

**Item Selection Methods based on  
Multiple Objective Approaches for  
Classification of Respondents into Multiple Levels**

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# Current Study

Current Study  
Classification Testing  
Item Selection  
Simulation Study  
Conclusion

Is it possible to develop new item selection methods which take advantage of the fact that we want to classify into multiple categories?

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# Classification Testing

- Classification into one of several, mutually exclusive categories



Wald, 1947; Eggen & Straetmans, 2000

- Selecting the (next) item based on some criterion
- Objective:  
Maximization of Fisher information at some point on the ability scale

Sequential Classification Testing

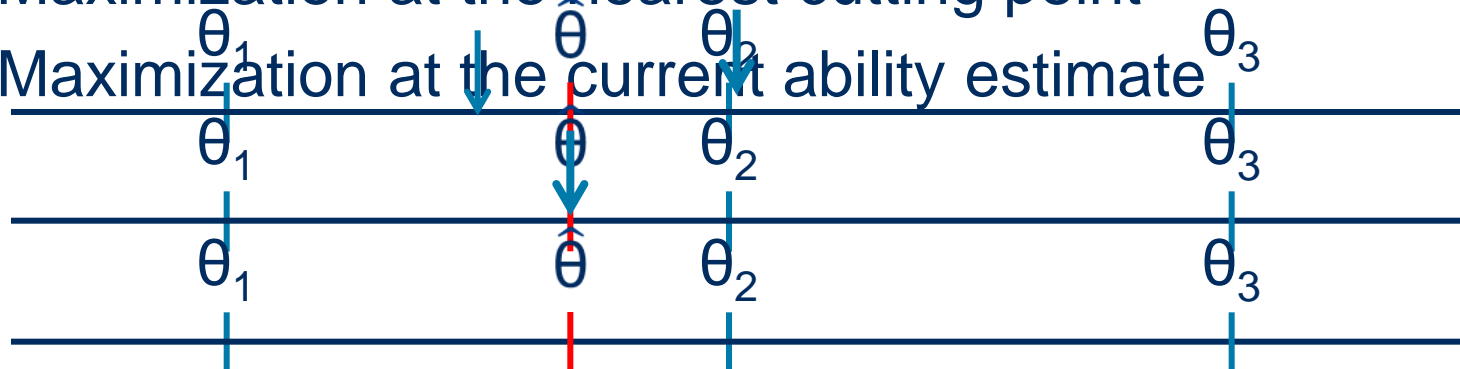


Adaptive Classification Testing



## Current methods (Eggen & Straetmans, 2000)

- Randomization
- Maximization at the middle of the cutting points
- Maximization at the nearest cutting point
- Maximization at the current ability estimate



## New methods

- Taking multiple points on the ability scale into account
- Based on multiple objective approaches (Veldkamp, 1999)

