



# eRm – Extended Rasch Modeling

An R Package for the Analysis of Extended Rasch Models

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## Rasch Measurement

### • Item Response Theory (IRT)

- Analysis of response patterns in tests and questionnaires.
- Functional relationship between the probability to solve an item and some item and person parameter.
- A simple model is the Rasch model with one parameter  $\beta_i$  for each item and one parameter  $\theta_v$  for each subject.
- Rasch model as a seal of approval of a test (fairness, scaling).

### • Rasch measurement scale

- Example: A temperature scale in physics is clearly defined. What about a scale for mathematical ability?
- The Rasch model generates a scale for a latent trait  $\Psi$ .

## Contents

- **Introduction to Rasch measurement**
  - Item response theory
  - Rasch measurement scale
- **The Rasch model and extensions**
  - RM and LLTM
  - RSM and LRSM
  - PCM and LPCM
- **Model hierarchy**
- **A unified CML approach**
- **Organization of the eRm routine**

## The Rasch Model

### • Rasch model equation for dichotomous items

	X					
	I <sub>1</sub>	I <sub>2</sub>	I <sub>3</sub>	I <sub>4</sub>	I <sub>5</sub>	R <sub>v</sub>
P <sub>1</sub>	1	1	1	1	1	5
P <sub>2</sub>	1	1	1	0	1	4
P <sub>3</sub>	0	0	1	1	1	3
P <sub>4</sub>	1	0	0	0	0	1
P <sub>5</sub>	0	0	0	0	0	0
...						

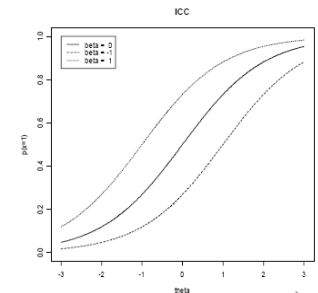
$$P(X_{vi} = 1 | \beta_i, \theta_v) = \frac{\exp(\theta_v - \beta_i)}{1 + \exp(\theta_v - \beta_i)}$$

$\beta_i$  ... item parameter

$\theta_v$  ... person parameter

### • Assumptions

- Unidimensionality
- Sufficiency of the raw score
- Parallel item characteristic curves
- Local independence



## Linear Logistic Test Model

### • LLTM (Linear Logistic Test Model)

- Linear reparameterization of the Rasch model.

$$\beta_i = \sum_{j=1}^p w_{ij} \eta_j$$

- LLTM as a more parsimonious model.
- LLTM as a more general model in terms of repeated measurements and certain effects.
- Concept of virtual items.

	$\beta_1$	$\beta_2$	...	$\beta_k$	$\tau$	$\delta$	$v$	$\rho$	
$\beta_1^*$	1								B <sub>1</sub>
$\beta_2^*$		1							
$\vdots$			$\ddots$						
$\beta_k^*$				1					B <sub>2</sub>
$\beta_{k+1}^*$	1				1				
$\beta_{k+2}^*$		1				1			
$\vdots$			$\ddots$						
$\beta_{2k}^*$					1	1			B <sub>3</sub>
$\beta_{2k+1}^*$	1					1	1		
$\beta_{2k+2}^*$		1					1	1	
$\vdots$			$\ddots$						
$\beta_{3k}^*$					1	1	1		B <sub>4</sub>
$\beta_{3k+1}^*$	1					1	1	1	
$\beta_{3k+2}^*$		1					1	1	
$\vdots$			$\ddots$						
$\beta_{4k}^*$					1	1	1	1	B <sub>5</sub>
$\beta_{4k+1}^*$	1					1	1	1	
$\beta_{4k+2}^*$		1					1	1	
$\vdots$			$\ddots$						
$\beta_{5k}^*$					1	1	1	1	

## Partial Credit Models

### • PCM (Partial Credit Model)

- Each item category gets a partial credit (item-category parameter)
- Different number of categories per item allowed

$$P(X_{vik} = 1 | \theta_v, \beta_{ik}) = \frac{\exp(k\theta_v + \beta_{ik})}{\sum_{h=0}^{m_j} \exp(h\theta_v + \beta_{ih})}$$

$\beta_{ih}$  ... item-category parameter

### • LPCM (Linear Partial Credit Model)

- Linear decomposition of the item-category parameter

$$\beta_{ih} = \sum_{j=1}^p w_{ij} \eta_j$$

## Rating Scale Models

### • RSM (Rating Scale Model)

- Item response categories are rating scales (polytomous)

$$P(X_{vi} = k | \theta_v, \beta_i, \omega_0, \dots, \omega_m) = \frac{\exp(k(\theta_v + \beta_i) + \omega_k)}{\sum_{h=0}^m \exp(h(\theta_v + \beta_i) + \omega_h)}$$

$h, k \dots$  categories;  $h = 0, \dots, m$

$\omega_h \dots$  category parameter

### • LRSM (Linear Rating Scale Model)

- Linear decomposition of the item parameter

$$\beta_i = \sum_{j=1}^p w_{ij} \eta_j$$

## Model Hierarchy

### • Model Nesting

Rasch	⊂	RSM	⊂	PCM
	⊂		⊂	LPCM
LLTM	⊂	LRSM		

### • LPCM is the most general model

- All other models can be viewed as special cases of LPCM.
- Parameterization through appropriate choice of  $\mathbf{W}$ .
- Unified CML procedure which is able to estimate these models.

## A Unified CML Approach

- **Linear item parameter decomposition**

$$\beta = W\eta$$

$W$ ...design matrix  
 $\eta$ ...parameter vector

- **CML approach**

- Estimation of  $\hat{\eta}$
- Likelihood conditioned on the raw score  $\rightarrow \theta$  vanishes

$$\log L_C = \sum_l \sum_{h_l} x_{+lh_l} \beta_{lh_l} - \sum_r n_r \log \gamma_r(\varepsilon)$$

$$\frac{\partial \log L_C}{\partial \eta_a} = \sum_l \sum_{h_l} w_{lh_a} \left( x_{+lh_l} - \varepsilon_{lh_l} \sum_r n_r \frac{\gamma_{r-h_l}^{(l)}(\varepsilon)}{\gamma_r(\varepsilon)} \right)$$

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## Organization of the eRm Routine

