



# **Look-ahead content balancing method in variable-length computerized classification testing**

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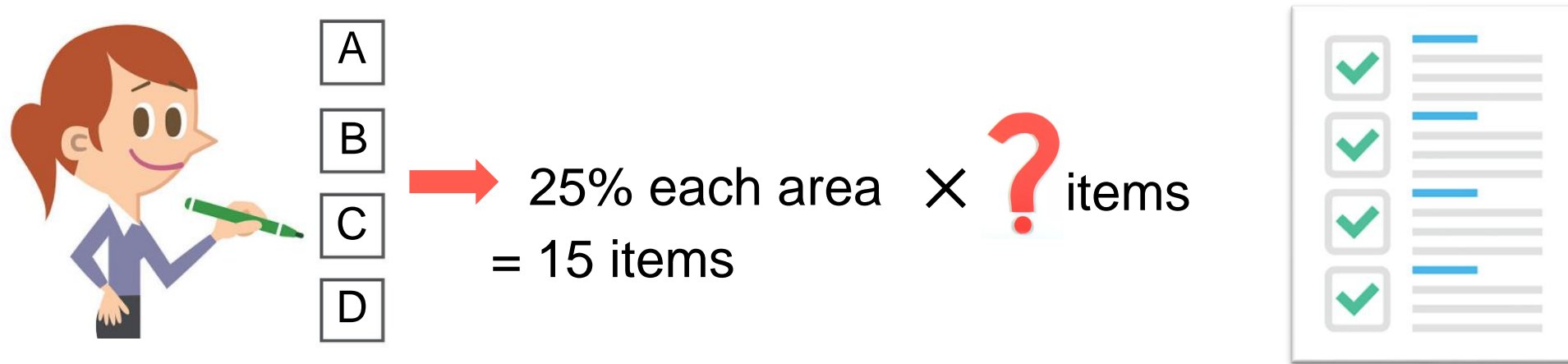
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# Introduction

- Computerized classification testing (CCT)
  - **statistical constraints:** max FI at current  $\hat{\theta}$  ...
  - **non-statistical constraints:** content balancing, exposure control ...



variable-length computerized classification testing (VL-CCT)



proposing **two feasible methods** gains control over content coverage in **VL-CCT** programs

- the constraint CAT method (Kingsbury & Weiss, 1983)
- the modified multinomial model method (Chen & Ankenman, 2004)
- the modified constraint CAT method (Leung, Chang, & Hau, 2000)
- the maximum priority index (MPI) (Cheng & Chang, 2009)
- the content-weighted item selection index (CWI) (Huo, 2009)

## How to assemble a test?

- Maximum Fisher information method

$$I_j(\theta) = \frac{(1 - c_j)a_j^2 e^{a_j(\theta - b_j)}}{[1 + e^{a_j(\theta - b_j)}]^2 \{1 - c_j + c_j[1 + e^{a_j(\theta - b_j)}]\}}$$

- Maximum priority index

$I_j(\theta) \times$  items' contribution towards meeting constraints

$$PI_j = I_j \prod_{k=1}^K (\omega_k f_k)^{c_{jk}}$$

# Methods

- Maximum priority index (MPI)

$$PI_j = I_j \prod_{k=1}^K (\omega_k f_k)^{c_{jk}}$$

$C =$

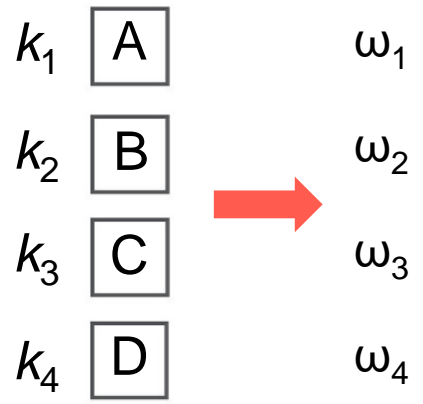
	$k_1$	$k_2$	$k_3$	$k_4$	
	0	1	0	0	item 1
	0	0	1	0	item 2
	1	0	0	0	item 3
	...	...	...	...	...
	0	0	0	1	item J

$$f_k = \frac{X_k - x_k}{X_k}$$

$$= \frac{15 - (\text{items have been selected})}{15}$$

- especially for exposure control

$$f_{kr} = \frac{r - n/N}{r}$$



~~25% each area~~

?  $l_k$   $\leq \mu_k \leq u_k$

**– first phase**

$$f_k = \frac{(l_k - x_k)}{l_k}$$

↓

all the lower bounds will be met at the end of the first phase

**– second phase**

$$f_k = \frac{(u_k - x_k)}{u_k}$$

- Content-weighted item selection index (CWI)
  - two-stage a-stratified method

1. first phase: fixed-length testing course

$$\min|\hat{\theta} - b| \text{ until min test length} \quad \underline{CWI_1} = \frac{l_k}{l_k - x_k + 1} |\hat{\theta} - b|$$