## New methods

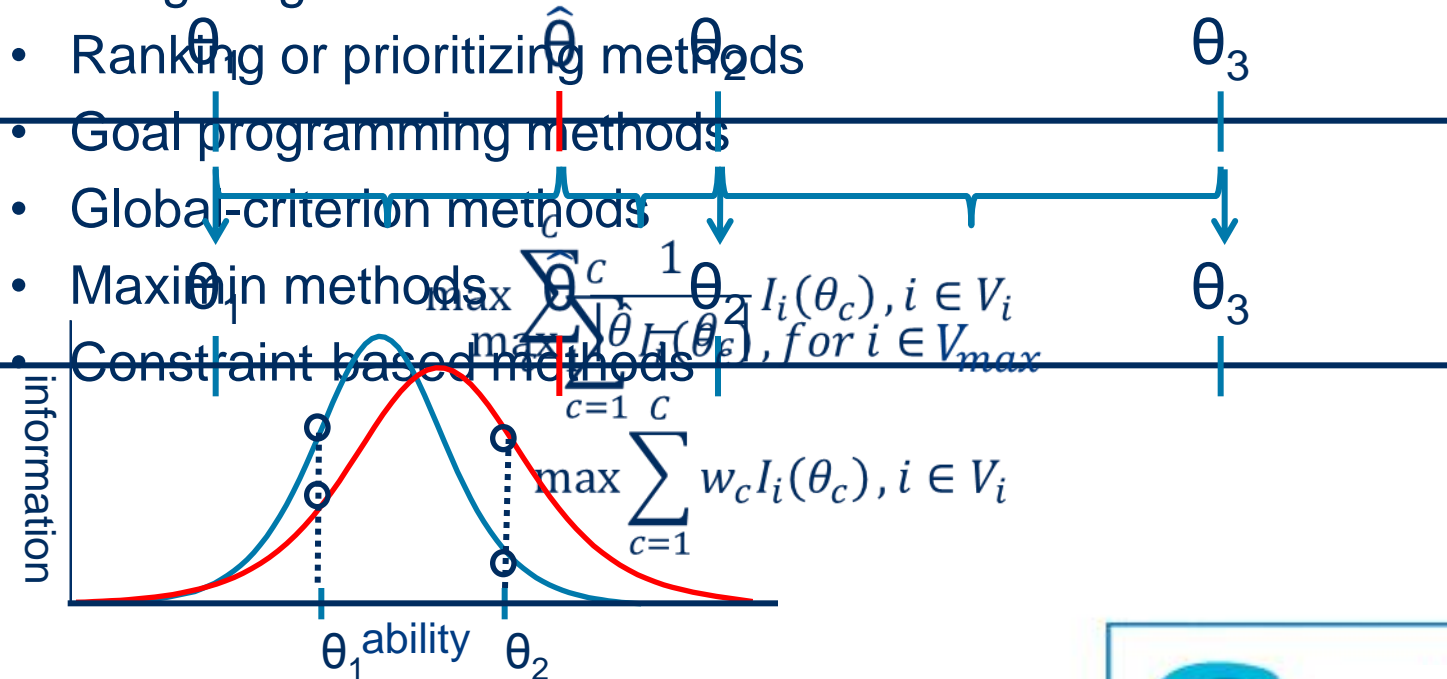
- Multiple objective approaches

- Weighting methods
- Ranking or prioritizing methods
- Goal programming methods

• Global-criterion methods

• Maximin methods

• Constraint based methods



# Simulation Study

Two item pools:

- 500 items
- $\alpha \sim N(1.0, 0.25)$
- $\beta \sim N(0.0, 1.0)$  &  $\beta \sim N(0.0, 2.0)$

Simulees:

- 1000 simulees per item selection method,  $\theta \sim N(0.0, 1.0)$

SPRT:

- $\alpha = \beta = 0.05$
- $\delta = 0.10$
- Cutting points: -1.0 & 1.0

8 item selection methods

# Simulation Study: Results

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Item selection method	Broad item pool		Peaked item pool	
	ATL	PCD	ATL	PCD
Random	99.0	0.77	95.6	0.77
Estimate Based	77.1	0.88	74.5	0.89
Middle cutting points	78.6	0.89	76.6	0.87
Nearest cutting point	80.5	0.86	75.0	0.89
Weighting method	79.3	0.87	74.4	0.89
Goal programming method	82.3	0.87	79.9	0.86
Global-criterion method	86.0	0.85	83.1	0.85
Maximin method	85.6	0.85	82.9	0.84

Note: ATL = average test length, PCD = percentage of correct decisions.

- Sequential Classification Tests **higher ATL** than Adaptive Classification Tests
- Sequential Classification Tests **slightly lower PCD** than Adaptive Classification Tests
- Results also hold with three and four cutting points

## Concluding remarks:

- Other item pools
- Other SPRT settings
- Other ability distributions
- Lower maximum number of items
- High average test length
- Other methods can be based on multiple objective approaches