#### Item Response Theory: Basic Notions

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#### Overview

- Some problems with Classical Theory
- The structure of a theory
- The prototype of IRT: Scalogram analysis
- The logistic function (a bit of mathematics)
- The Rasch model
- Our tasks
- What to do if our hypothesis fails
- The Partial Credit Model
- Closing the circle

# An interesting, yet often ignored aspect of classical theory

- An old test, X, is replaced by a new test, Y
- An interesting question: do X and Y measure the same construct?
- Possible answer: if they do, then the correlation between both tests  $\rho(X, Y)$  must equal one.
- Objection: What we observe (scores) is blurred (polluted) by measurement error, and these errors will suppress ('attenuate') the correlation

Correction for attenuation  

$$\rho(T_X, T_Y) = \frac{\rho(X, Y)}{\sqrt{\rho(X, X') \times \rho(Y, Y')}}$$

Example: 
$$\rho(X, Y) = 0.70$$
  
but  $\rho(X, X') = 0.85$  and  $\rho(Y, Y') = 0.60$ 

So, we find: 
$$\rho(T_X, T_Y) = \frac{0.70}{\sqrt{0.85 \times 0.60}} = 0.98$$

#### Some weak points of classical theory

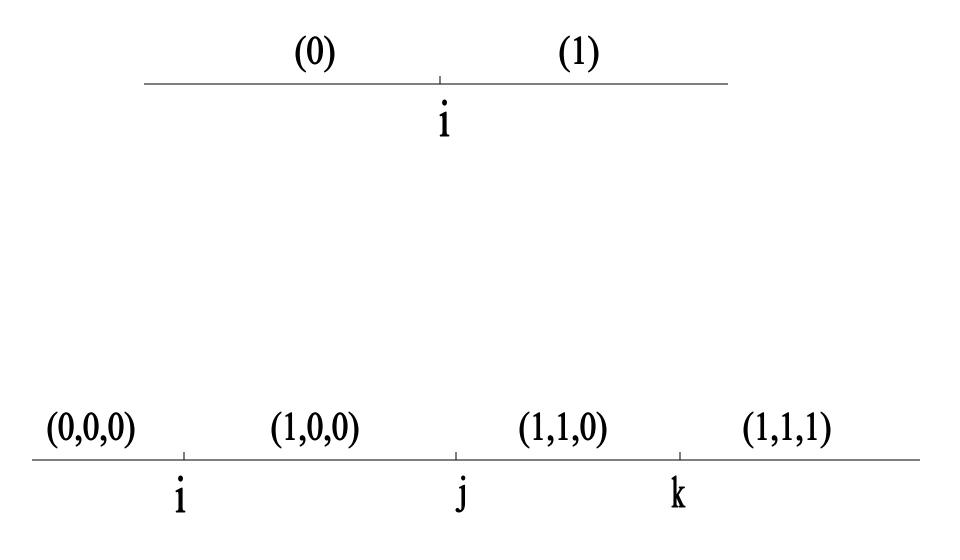
- The construct plays a minor role; central concept is the score
- Indices of difficulty and discrimination are population dependent
- Comparison of scores on different tests is very difficult and often impossible
  - Example: at university P students of faculties a, b and c take a test of English in March. In May students of faculties d, e and f take another exam of English
  - How can we compare the performance of faculties a and d?

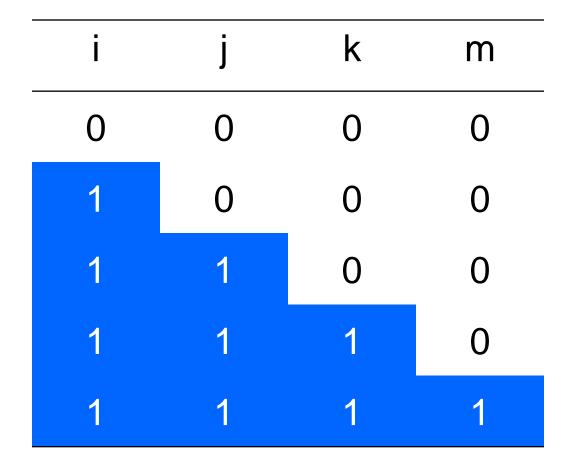
#### The structure of a theory

- A theory is a narrative about the observable world
  - Using concepts and relations between concepts
  - Clarifying the relation between concepts and observable phenomena
  - Imposing restrictions on 'possible' phenomena
  - Falsifiable
- Is Classical Test Theory a theory in this sense?