2. second phase: variable-length testing part

$$\min|\hat{\theta} - b| \text{ until threshold} \quad CWI_2 = \frac{u_k}{u_k - x_k + 1} |\hat{\theta} - b|$$

&

circularly increasing / decreasing

strata1 < strata2 < strata3 < strata4

Ca → item1    item2    item3    item4  
Cd → item4    item3    item2    item1

# Methods

- MPI and CWI
  - upper bound

$$u_k = U \times u_k\%$$

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	test	content area 1 (20%)	content area 2 (20%)	content area 3 (30%)	content area 4 (30%)
Lower bound	20	4	4	6	6
Upper bound	40	8	8	12	12

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↓  
can be much larger than the actual ones

➔ “step size” + existing number of selected items

1. constant  $S$
2. adaptive  $S$

- Look-ahead content balancing (LA-CB): based on MPI

– first phase  $f_k = \frac{l_k - x_k}{l_k}$

– second phase  $f_k = \frac{(u_k - x_k)}{u_k}$

➔ the total test length (TL)  $\leq$  the maximum test length ( $U$ )

**What would be the remaining length?**

$$x_k + S \times u_k \% \leq U \times u_k \%$$

$$1 \leq S \leq U - \sum_{k=1}^K x_k$$

- LA-CB-C: with constant step size

$S^{\text{constant}}$  ➔ a constant integer within the range

**How to be adaptive?**



# Methods

- LA-CB-A: with adaptive step size

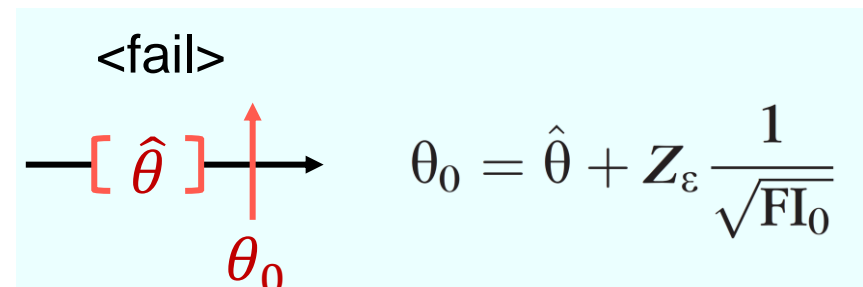
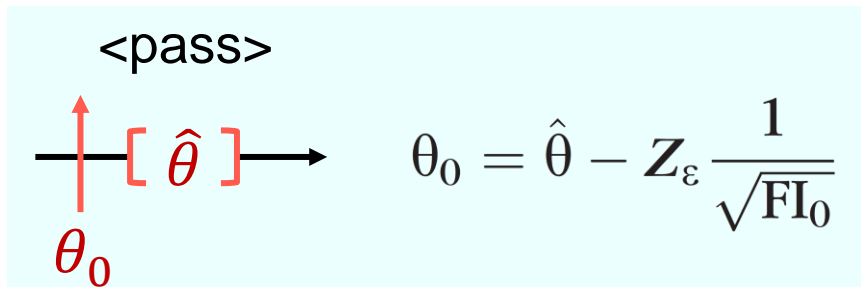
$$1 \leq S \leq \textcircled{U} - \sum_{k=1}^K x_k \quad ?$$

– inspired by the ability confidence interval (ACI)

→  $\hat{\theta} - Z_\alpha \times \text{SEM} < \theta < \hat{\theta} + Z_\alpha \times \text{SEM} \rightarrow SD(\hat{\theta}) \rightarrow \frac{1}{\sqrt{\sum_{j=1}^n I_j(\theta)}}$

the normal deviate for a 100(1 - α)% CI

→  $\hat{\theta} - Z_\alpha \frac{1}{\sqrt{\sum_{j=1}^k I_j(\theta)}} < \theta < \hat{\theta} + Z_\alpha \frac{1}{\sqrt{\sum_{j=1}^k I_j(\theta)}}$



- LA-CB-A

$$\theta_0 = \hat{\theta} \pm Z_\varepsilon \frac{1}{\sqrt{FI_0}}$$

$$\theta_0 - \hat{\theta} = \pm Z_\varepsilon \frac{1}{\sqrt{FI_0}}$$

$$[\theta_0 - \hat{\theta}]^2 = \left[ Z_\varepsilon \frac{1}{\sqrt{FI_0}} \right]^2$$

$$FI_0 = \left[ \frac{Z_\varepsilon}{\theta_0 - \hat{\theta}} \right]^2$$

$$\frac{FI_0 - \sum_{j=1}^k I_j(\hat{\theta})}{\max(I_{\text{unselected}})} < \text{no. of items} < \frac{FI_0 - \sum_{j=1}^k I_j(\hat{\theta})}{\min(I_{\text{unselected}})}$$

