

Dalla parte degli item

Georg Rasch e i modelli dell'Item Response Theory

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Dalla base

1. $1 + 2$

2. $\frac{2}{3} + \frac{3}{5}$

3. $2x^2 + 3x + 4 = 0$

$\frac{A_{soggetto}}{D_{item}}$:

• $A_{soggetto} > D_{item} \rightarrow > 1$

• $A_{soggetto} < D_{item} \rightarrow < 1$

Georg Rasch

Qualche trasformazione dopo...

$$P(x_{si} = 1 | \theta_s, b_i) = \frac{\exp(\theta_s - b_i)}{1 + \exp(\theta_s - b_i)}$$

Generalized Linear Model

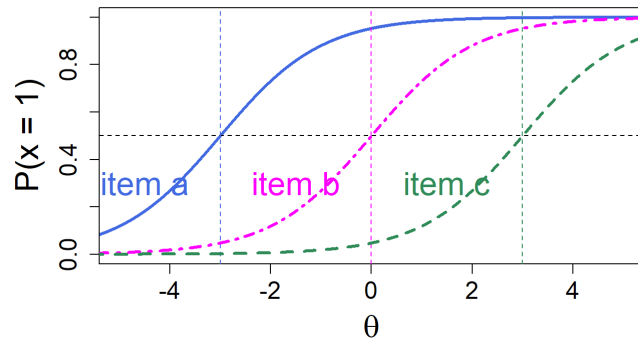
$$\text{logit} = \ln\left(\frac{P(x=1)}{1-P(x=1)}\right)$$

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

where

$$\eta = \alpha + \beta X$$

θ_s, b_i e il tratto latente



Rasch in pillole

- I marginali di riga (colonna) sono statistiche sufficienti
- Sono i dati che si adattano al modello
- La triade:
 1. Indipendenze locale
 2. Oggettività specifica
 3. Invarianza di misurazione

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Così simili e così diversi

- I parametri degli item
- Il modello si adatta ai dati
- I marginali di riga (colonna) non sono più statistiche sufficienti

Item Response Theory

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1PI

- Confuso con il modello di Rasch
- Il parametro a che c'è e non c'è (dipende a chi chiedi)
- 1 parametro e mezzo

e.g. De Mars (2010)

$$P(x = 1|\theta, b, a) = \frac{\exp[a(\theta_s - b_i)]}{1 + \exp[a(\theta_s - b_i)]}$$

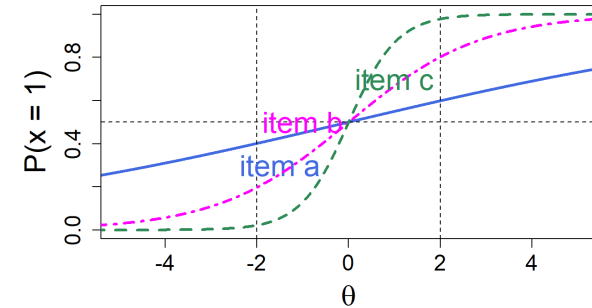
Da altre parti

$$P(x = 1|\theta, b) = \frac{\exp(\theta_s - b_i)}{1 + \exp(\theta_s - b_i)}$$

2PL (cambia la pendenza)

a (discrimination):

$$P(x = 1|\theta, b, a) = \frac{\exp[a_i(\theta_p - b_i)]}{1 + \exp[a_i(\theta_p - b_i)]}$$



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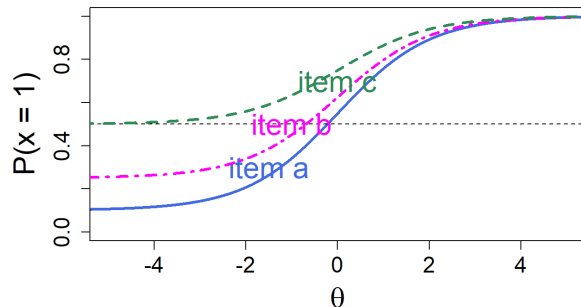
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Spostando gli asintoti

3PL (alza in basso)

c (lucky guess):

$$P(x = 1|\theta, b, a, c) = c_i + (1 - c_i) \frac{\exp[a(\theta_s - b_i)]}{1 + \exp[a(\theta_s - b_i)]}$$

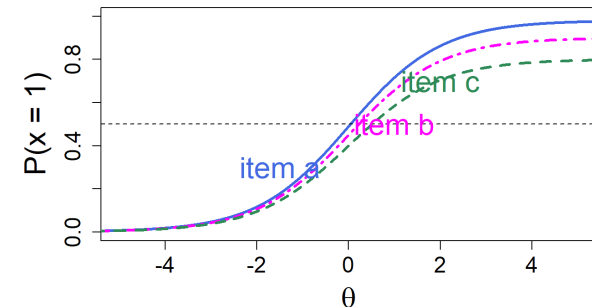


Spostando gli asintoti

4PL (abbassa in alto)

e (careless error):