#### Guttman's scalogram

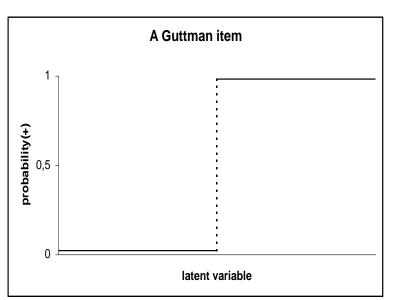
- The targeted concept is represented by a line (continuous variable) (for example, an attitude, a competence)
- Persons and items are represented by points on the line
- The observable responses reflect the ordinal relationship between the points 'person' and 'item' on the underlying line. This 'line' is nonobservable or latent





#items	possible	allowed
4	16	5
n	2 <sup>n</sup>	n+1

#### Guttman items



- Not-decreasing
- Not continuous ('step function')
  - The model is deterministic

### A bit of algebra: exponentiation $3^5 \times 3^7 = 3^{5+7} = 3^{12}$ Definition: $3^{-5} = \frac{1}{3^5}$

$$1 = 3^5 \times \frac{1}{3^5} = 3^5 \times 3^{-5} = 3^{5-5} = 3^0 = 1$$

For any positive number c, it holds that  $c^0 = 1$ 

 $3^{5}: \begin{cases} 3 \text{ is the basis} & (\text{positive}) \\ 5 \text{ is the exponent} & (\text{arbitrary number}) \end{cases}$ 

#### The number e

- A famous number which deserves its own name.
- The basis of the natural logarithms
- *e* = 2.7182818...
- The exponentiel function (see 'webinar2.xls', page 'exp-function'):

$$y = e^{x}$$
$$y = \exp(x)$$

#### The logistic function

$$f(x) = \frac{e^x}{1 + e^x}$$

$$f(x) = \frac{\exp(x)}{1 + \exp(x)}$$

"the exp of something (*x*) divided by one plus the exp of the same something"

#### The Rasch model (1)

• Definition of the item response function:

$$f_i(\theta) = P(X_i = 1 | \theta)$$

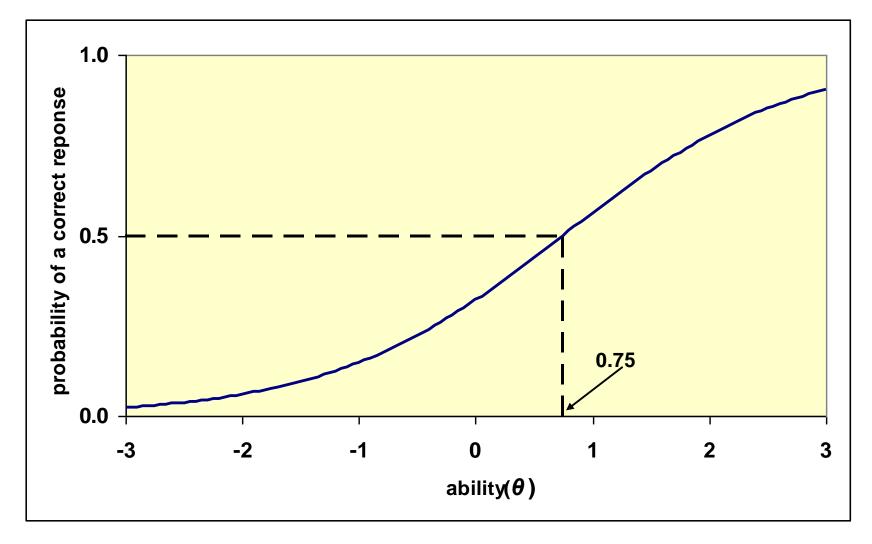
- X<sub>i</sub> stands for 'the score on item i'
- $\beta_i$  is a non-specified number (parameter)

