

# Item Response Theory Using the `ltm` Package

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# 1 Let's Start with An Example

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- Situation:
  - ▷ A teacher offers a course on Calculus
  
- Question:
  - ▷ How can she find out which students have sufficiently understood the material?
  
- Solution:
  - ▷ Exams – Students need to take a test with questions on Calculus

# 1 Let's Start with Some Questions (cont'd)

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- What are exams trying to measure:



The Students' **Ability** in Calculus

- Features of Ability
  - ▷ something that is abstract
  - ▷ something that cannot be directly measured
  - ▷ something that is latent

# 1 Multivariate Data Set

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- A sample data set ('1' correct response; '0' wrong response)

Student	Item 1	Item 2	Item 3	...
1	0	0	0	...
2	0	1	1	...
3	1	1	1	...
4	1	0	1	...
⋮	⋮	⋮	⋮	⋮

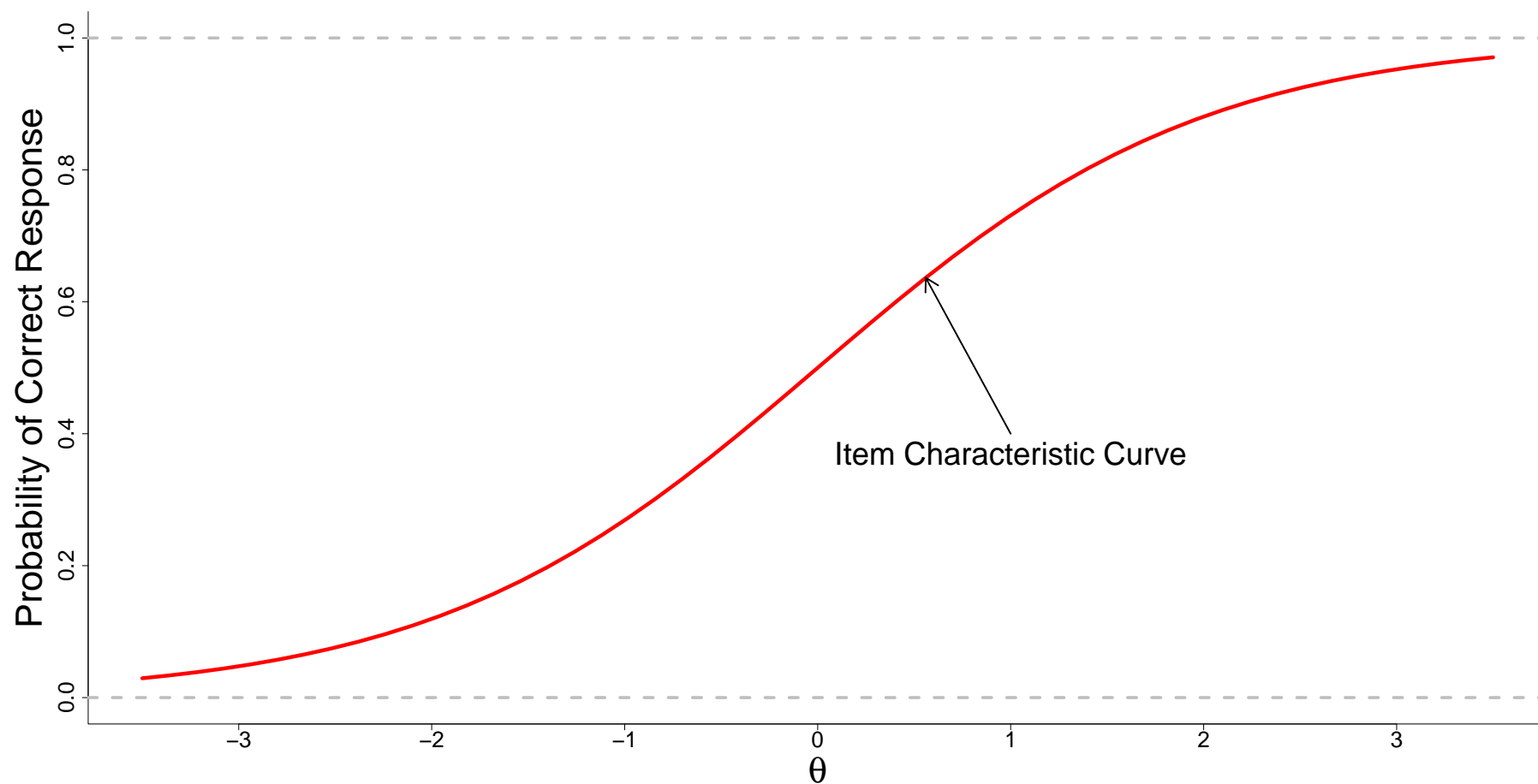
## 2 Item Characteristic Curve

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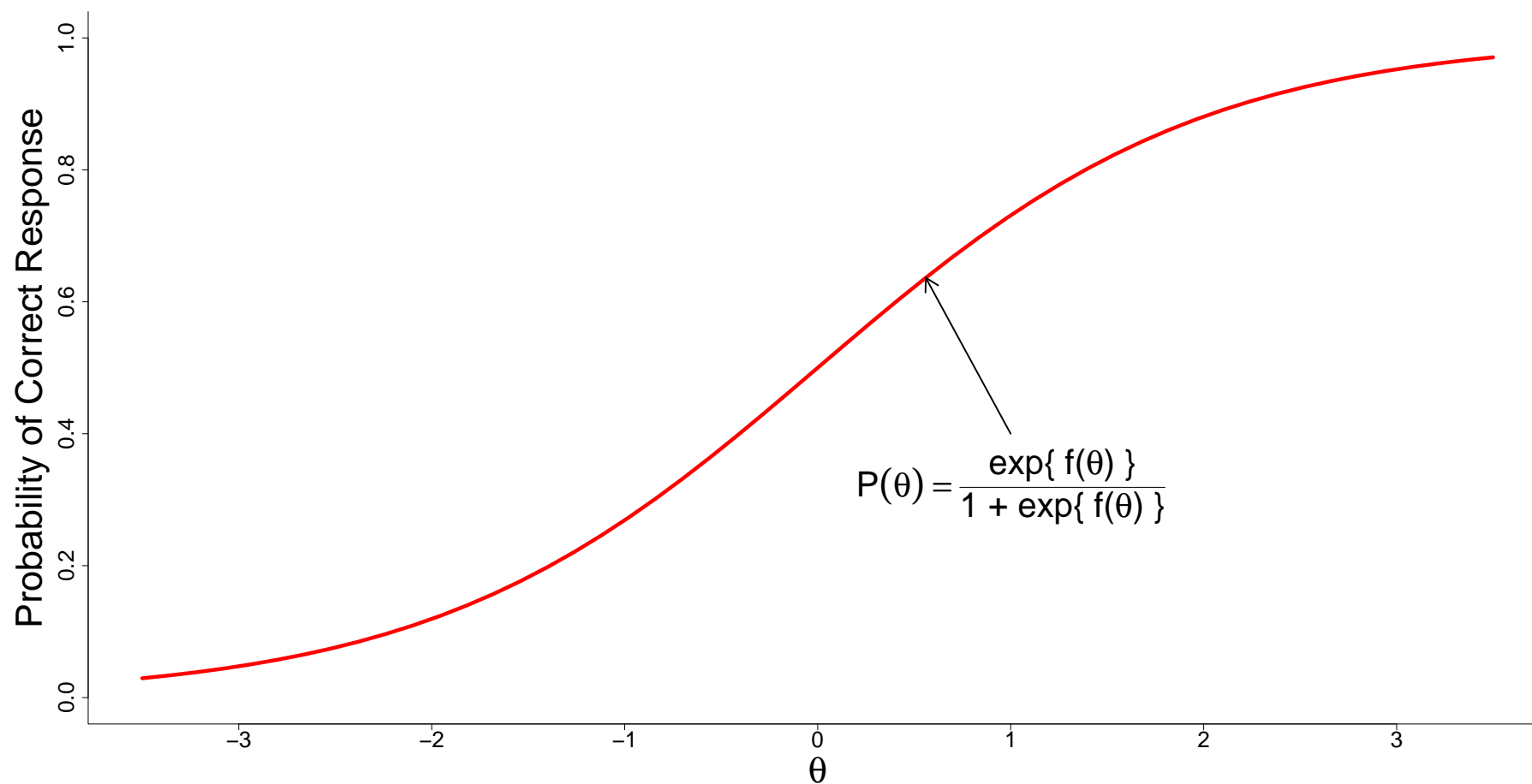
- A pool of items measuring a single latent trait
- Basic components
  - ▷  $\theta \in (-\infty, \infty)$ : latent ability
  - ▷  $P_i \in (0, 1)$ : probability of responding correctly in item  $i$

Item Characteristic Curve: functional relationship between  $\theta$  and  $P_i$

## 2 Item Characteristic Curve (cont'd)



## 2 Item Characteristic Curve & IRT Models



## 2 Item Characteristic Curve & IRT Models (cont'd)

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- Two Parameter Logistic Model

$$\log \frac{P_i(\theta)}{1 - P_i(\theta)} = \alpha_i(\theta - \beta_i), \quad i \text{ denotes the item}$$