as the look-ahead upper bound  $U$

$$S_0^{\text{adaptive}} = \frac{FI_0 - \sum_{j=1}^k I_j(\hat{\theta})}{\max(I_{\text{unselected}})} - \sum_{k=1}^K x_k$$

in a relatively early stage, not yet close to  $FI_0$

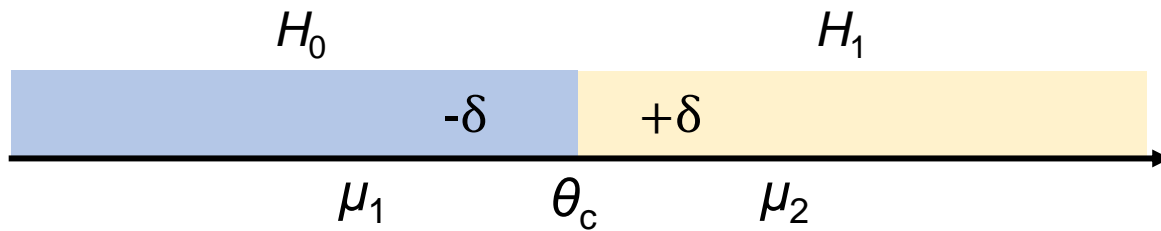
$$S^{\text{adaptive}} = \min\{S_0^{\text{adaptive}}, S^{\text{constant}}\}$$

the remaining length!

$$\rightarrow f_k = \frac{(u_k - x_k)}{u_k} = \frac{S \times u_k \%}{x_k + S \times u_k \%} \rightarrow \begin{matrix} S^{\text{constant}} \\ S^{\text{adaptive}} \end{matrix}$$

## How to stop?

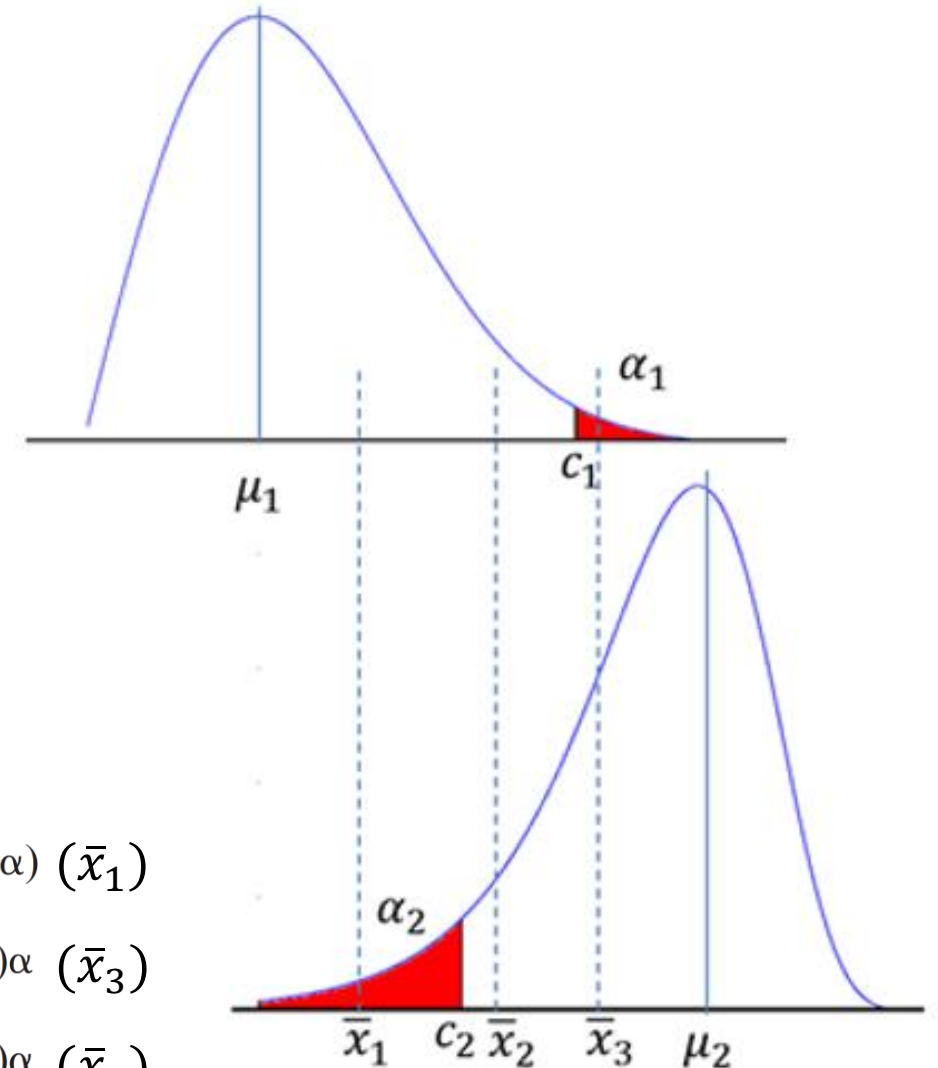
- The sequential probability ratio test (SPRT)



$$L(\theta; \mathbf{x}) = \prod_{i=1}^k p_i(\theta)^{x_i} [1 - p_i(\theta)]^{1-x_i}$$

$$LR(\theta_c + \delta; \theta_c - \delta) = \frac{L(\theta_c + \delta; \mathbf{x})}{L(\theta_c - \delta; \mathbf{x})}$$