$$f_i(\theta) = \frac{\exp(\theta - \beta_i)}{1 + \exp(\theta - \beta_i)}$$

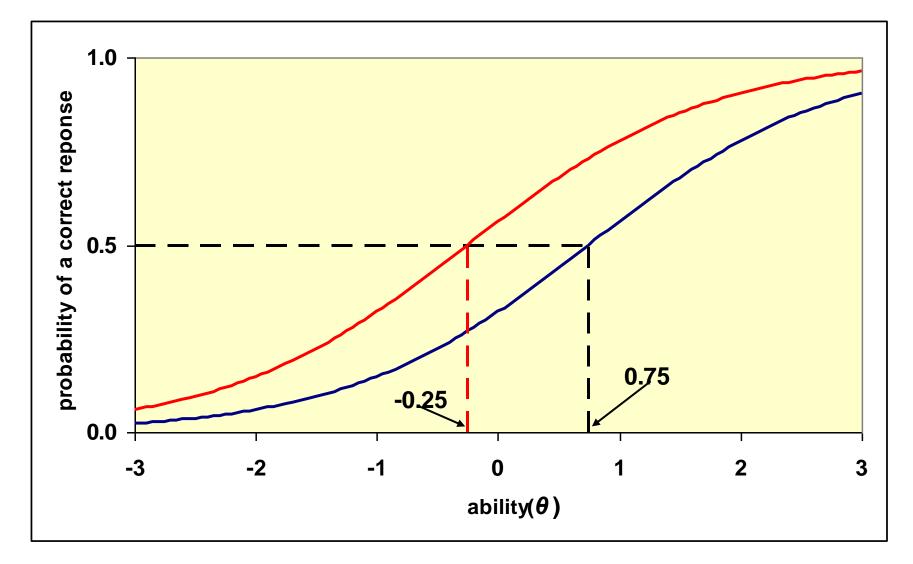
The Rasch model (2)  
$$f_i(\theta) = \frac{\exp(\theta - \beta_i)}{1 + \exp(\theta - \beta_i)}$$

- $\theta$  is the symbol for the latent variable
- Each item has its own function; hence  $f_i$
- $\beta_i$  is a non-specified number (parameter)
- If  $\theta = \beta_i$ , or  $\theta \beta_i = 0$  (and remember  $\exp(0) = 1$ ), then  $f_i(\theta) = \frac{1}{2}$
- $\beta_i$  is 'the amount of ability' needed to grant a probability of exactly  $\frac{1}{2}$  for a correct response.
- The more ability needed, the more difficult the item.
- $\beta_i$  is called the **difficulty parameter**