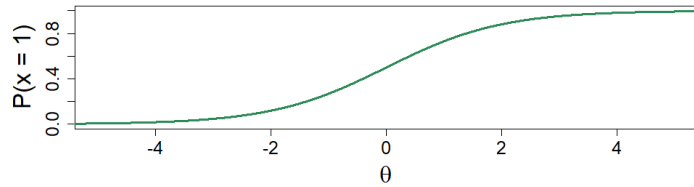
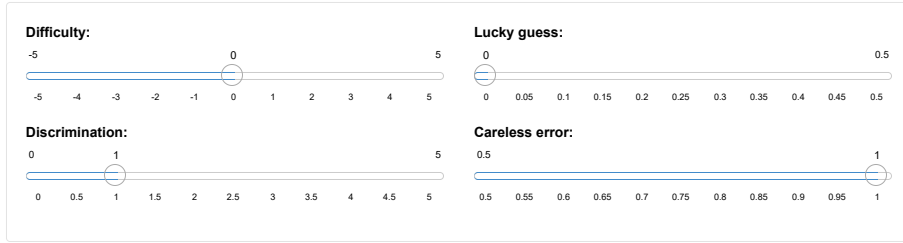
$$P(x = 1|\theta, b, a, c, e) = c_i + (e_i - c_i) \frac{\exp[a(\theta_s - b_i)]}{1 + \exp[a(\theta_s - b_i)]}$$



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Tutti quanti



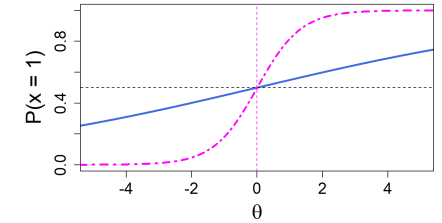
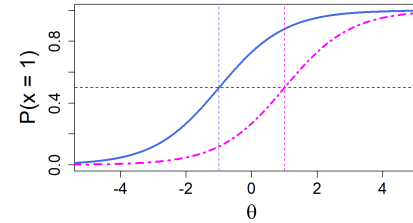
Differential Item Functioning

“Il cat calling è una molestia”:

- Si
- No

Uniforme

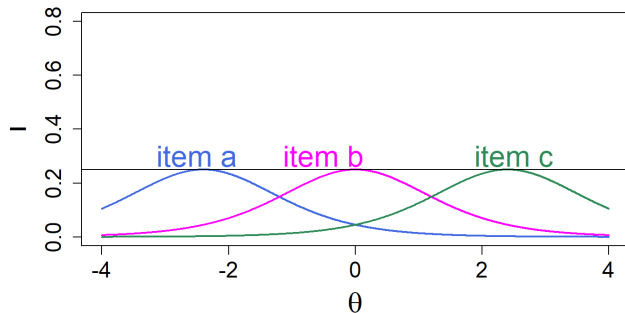
Non Uniforme



Item Information Function - Rasch

$$I_i(\theta) = P_i(\theta)(1 - P_i(\theta))$$

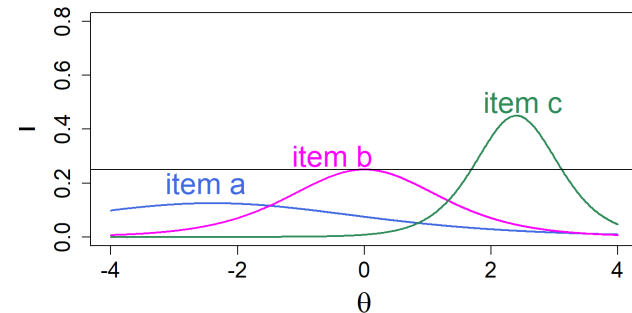
	item_a	item_b	item_c
difficulty	-3	0	3
discrimination	1	1	1



Item Information Function - 2PL

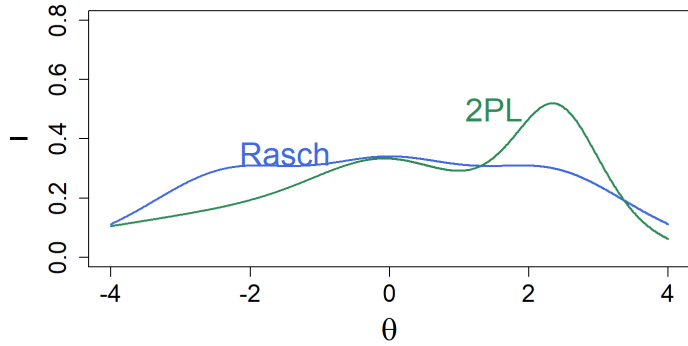
$$I_i(\theta) = a^2 P_i(\theta)(1 - P_i(\theta))$$