- ability below the cutting point if  $LR(\theta_c + \delta; \theta_c - \delta) \leq \beta(1 - \alpha)$  ( $\bar{x}_1$ )
- ability above the cutting point if  $LR(\theta_c + \delta; \theta_c - \delta) \geq (1 - \beta)\alpha$  ( $\bar{x}_3$ )
- administer another item if  $\beta(1 - \alpha) < LR(\theta_c + \delta; \theta_c - \delta) < (1 - \beta)\alpha$  ( $\bar{x}_2$ )



- Ability confidence interval (ACI)

$$\hat{\theta} - Z_{\alpha} \times \text{SEM} < \theta < \hat{\theta} + Z_{\alpha} \times \text{SEM}$$

the normal deviate for a 95% CI

$SD(\hat{\theta}) \rightarrow \frac{1}{\sqrt{\sum_{j=1}^n I_j(\theta)}}$

– 95% CI is above the cut-off score  $\theta_0$

– 95% CI is below the cut-off score  $\theta_0$

– 95% CI is equal to or within the cut-off score  $\theta_0$

- Study 1.  
to choose a preferable classification method (**ACI or SPRT** methods) with LA-CB-C
- Study 2.  
whether the **LA-CB-C** method controls content constraints better than the **existing MPI and CWI** methods
- Study 3.  
whether the **LA-CB-A** method further improves the content balancing performance on top of the **LA-CB-C** method

- Data generation
  - Item pool structure
    - $a \rightarrow 0.5, 1.0, 1.5, 2.0$
    - $b \sim N(0, 1)$
    - $c \sim U(0, 0.25)$
    - 3PLM with 400 items
    - 4 content areas, 100 items/area, 25%
    - weights are all = 10
    - [minimum, maximum] test lengths = [28, 60]
    - constraint  $k = 1, 2, 3, 4$  (bounded = [7, 15])
    - exposure rate  $< 0.2$
    - weight = 100
  - Examinee generation
    - 2000 examinees with  $\theta \sim N(0, 1)$

- Data generation

- Model settings

- $\delta = 0.2, \alpha = \beta = 0.05$

- cut-off score  $\theta_0 = 0$

- step size  $S = [3, 20]$ , i.e., 18 integral values

$$S^{\text{actual}} = \max \left\{ 1, \min \left\{ S, U - \sum_{k=1}^K x_k \right\} \right\} \rightarrow \begin{array}{l} \text{LA-CB-C: } S = S^{\text{constant}} \\ \text{LA-CB-A: } S = S^{\text{adaptive}} \end{array}$$

- the first three items are always selected randomly

- ➔ selected from the two best items:

- maximized priority index / Fisher information

- minimized weighted index ...

- Evaluation criteria

## 1. Classification accuracy

- Classification error rate (CER)
- Type I error rate (Type I ER)
- Type II error rate (Type II ER)
- Mean square error:

$$\text{MSE} = \frac{\sum_{i=1}^N (\hat{\theta}_i - \theta_i)^2}{N}$$

## 2. Content balancing

- The average of a test:

$$\bar{V} = \frac{\sum_{i=1}^N V_i}{N}$$

- The average of a step size:

$$\text{Average } \bar{V} = \frac{\sum_{p=P_0}^P \bar{V}_p}{P - P_0 + 1}$$

## 3. Exposure control

- The maximum item exposure rate
- The proportion of over-exposed items (exposure rate > 0.2)
- The proportion of unused items
- Observed vs. expected exposure rates (ER):

