of the scale.

Wright maps showing the person distribution with item labels



Differential Item Functioning

For both English and Mathematics items, we used the Rasch-Welsh Test and a significance level of 0.05 based on the absolute value of the DIF contrast (Linacre, 2022).

- Overall, there was no support for any individual items that favored boys or girls.
- However, for location (urban versus rural), **15%** of the mathematics items showed moderate to high DIF ($|DIF| \geq 0.64$ logits) and slight to moderate DIF ($|DIF| \geq 0.43$ logits), (Zwick, Thayer & Lewis, 1999).
- The English test had **17.5%** of items showing moderate to high DIF. For example, the first test item which asked students to identify a school bag, had a DIF contrast of .82. This item was harder for students in rural schools.

Examples of some English Items that exhibited location DIF

1 Select the correct word for the picture.



- A Bat
- B Bag
- C Pad
- D Box

8 Who among the following is most likely to use this object for their work?



- A Lawyer
- B Doctor
- C Farmer
- D Carpenter

A picture of a school in rural Ethiopia



Source: Ark Republic

CONCLUSIONS

Based on the literature from LMICs, most large-scale test developers seem to focus on creating unbiased questions to avoid gender DIF than they do for location. Thus, it is no surprise that this same trend is seen within these YL data where the gender-based differences were negligible in terms of DIF but potentially biased items are indicated when location is considered.

E-mail Wilberforce.2@osu.edu with suggestions.

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