

# Pauci sed moni: An Item Response Theory approach for shortening tests

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#### Introduction

Item Response Theory (IRT) is the theoretical framework often used for shortening existing tests. IRT models describe the probability of observing a response as a function of the characteristics of respondent p (i.e., the latent trait level  $\theta$ ) and the characteristics of item s. IRT models provide detailed information on how well each item measures a certain  $\theta$  level (i.e., *item information function*, *IIF*). Two types of short forms can be created by exploiting the *IIF*s:

1. **Adaptive short forms**: Ad-hoc tests for each person (i.e., Computerized Adaptive Testing, CAT. The items administered to each respondent vary according to the responses that this respondent gave to the previously administered items)  $\rightarrow$  the information is maximized for each level of  $\theta$  (i.e., each respondent)

**Issue**: Different short test forms for each respondent  $\rightarrow$  Unfair assessments in recruitment or admissions tests

2. **Static short forms**: Static tests equal for all respondents (i.e., only the items from the full-length test that provide the highest information are included in the short form)  $\rightarrow$  the information is maximized across  $\theta$  levels (i.e., across all respondents)

**Issue**: Not being tailored to any  $\theta$  level of interest  $\rightarrow$  Potentially more items are needed to cover a wide range of  $\theta$ s

#### Aim

New IRT-based procedures for the development of short test forms combining the advantages of adaptive short test forms (i.e., tailoring the tests to different  $\theta$  levels) and those of static short forms (i.e., being equal for all respondents).

The new item selection procedures are based on the definition of trait levels of interest (i.e.,  $\theta$  targets, denoted as  $\theta'$ )  $\to$  The items that best assess the trait levels represented by the  $\theta'$  targets (i.e., optimal items with highest *IIF*s for each  $\theta'$ ) are included in the short form.

## Item Response Theory and information functions

This illustration is based on the 2-parameter logistic model (2PL) for dichotomous responses:

$$P(x_{ps} = 1 | \theta_p, b_s, a_s) = \frac{exp[a_s(\theta_p - b_s)]}{1 + exp[a_s(\theta_p - b_s)]}$$
(1)

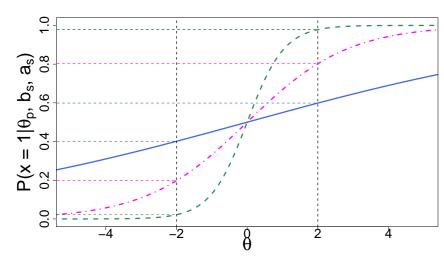
where  $P(x_{ps} = 1 | \theta_p, b_s, a_s)$  is the probability of respondent p to respond correctly to item s given the ability  $(\theta)$  of p and difficulty (b) and discrimination (a) of s. The *Item Characteristics Curves* (*ICCs*) of three items with same difficulty but different discriminations are illustrated in Figure 1a.

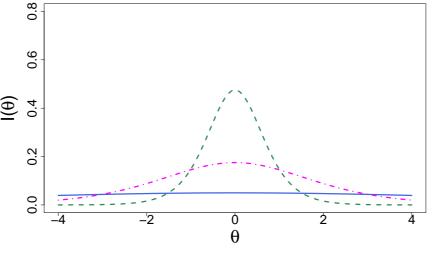
The *item information function* (*IIF*) informs about the precision with which the item measures the abilities  $\theta$ s. In the 2PL model, the *IIF* is obtained as:

$$IIF = a^{2}[P(\theta)(1 - P(\theta))], \tag{2}$$

where  $P(\theta)$  is the probability of a respondent with a certain  $\theta$  of responding correctly to an item, and  $1 - P(\theta)$  is their probability of responding incorrectly to the same item. The *IIF*s of the items depicted in Figure 1a are illustrated in Figure 1b.

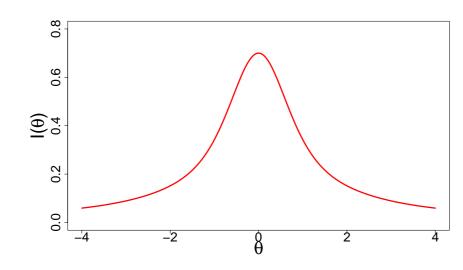
The test information function (TIF) is obtained by summing the IIFs across items (test information function,  $TIF = \sum_{s=1}^{S} IIF_s$ , Figure 1c).





(a) Item Characteristics Curves (*ICC*s) of items with b = 0, and a = 0.20, a = 0.70, a = 1.90

(b) Item Information Functions (*IIF*s) of items with b=0, and a=0.20, a=0.70, a=1.90



(c) Test Information Function (*TIF*) of the test composed of the items in Fig. 1a and Fig. 1b.