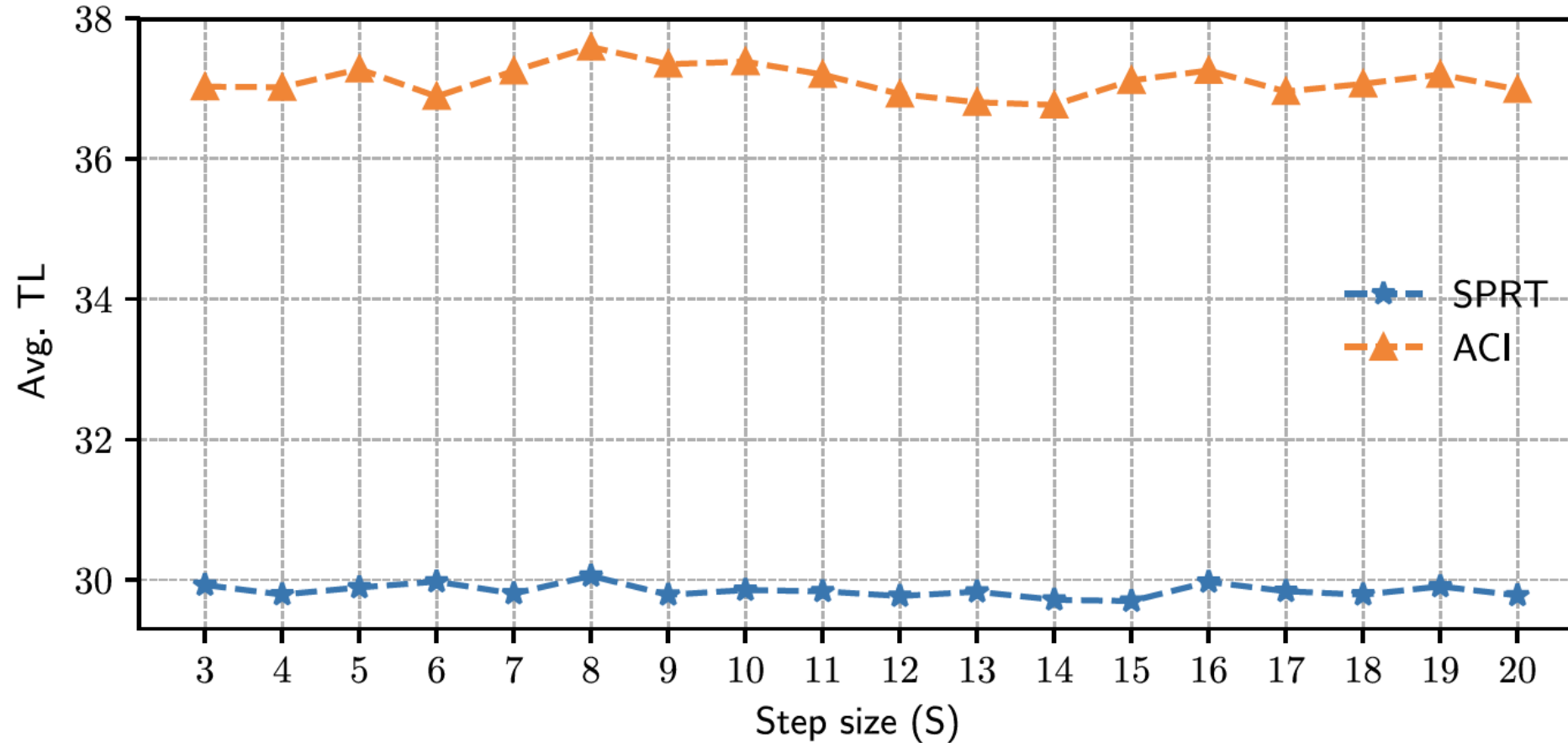
$$\chi^2 = \sum_{j=1}^K \frac{(\text{ER}_j - \overline{\text{ER}})^2}{\overline{\text{ER}}}$$

## 4. Test efficiency

- The average across various examinees:

$$\overline{\text{TL}} = \frac{\sum_{i=1}^N \text{TL}_i}{N}$$



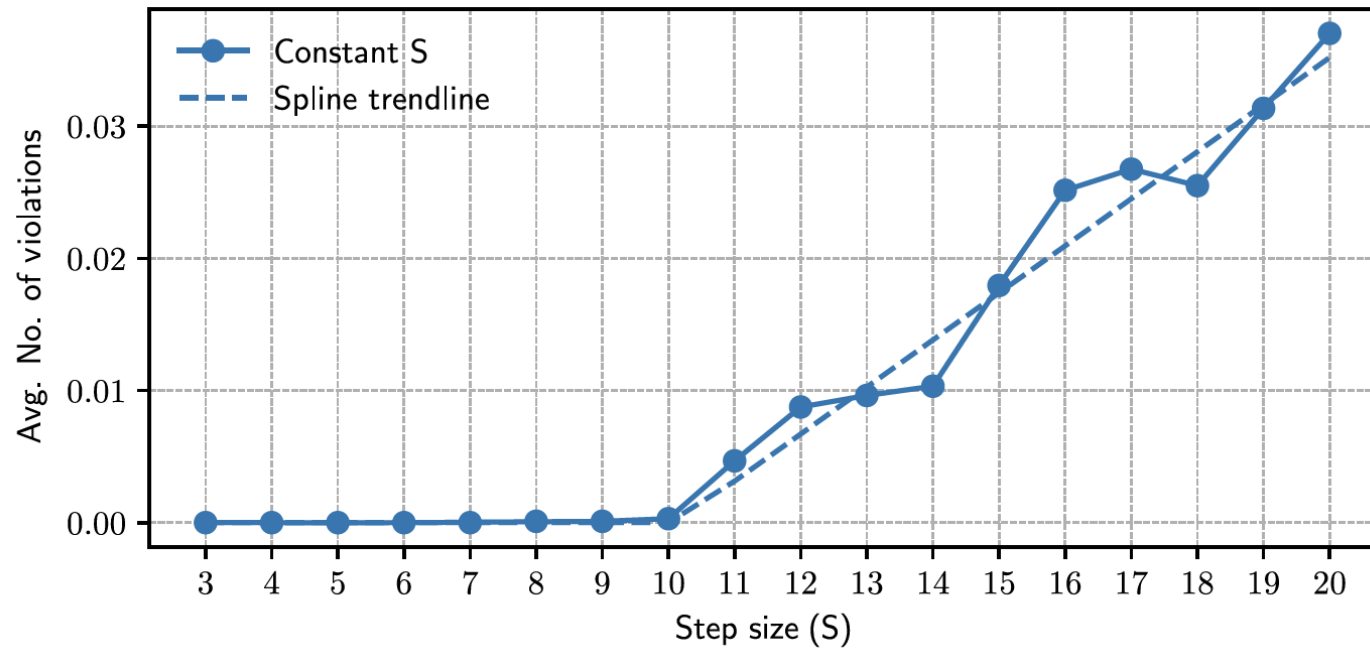
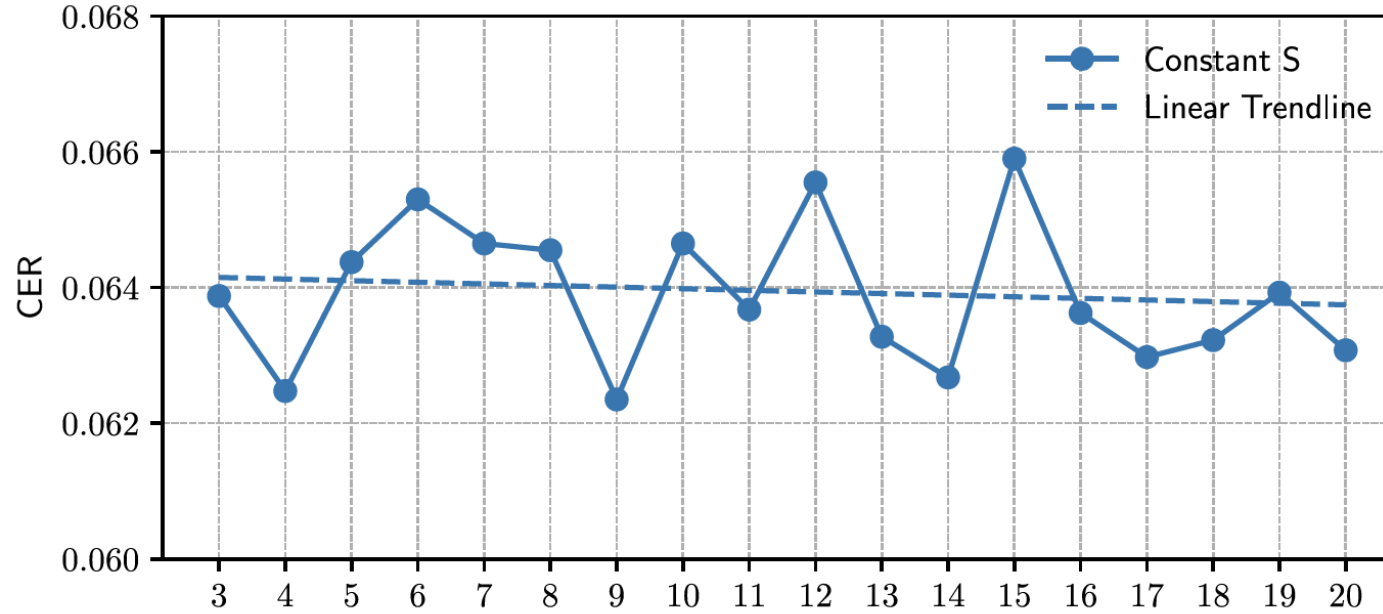


**Figure 1.** Average test length (TL) of sequential probability ratio test (SPRT) and ability confidence interval (ACI). [Colour figure can be viewed at [wileyonlinelibrary.com](http://wileyonlinelibrary.com)]

**Table 1.** Overall performance of sequential probability ratio test (SPRT) and ability confidence interval (ACI) classification methods

Methods	SPRT	ACI
Avg. test length TL	29.85	37.11
Grand avg. violated constraints $\bar{V}$	0.011	0.017
Average classification error rate	0.063	0.060
Average Type I error rate	0.033	0.030
Average Type II error rate	0.030	0.030