# The item response function for an item with difficulty parameter 0.75



### The item response functions for two Rasch items



#### Our tasks

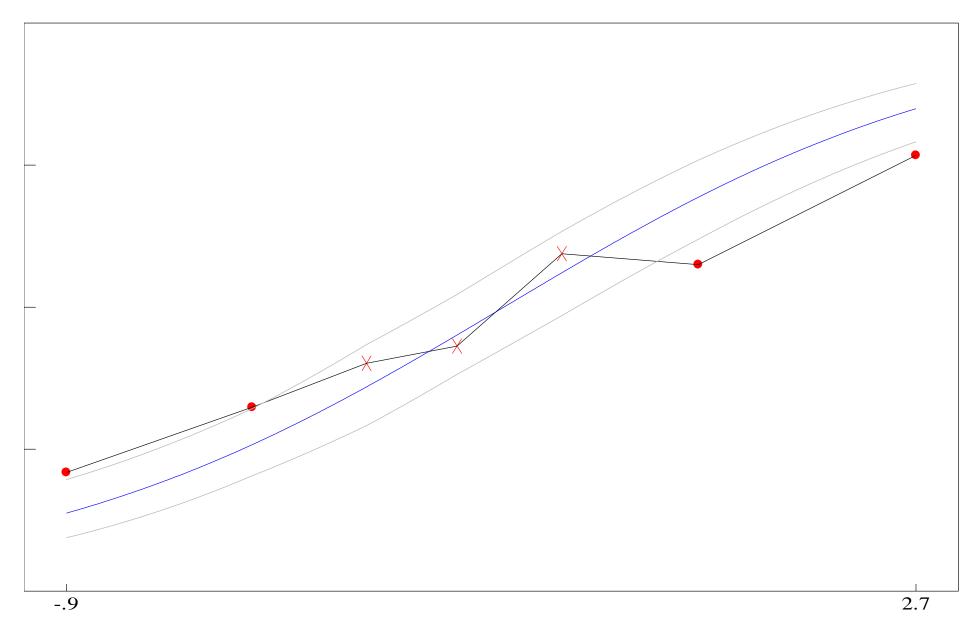
- What do we have?
  - A narrative (hypothesis)
  - A binary table with realisations of  $X_{vi}$  (0 or 1)
- What do we have to do?
  - Estimate the parameters
  - Check the narrative
  - Accept or reject the narrative
  - Use the test: go and measure

#### Parameter estimation

- Is a technically difficult problem
- Still a controversy about some methods
- The non-technical user can use public software, but beware...
- The matter is too complex to be discussed in a webinar
- This shows the necessity of interdisciplinary cooperation

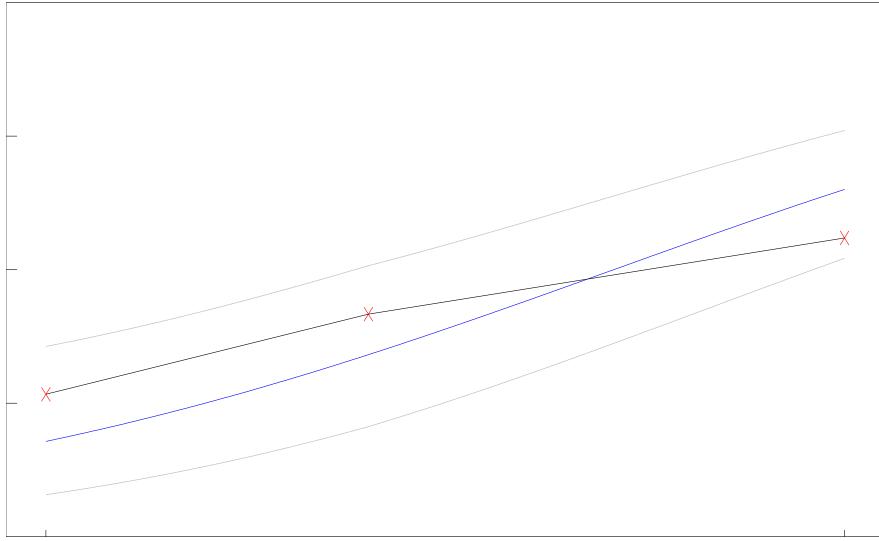
#### Check the narrative

- As the model is probabilistic, testing is not straightforward.
- Often statistical tests are used.
  - We need to understand the logic of statistical tests
- Graphical aids are helpful too



#### with n = 133 instead of n = 1332

Rel. item #: 4 Abs. item #: 4 Label: Item\_4 [:1]