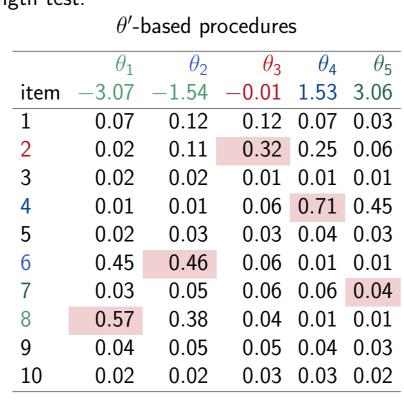
Figure 1: 2-PL and information functions

## Item Selection Procedures

- ▶ **Benchmark**: The *N* items with the highest *IIF*s are selected from the full-length test to be included in the static short form, where *N* is the desired length of the short form (Benchmark Procedure, BP).
- ▶ Procedures based on  $\theta'$ :
- ▶ Cluster: The latent trait is clustered in N clusters, where n is the number of items to be included in the short form. The centroids of the clusters are the  $\theta'$  (Unequal Intervals Procedure, UIP).
- ▶ Intervals: The latent trait is segmented into N+1 intervals. Each interval is defined by  $[\theta'_{n-1}; \theta'_n]$ . The  $\theta'$ s are obtained by averaging between the lower and upper bound of each interval to avoid that the first and the last  $\theta'$ s correspond to the minimum and maximum  $\theta$  values (Equal Intervals Procedure, EIP).
- ▶ Random: Items are randomly selected from the full-length tests (RP).

Development of a 5-item short form from a 10-item full-length test:

| Typical procedure |       |      |      |
|-------------------|-------|------|------|
| item              | Ь     | а    | IIF  |
| 1                 | -2.51 | 1.68 | 0.10 |
| 2                 | -2.43 | 0.25 | 0.02 |
| 3                 | -2.28 | 1.62 | 0.13 |
| 4                 | -0.67 | 0.71 | 0.11 |
| 5                 | -0.66 | 0.44 | 0.05 |
| 6                 | 0.50  | 1.19 | 0.27 |
| 7                 | 0.64  | 0.50 | 0.06 |
| 8                 | 0.72  | 0.33 | 0.03 |
| 9                 | 1.72  | 0.39 | 0.03 |
| 10                | 2.12  | 1.98 | 0.16 |



### Method

Comparison between the item selection procedures:

- ► Benchmark procedure (BP)
- ► Unequal Intervals Procedure (UIP)
- ► Equal Interval Procedure (EIP)
- ► Random Procedure (RP)

in the development of 10, 30, 50, 70, 90 items test short forms from a 100-item full-length test (For each short test form, RP randomly selects the items 10 times).

## 1000 respondents p

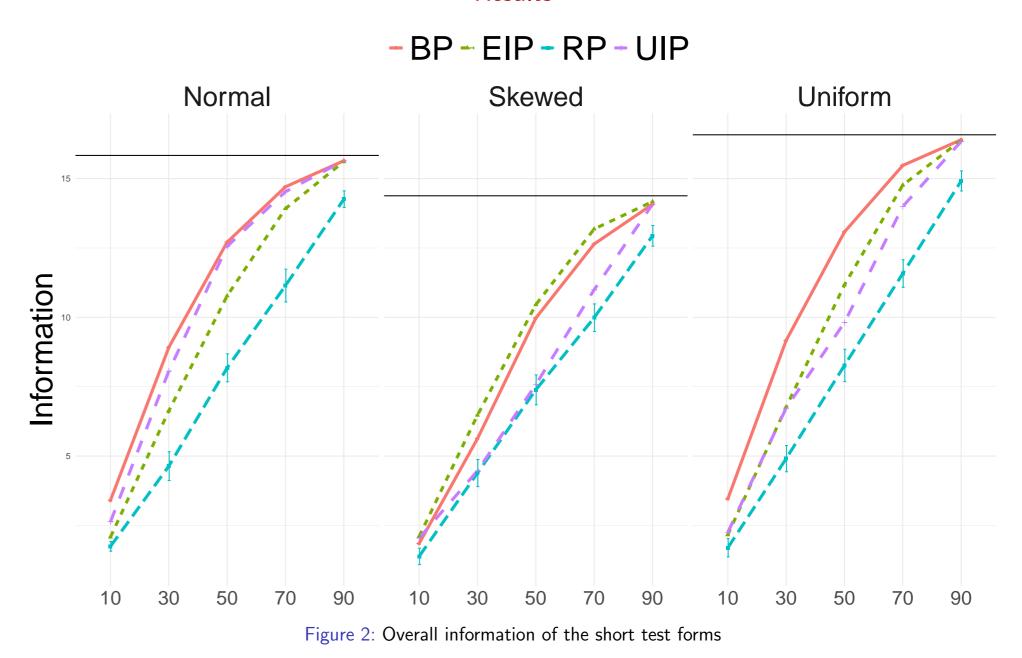
100 items *s*:

Three  $\theta$  distributions:

▶  $b \sim \mathcal{U}(-3,3)$ 

- **1**. Normal distribution  $p \sim \mathcal{N}(0,1)$
- ightharpoonup  $a \sim \mathcal{U}(0.40, 2)$
- 2. Positive skewed distribution  $p \sim Beta(1, 100)$  (linearly transformed to obtain negative values)
- 3. Uniform distribution  $p \sim \mathcal{U}(-3,3)$

## Results



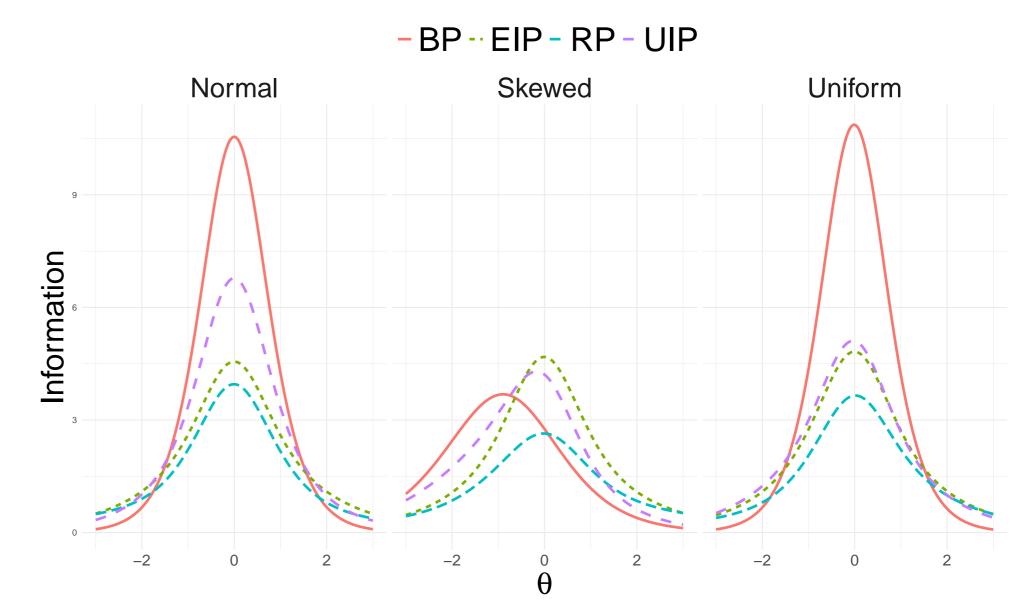


Figure 3: Detailed information of the short test forms

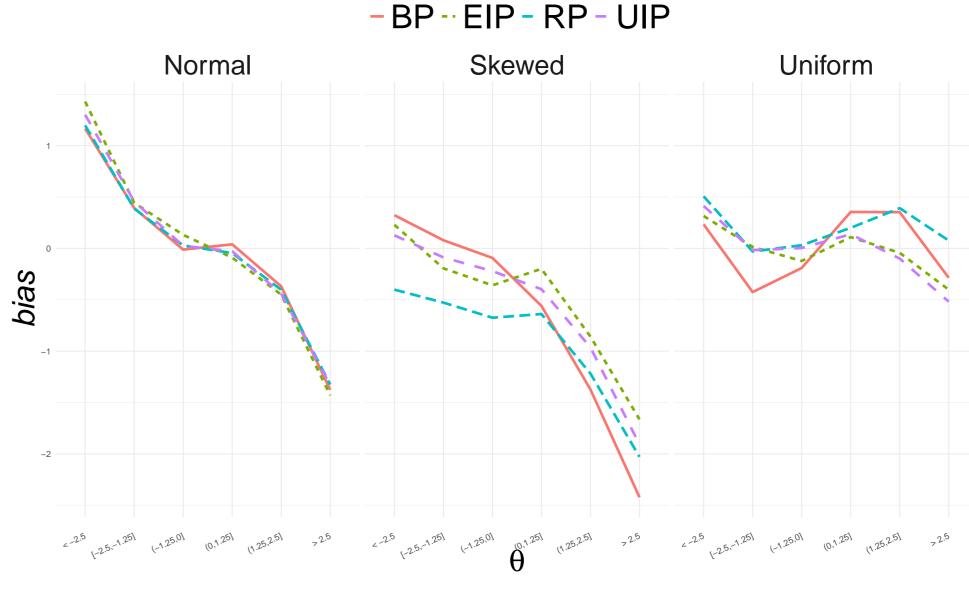


Figure 4: Bias for different group of  $\theta$ 

## Discussion

- ightharpoonup Different methods for different heta distributions
- lacktriangle Better performance of heta-based procedures on the extreme ends of the distributions
- lackbox By considering the heta' in the item selection procedures o not the highest information but best coverage of the entire latent trait