# Results – Study 2



# Results – Study 2

**Table 2.** Classification error rates (ER) and mean square error (MSE) of the LA-CB-C method and three other methods for 18 step sizes  $S$

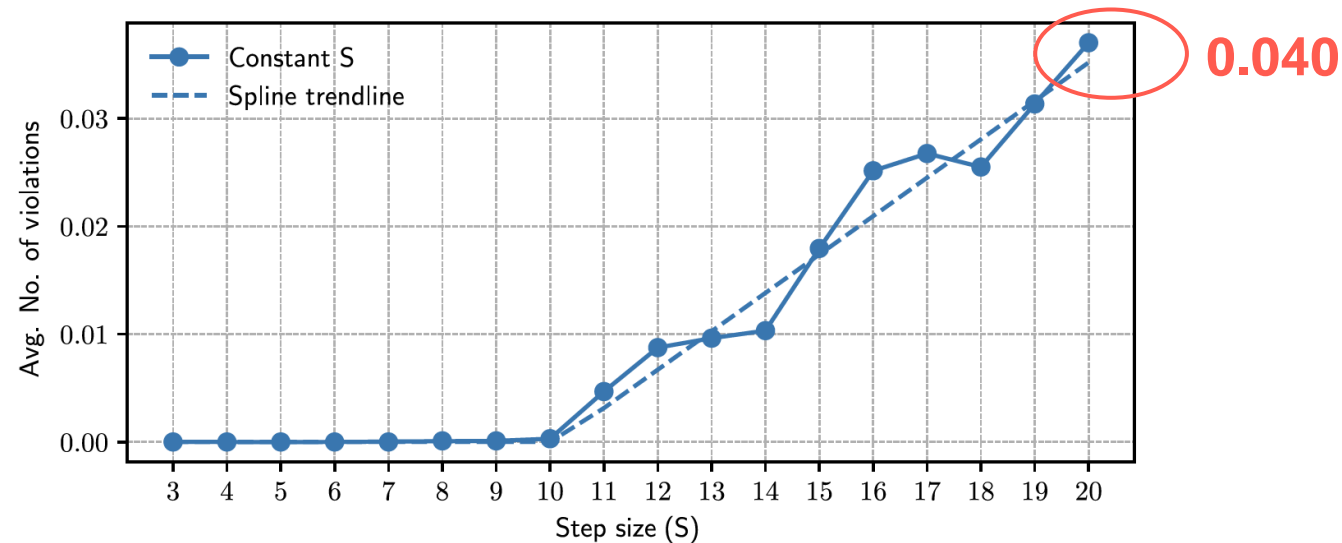
$S$	CER	Type I ER	Type II ER	MSE
3	0.064	0.034	0.030	0.074
4	0.062	0.033	0.030	0.074
5	0.064	0.033	0.031	0.075
6	0.065	0.034	0.031	0.075
7	0.065	0.033	0.031	0.075
8	0.065	0.033	0.032	0.074
9	0.062	0.033	0.030	0.075
10	0.065	0.034	0.031	0.076
11	0.064	0.032	0.032	0.075
12	0.066	0.034	0.031	0.075
13	0.063	0.033	0.030	0.075
14	0.063	0.033	0.030	0.075
15	0.066	0.034	0.032	0.075
16	0.064	0.033	0.031	0.074
17	0.063	0.033	0.030	0.075
18	0.063	0.033	0.030	0.075
19	0.064	0.032	0.032	0.075
20	0.063	0.033	0.030	0.075
LA-CB-C average	0.064	0.033	0.031	0.075
Maximum priority	0.062	0.032	0.030	0.074
Content-weighted	0.078	0.040	0.039	0.118
Maximum information	0.054	0.032	0.022	0.054
Randomized	0.084	0.042	0.042	0.254

# Results – Study 2

**Table 5.** Summary of content constraint violations ( $\bar{V}$ )

Measures	Average $\bar{V}$	Max $\bar{V}$	Min $\bar{V}$
LA-CB-C	0.0110	0.0370	0
Maximum priority*	0.0540	–	–
Content-weighted*	2.2295	–	–
Maximum information*	8.1380	–	–
Randomized*	7.0230	–	–

*Note.* This table summarizes the statistics of  $\bar{V}$  for 18 step sizes. Methods with (\*) do not include step sizes to make item selections and therefore maximum and minimum  $\bar{V}$  are not applicable.