	item_a	item_b	item_c
difficulty	-3.0	0	3.0
discrimination	0.5	1	1.8



Test Information Function (TIF)

$$I(\theta) = \sum_{i=1}^n I_i(\theta)$$

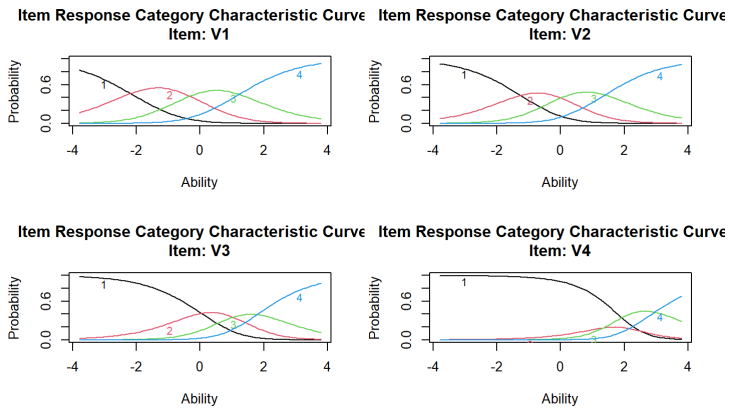


Oltre il dicotomico

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Polytomous Rasch Model

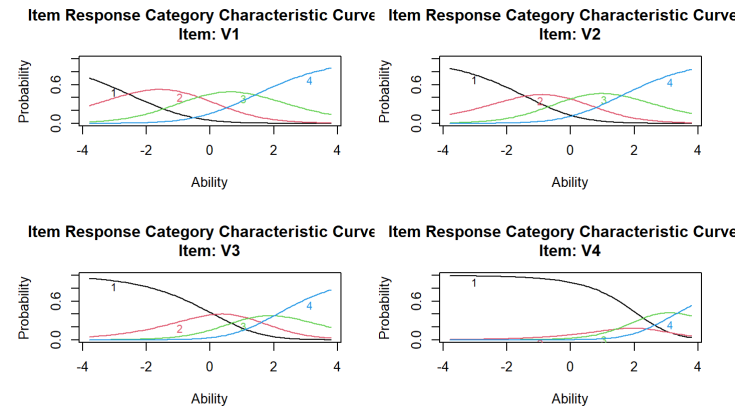
$$P(X = x, x > 0 | \theta_s, \tau_{ic}) = \frac{\exp(\sum_{c=1}^m \theta_s - \tau_{ic})}{1 + \exp(\sum_{c=1}^m \theta_s - \tau_{ic})}$$



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Partial credit model

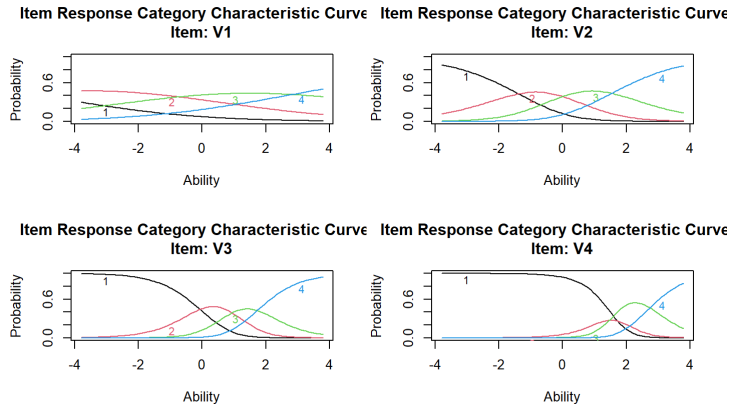
$$P(X = x, x > 0 | \theta_s, \tau_{ic}, a) = \frac{\exp[\sum_{c=1}^m a(\theta_s - \tau_{ic})]}{1 + \exp[\sum_{c=1}^m a(\theta_s - \tau_{ic})]}$$



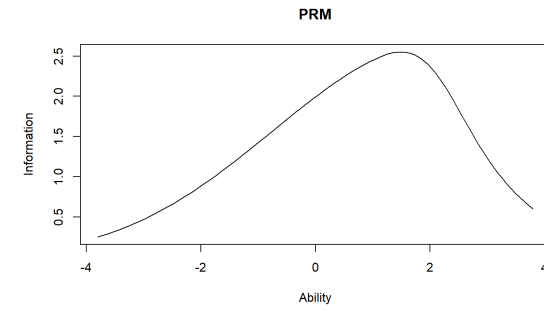
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Graded Response Model

$$P(X = x, x > 0 | \theta_s, \tau_{ic}, a_i) = \frac{\exp[\sum_{c=1}^m a_i(\theta_s - \tau_{ic})]}{1 + \exp[\sum_{c=1}^m a_i(\theta_s - \tau_{ic})]}$$



PRM, PCM, GRM: TIF



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Ackoejkl nkjgnkg

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