Detecting Intersectional Differential Item Functioning: A Comparison of Two Methods

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Outline

- > Background
- > Methods
- > Empirical study
- > Discussions



Background

- > Differential item functioning (DIF)
- > Intersectionality
 - Acknowledges that most demographic factors do not function in isolation, but rather as interlinked phenomena that collectively shape complex social inequalities (Bauer et al., 2021; Grabe, 2020)
- > Intersectional DIF
 - Not only incorporates multiple demographic variables in a single model but emphasizes the DIF effect that reflects the interactions of various identities rather than the mere additive effect





Methods: Regularized DIF (reg-DIF) Method

$$P(y_{ij} \mid \theta_i) = \frac{1}{1 + \exp\left[-\left(\left(a_j + \boldsymbol{X}_i^{\mathrm{T}} \boldsymbol{\gamma}_j\right) \theta_i + d_j + \boldsymbol{X}_i^{\mathrm{T}} \boldsymbol{\beta}_j\right)\right]}$$

DIF parameters on which reg-DIF assign penalty

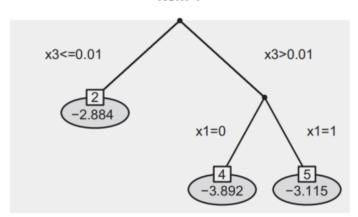
> The reg-DIF method incorporates regularization regression into the explanatory item response theory (EIRT).



Methods: Item-focused Tree (IFT) Method

$$\log \left[\frac{P(y_{ij} = 1 | S_i, \mathbf{X}_i)}{P(y_{ij} = 0 | S_i, \mathbf{X}_i)} \right] = \left[\sum_{k=1}^{K} \gamma_{jk} node_{jk}^{[U]}(\mathbf{X}_i) \right] S_i + \sum_{k=1}^{K} \beta_{jk} node_{jk}^{[NU]}(\mathbf{X}_i)$$
Non-Uniform DIF
Uniform DIF

Item 1



> The IFT method integrates the tree structure—binary recursive partitioning of the feature space—with logistic regression.



Intersectional DIF with reg-DIF and IFT

> reg-DIF: All the variables of interest are synthesized into a single grouping variable, within which each category corresponds to a unique combination of different demographic variables. Dummy coding is then applied to the synthesized variable and incorporated into the model.

> IFT: It can directly use the original demographic variables. The intersectional DIF is represented by the sequentially split nodes in the model.



Empirical Study

- > Data is from the SCS1 assessment, an exam widely used for the undergraduate introductory programming course in computer science
- > 259 students completed the test, and their demographics are as below

Table 1. Demographics for the 259 students that completed the relevant questions when attached to the end of SCS1.

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Gender	n	Race/Ethnicity	n	Primary Language	n		
Female	96	White/Caucasian	104	English	160		
Male	159	Asian/Pacific	117	Chinese	42		
		Islander					
Other	4	Black	3	Korean	17		
		Hispanic	6	Vietnamese	7		
		Multiple	29	Spanish	4		
		_		Other/(No answer)	29		



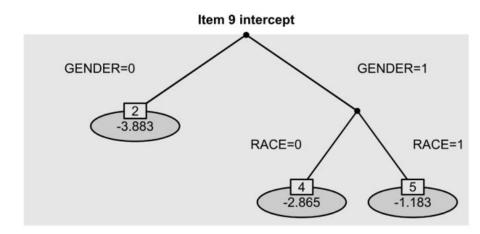
Results

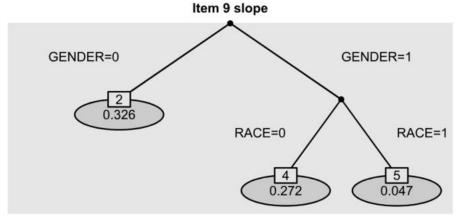
Table 2. Intersectional DIF magnitude estimates.

		U/NUDIF		
Item with DIF	reg-DIF WMref	reg-DIF NFref	IFT	IFT
Item 8	0.803 ^(NM)	0.801 ^(NM)	-	-
Item 9	-	-	-	#
Item 13	$0.821^{(NM)}$	$0.820^{(NM)}$	-	-
Item 20	-	$0.793^{(WM)}$	-	-

Note: reg-DIF WMref refers to the reg-DIF method with white male as the reference, and reg-DIF NFref refers to the reg-DIF method with non-white female as the reference. The superscript on the magnitude denotes the focal group inducing DIF effect, with NM representing non-white male and WM denoting white male. # refers to the item detected with U/NUDIF, and it is illustrated in Figure 1.

Figure 1. Tree structure of item 9 under U/NUDIF detection by gender and race.





Note: GENDER=0 refers to males, GENDER=1 refers to females, RACE=0 refers to the white, RACE=1 refers to the non-white.



Thanks! heren@uw.edu

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