**Table 3.** Overall exposure control indices

Methods	LA-CB-C	Maximum priority	Content-weighted	Maximum information	Randomized
Max. exposure rate	0.178	0.175	0.166	0.532	0.100
Over-exposed (%)	0	0	0	3.2	0
Never exposed (%)	0	0	0	0	0
$\chi^2$	20.297	20.228	4.009	83.041	0.153

# Results – Study 3

**Table 4.** LA-CB-A classification error rates (ER) and mean square error (MSE) for different step sizes  $S$

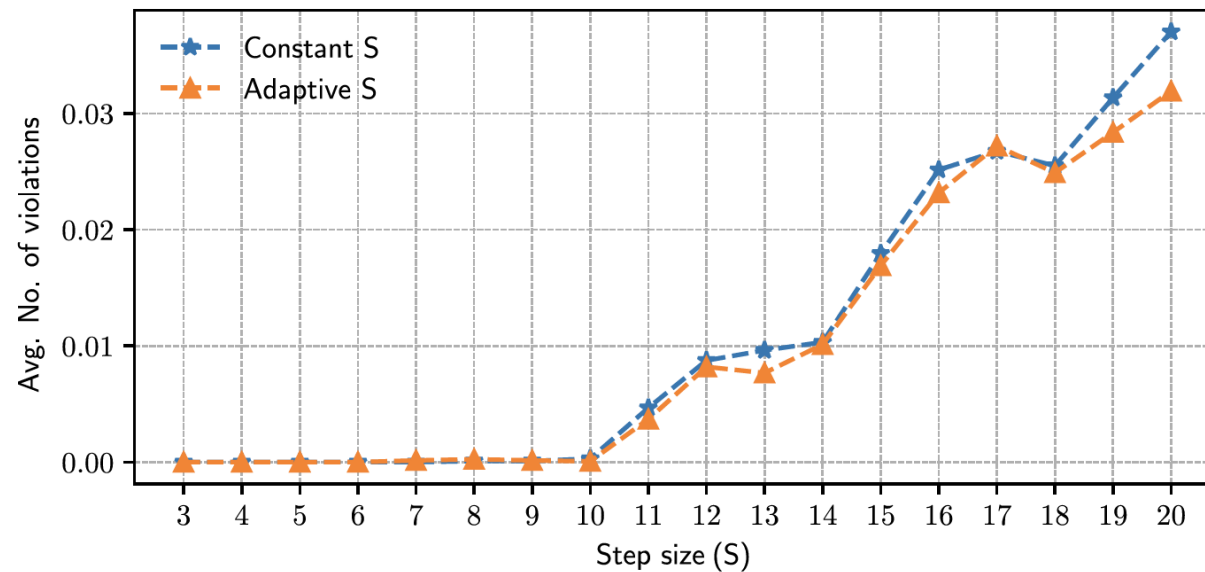
Constant $S$	CER	Type I ER	Type II ER	MSE
3	0.065	0.034	0.031	0.076
4	0.063	0.032	0.031	0.076
5	0.064	0.034	0.030	0.076
6	0.064	0.033	0.031	0.075
7	0.062	0.032	0.030	0.075
8	0.064	0.033	0.031	0.074
9	0.065	0.033	0.031	0.076
10	0.063	0.033	0.031	0.073
11	0.065	0.033	0.031	0.074
12	0.062	0.032	0.030	0.074
13	0.066	0.034	0.032	0.075
14	0.063	0.034	0.029	0.074
15	0.063	0.032	0.031	0.076
16	0.064	0.033	0.031	0.074
17	0.064	0.033	0.031	0.074
18	0.062	0.032	0.030	0.075
19	0.064	0.033	0.030	0.074
20	0.064	0.033	0.031	0.075
LA-CB-A average	0.064	0.033	0.031	0.075
Maximum priority	0.062	0.032	0.030	0.074
Content-weighted	0.078	0.040	0.039	0.118
Maximum information	0.054	0.032	0.022	0.054
Randomized	0.084	0.042	0.042	0.254

# Results – Study 3

**Table 5.** Summary of content constraint violations ( $\bar{V}$ )

Measures	Average $\bar{V}$	Max $\bar{V}$	Min $\bar{V}$
LA-CB-C	0.0110	0.0370	0
LA-CB-A	0.0102	0.0319	0
Maximum priority*	0.0540	–	–
Content-weighted*	2.2295	–	–
Maximum information*	8.1380	–	–
Randomized*	7.0230	–	–

*Note.* This table summarizes the statistics of  $\bar{V}$  for 18 step sizes. Methods with (\*) do not include step sizes to make item selections and therefore maximum and minimum  $\bar{V}$  are not applicable.





**Table 6.** Overall exposure control indices

Methods	LA-CB-C	LA-CB-A
Maximum exposure rate	0.178	0.178
Over-exposed (%)	0	0
Never exposed (%)	0	0
$\chi^2$	20.297	20.289

- the LA-CB methods **perform better** than the CWI and MPI methods
  - controlling constraints, while still maintaining high classification accuracy
- **different stopping rules** can be evaluated and optimally determined
- integrated with the **shadow test approach**
- how the LA-CB methods work when integrated with **other item selection methods**

*The End. Thanks  
for Listening!*



beijing normal  
university

**谢谢大家  
多谢晒~  
ありがとう  
Danke  
Merci**

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# Results – Study 1