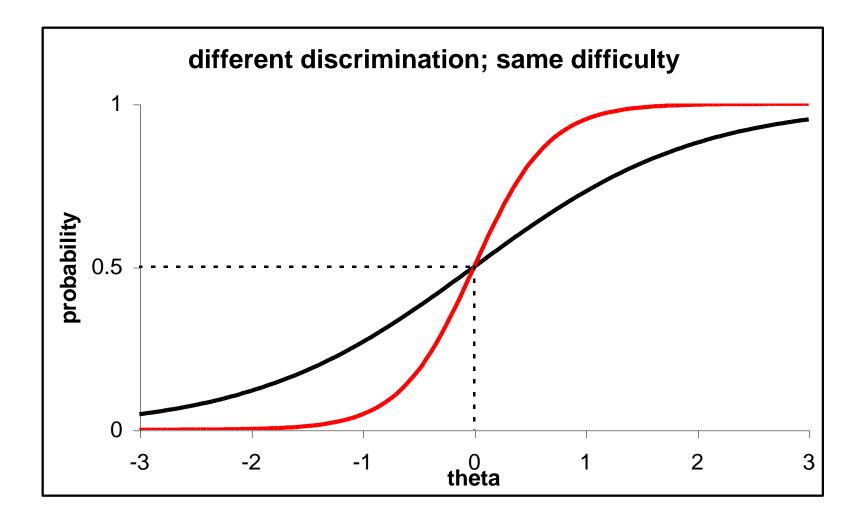
#### Power of a (statistical) test

- The probability that a deviation from the narrative will be discovered is dependent
  - On the seriousness of the deviation (which we do not know)
  - On the sample size (which is under our control)
- This probability is called the statistical power
- Discovered: yield a significant result
  - In every-day language: will ring the alarm
- The deviation from the narrative in the example has to do with discrimination
  - Item 4 discriminates less well than assumed by the narrative

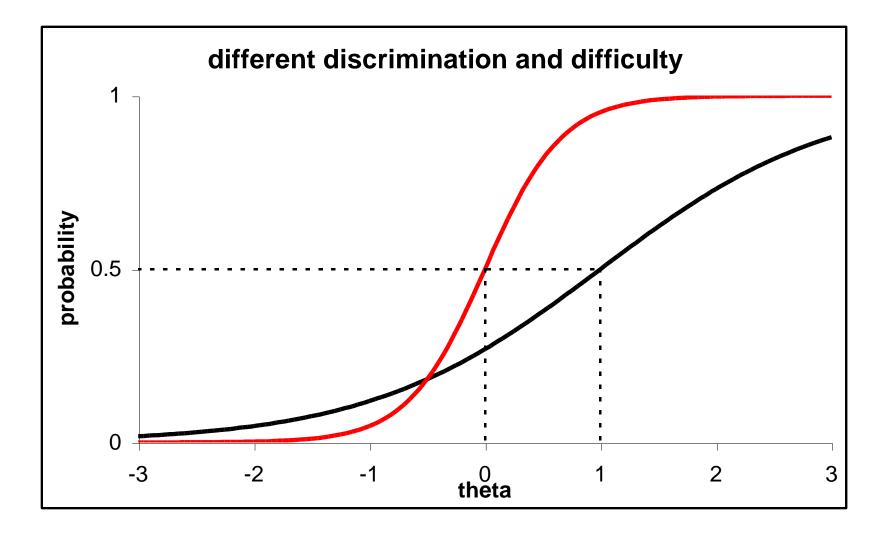
#### The narrative revisited

- The Rasch model as a narrative assumes
  - That all items discriminate equally
  - That the latent trait  $\theta$  is unidimensional
  - That there is local stochastic independence
- In most software, the statistical tests have little power if unidimensionality or local independence are not true.

#### Discrimination (1)



#### Discrimination (2)



#### **Discrimination and difficulty**

$$f_i(\theta) = P(X_i = 1 | \theta) = \frac{\exp[a_i(\theta - \beta_i)]}{1 + \exp[a_i(\theta - \beta_i)]}, \quad (a_i > 0)$$

This is known as the two-parameter logistic model (2PLM)

Rasch model:  $a_1 = a_2 = \cdots = a_k$ 

#### Local Independence: an example(1)

- Population: children in the age 6 to 14
- Test X: size of the feet
- Test Y: score on a reading comprehension test
- What is the sign of the correlation ρ(X, Y) in this population?
  - Negative?
  - Zero?
  - Positive?
- Why?