- Parameters

- ▷ item difficulty parameter:  $\beta$
- ▷ item discrimination parameter:  $\alpha$
- ▷ person ability parameter:  $\theta$



## 2 Special Case: The Rasch Model

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- proposed by Georg Rasch (Danish mathematician) in 1960

$$\log \frac{P_i(\theta)}{1 - P_i(\theta)} = \theta - \beta_i, \quad i \text{ denotes the item}$$

- Properties and Features
  - ▷ closed-form sufficient statistics
  - ▷ restrictive  $\Rightarrow \alpha_i = 1$  for all  $i$
  - ▷ widely used

## 3 IRT Using the `ltm` Package

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- `ltm` package has been designed for user-friendly IRT analyses
- Functions for:
  - ▷ descriptive analyses
  - ▷ fitting common IRT models
  - ▷ post-processing of the fitted models
  - ▷ extra features

## 3 Descriptive Analyses

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```
>R descript(LSAT)
```

Descriptive statistics for the 'LSAT' data-set

Sample:

5 items and 1000 sample units; 0 missing values

Proportions for each level of response:

	0	1	logit
Item 1	0.076	0.924	2.4980
...			

Frequencies of total scores:

	0	1	2	3	4	5
Freq	3	20	85	237	357	298

Biserial correlation with Total Score:

	Included	Excluded
Item 1	0.3618	0.1128
...		

Cronbach's alpha:

	value
All Items	0.2950
Excluding Item 1	0.2754
...	

Pairwise Associations:

Item i	Item j	p.value
1	5	0.565
...		

### 3 Fit IRT Models

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```
>R fitRasch <- rasch(LSAT)
>R summary(fitRasch)
```

Call:

```
rasch(data = LSAT)
```

Model Summary:

log.Lik	AIC	BIC
-2466.938	4945.875	4975.322

Coefficients:

	value	std.err	z.vals
Dffclt.Item1	-3.6153	0.3266	-11.0680
Dffclt.Item2	-1.3224	0.1422	-9.3009
...			
Dscrmn	0.7551	0.0694	10.8757

Integration:  
method: Gauss-Hermite  
quadrature points: 21

Optimization:  
Convergence: 0  
max(|grad|): 2.9e-05  
quasi-Newton: BFGS

### 3 Fit IRT Models (cont'd)

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```
>R fit2PL <- ltm(LSAT ~ z1)
>R summary(fit2PL)
```

Call:

```
ltm(formula = LSAT ~ z1)
```

Model Summary:

log.Lik	AIC	BIC
-2466.653	4953.307	5002.384

Coefficients:

	value	std.err	z.vals
Dffclt.Item1	-3.3597	0.8669	-3.8754
...			
Dscrmn.Item1	0.8254	0.2581	3.1983
...			

Integration:  
method: Gauss-Hermite  
quadrature points: 21

Optimization:  
Convergence: 0  
max(|grad|): 0.024  
quasi-Newton: BFGS



### 3 Compare Fits with an LRT

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```
>R anova(fitRasch, fit2PL)
```

Likelihood Ratio Table

	AIC	BIC	log.Lik	LRT	df	p.value
fit1	4945.88	4975.32	-2466.94			
fit2	4953.31	5002.38	-2466.65	0.57	4	0.967

### 3 Ability Estimates

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```
>R factor.scores(fit2PL)
```

Call:

```
ltm(formula = LSAT ~ z1)
```

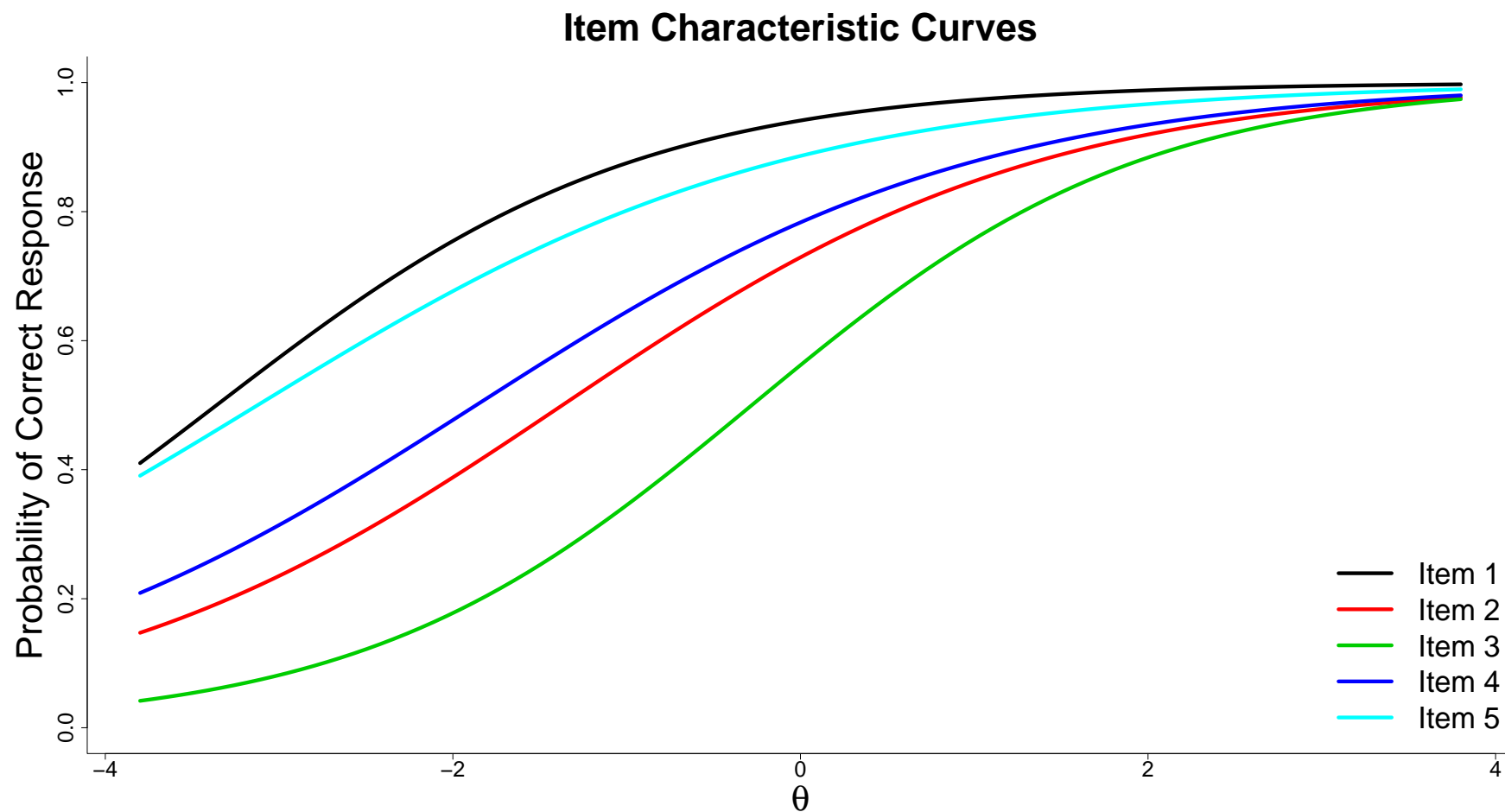
Scoring Method: Empirical Bayes

Factor-Scores for observed response patterns:

	Item 1	Item 2	Item 3	Item 4	Item 5	Obs	Exp	z1	se.z1
1	0	0	0	0	0	3	2.277	-1.895	0.795
2	0	0	0	0	0	1 6	5.861	-1.479	0.796
...									
29	1	1	1	1	0	28	29.127	0.139	0.833
30	1	1	1	1	1	298	296.693	0.606	0.855

### 3 Plot ICCs

```
>R plot(fit2PL, legend = TRUE, cx = "bottomright")
```



## 4 Extra Features of ltm

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- IRT Models:

- ▷ Graded Response Model for polytomous items  $\Rightarrow$  `grm()`
- ▷ Latent Trait Model with 2 latent variables  $\Rightarrow$  `ltm()`
- ▷ Birnbaum's Three Parameter Model  $\Rightarrow$  `tpm()`

- Goodness-of-Fit:

- ▷ Fit on the margins  $\Rightarrow$  `margins()`
- ▷ Bootstrap Pearson  $\chi^2$  test  $\Rightarrow$  `GoF.rasch()`
- ▷ Item- and Person-fit statistics  $\Rightarrow$  `item.fit()` & `person.fit()`

## 4 Extra Features of ltm (cont'd)

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- Plotting
  - ▷ Item and Test Information Curves
  - ▷ Item Person Maps
  
- A lot of other options . . .

**Thank you for your attention!**

**More Information for ltm is available at:**

<http://wiki.r-project.org/rwiki/doku.php?id=packages:cran:ltm>