#### Local Independence: an example(2)

- The correlation is positive because
  - The older the children, the bigger their feet
  - The older the children, the better they read
- The variation in age 'explains' the correlation
- Proof: in a population of children of the same age (i.e., local) the correlation will vanish

#### Local independence in IRT

- In an arbitrary group (e.g., a class), the correlation between the answers on item *i* item *j* is (usually) positive, i.e., ρ(X<sub>i</sub>, X<sub>j</sub>) > 0
- In a population of students with the same value of  $\theta$ , this correlation is zero, i.e.,  $\rho(X, Y \mid \theta) = 0$
- This is an assumption, and the test to find out if it is tenable is very difficult because we cannot form a group of people with the same ability (ability is latent, i.e., not observable)

# Another look at conditional independence

- Given the latent variable, the probability of a correct answer must not depend on the answer to another item.
- Remember multiple matching items: if Tallin is assigned to Portugal, the probability of a correct answer for Lisbon is zero

Iceland	Tallin
Portugal	Bucharest
Roumania	Sofia
Bulgaria	Lisbon
Estonia	

#### Multidimensionality

- Example 1: a test of 'communicative competence' containing Reading items and Listening items
  - Probably better to analyse Reading and Listening separately
- Example 2: a Reading Comprehension test, consisting of (say) six text fragments and 5 questions per fragment.
  - In a fragment about history, students with more interest in or knowledge of history will probably do better than students weaker in this respect, even if their reading ability is equal

#### The testlet or task problem

- A set of items belonging together for some obvious reason (matching, text fragment) is often called an item bundle or a testlet.
- One can avoid the influence of lack of independence or multidimensionality by considering the testlet as a partial credit item instead of as a collection of binary items.
- And we have a beautiful model for this: the partial credit model

#### The Partial Credit Model (PCM)

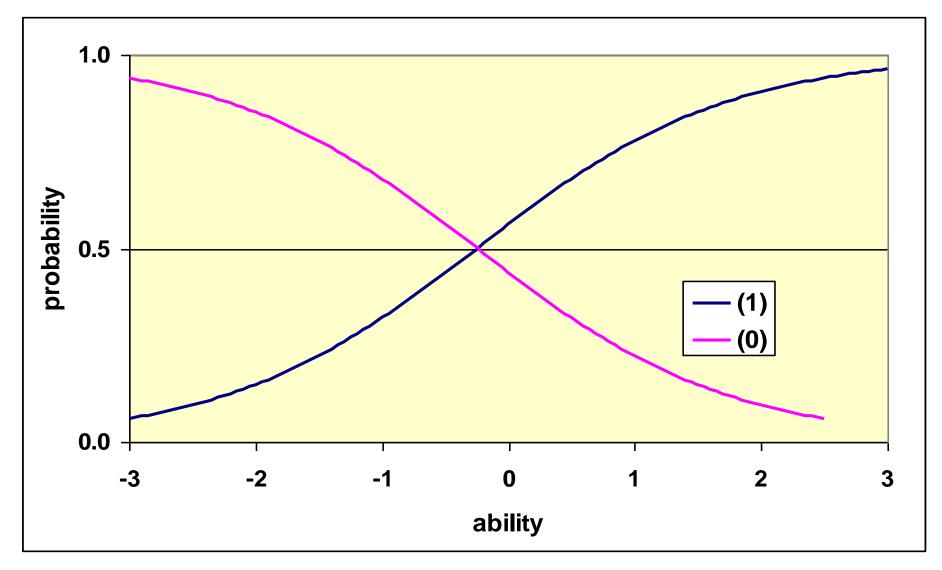
- Possible scores are 0, 1, 2,...,*m*
- In the Rasch model *m* = 1 and 1 item parameter
- In the PCM: *m* item parameters

score 1:  $\theta - \beta_{i1}$ score 2:  $2\theta - (\beta_{i1} + \beta_{i2})$ score 3:  $3\theta - (\beta_{i1} + \beta_{i2} + \beta_{i3})$ 

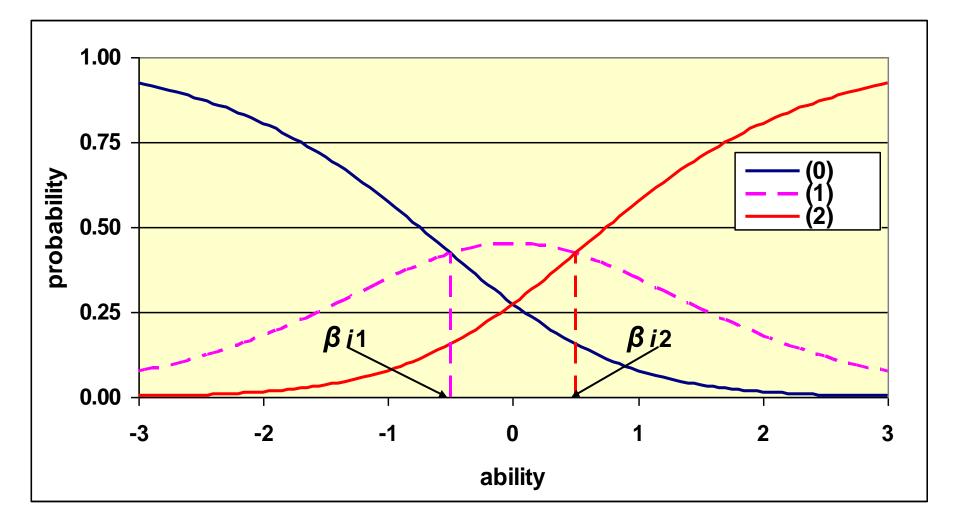
### PCM: exact formula (for m = 2) $P(X_i = 0 \mid \theta) = \frac{1}{D}$ $P(X_i = 1 | \theta) = \frac{\exp(\theta - \beta_{i1})}{D}$ $P(X_i = 2 \mid \theta) = \frac{\exp[2\theta - (\beta_{i1} + \beta_{i2})]}{1 - 1}$

*D* is the sum of the three numerators. (This guarantees that the sum of the 3 probabilities equals one.)

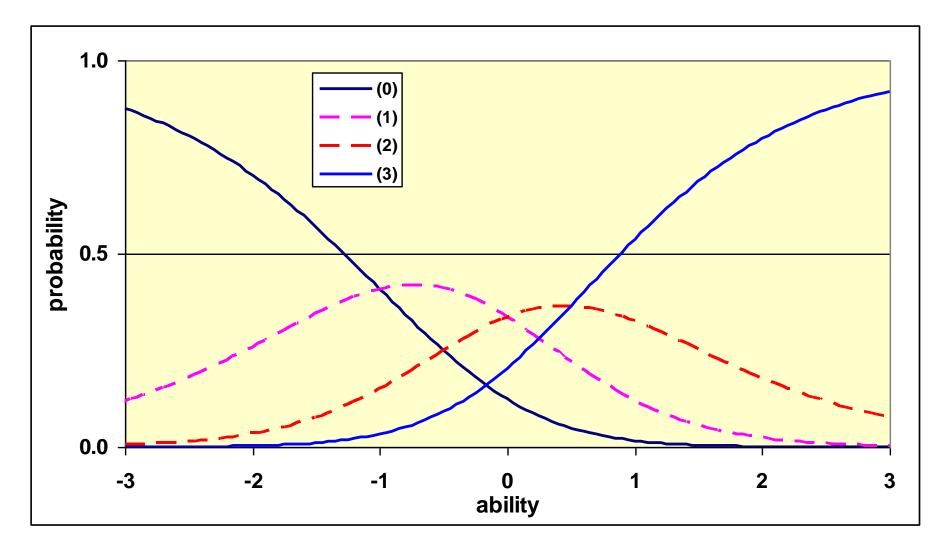
### Category reponse functions in the Rasch model



# Category reponse functions in the PCM (m = 2)



## Category reponse functions in the PCM (m = 3)



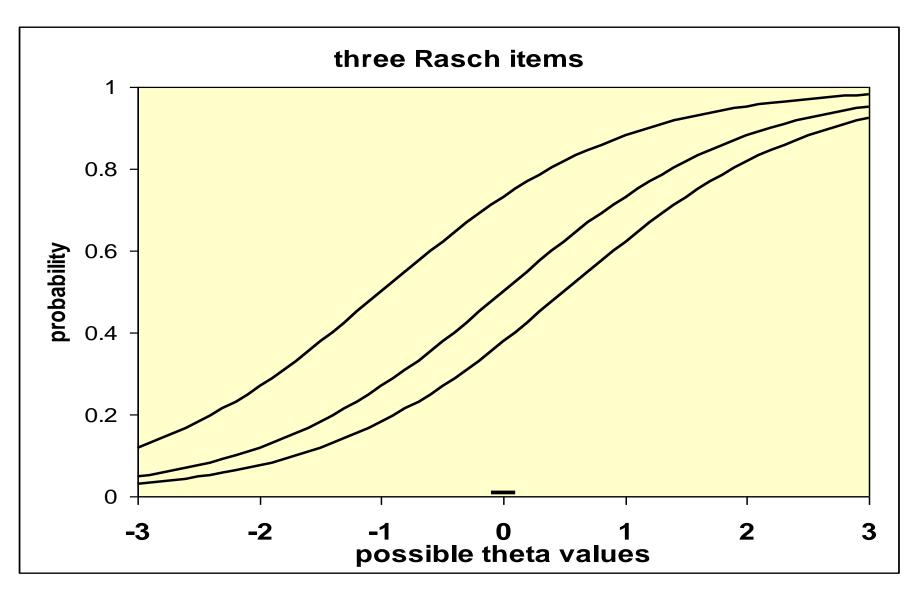
### The big advantage of IRT

- Performances on different tests can be meaningfully compared
  - If the tests measure the same construct
  - If they are calibrated together
  - If they have items in common
- Partial credit items and binary items can appear in any mixture

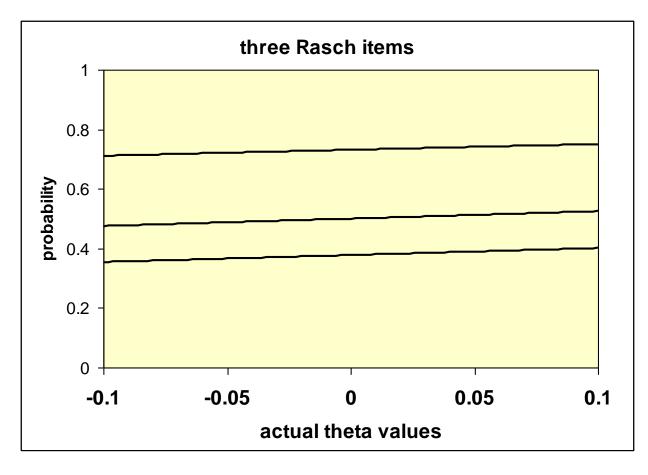
– Do not exaggerate *m* in the PCM

• But...

#### IRT is not the ultimate salvation



### What happens if the variation of the abilities is small?



- Reliability goes down
  - Validity of
     the Rasch
     model does
     not imply
     one has a
     reliable test

#### Thank you

### For questions related to this webinar, you can write me via e-mail:

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