



# LST Modeling with Logistic Item Response Models 1

## Purpose

Modeling the relationship between the item and time-specific latent variable and the manifest answer pattern  $Y = y$

## How?

For every item  $i$  we consider the relationship between the probability to answer in a particular category of this item ( $Y_i = y$ ) and the item and time-specific latent variable.

Steyer, R. and Partchev, I. (2001). Latent state-trait modeling with logistic item response models. In: R. Cudeck, S. du Toit, and D. Sörbom (Eds.), *Structural Equation Modeling: Present and Future* (pp. 481-520). Chicago: Scientific Software International.



## The Set of Possible Outcomes

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$$\Omega = \Omega_U \times \Omega_{O_1} \times \dots \times \Omega_{O_t} \times \dots \times \Omega_{O_n}$$

$$\Omega = \Omega_U \times (\Omega_{S_1} \times \Omega_{O_1}) \times \dots \times (\Omega_{S_t} \times \Omega_{O_t}) \times \dots \times (\Omega_{S_n} \times \Omega_{O_n})$$

$$Y_{it}: \Omega \rightarrow \{0, 1, 2, \dots, c_i\}$$

$$U: \Omega \rightarrow \Omega_U,$$

$$S_t: \Omega \rightarrow \Omega_{S_t},$$



## Assumptions

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$$P(Y_{it} = 0 | U, S_t) = \frac{1}{1 + \sum_{j=1}^{c_{it}} \exp \sum_{k=1}^j (q_{it} + \mu_{it} - ?_{itk})}$$

$$P(Y_{it} = y | U, S_t) = \frac{\exp \sum_{j=1}^y (q_{it} + \mu_{it} - ?_{itj})}{1 + \sum_{j=1}^{c_{it}} \exp \sum_{k=1}^j (q_{it} + \mu_{it} - ?_{itk})}, \quad y = 1, 2, \dots, c_{it}$$

### local stochastic independence

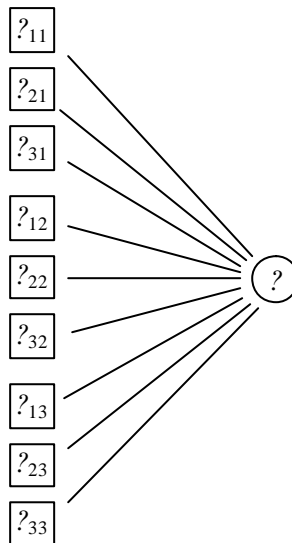
$$P(Y_{it} = y | \mathbf{q}_{it}, \mathbf{y}_{it}) = P(Y_{it} = y | \mathbf{q}_{it}),$$

where  $\mathbf{y}_{it}$  denotes the vector of all other items  $Y_{js}$ ,  $(j, s) \neq (i, t)$ .



## Singletrait Model: *path diagram*

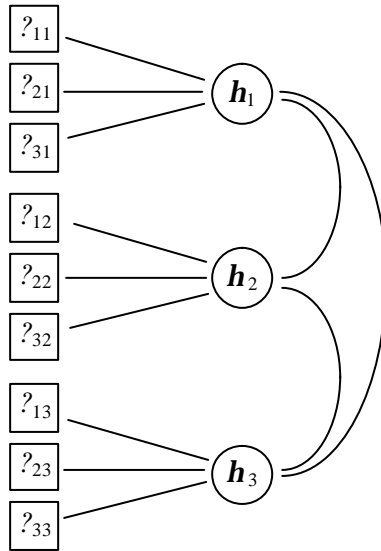
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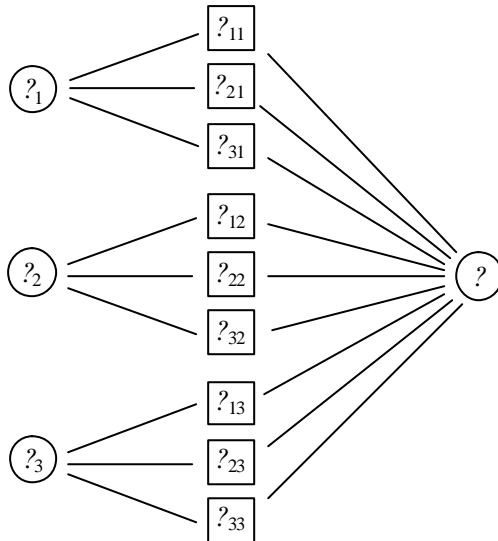
### Multistate Model: *path diagram*

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### Singletrait-Multistate Model: *path diagram*

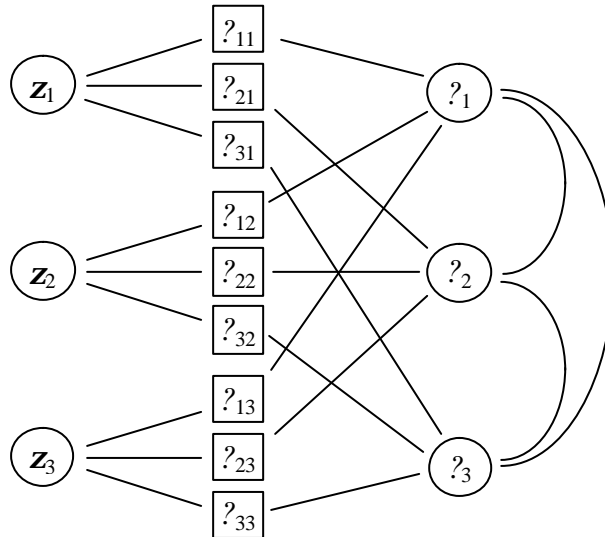
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## Multitrait-Multistate Model: *path diagram*

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## Models for Latent Variables

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**Table 1.** Assumptions on the item-and-time-specific latent state variables  $\mathbf{q}_{it}$  defining different latent state-trait models

Model	Assumptions on $\mathbf{q}_{it}$
(a) Singletrait	$(a_1) \mathbf{q}_{it} = \mathbf{x}$ <span style="float: right;"><math>i = 1, \dots, m,</math></span> $(a_2) E(\mathbf{x}) = 0$ <span style="float: right;"><math>t = 1, \dots, n</math></span>
(b) Multistate	$(b_1) \mathbf{q}_{it} = \mathbf{h}_t$ <span style="float: right;"><math>i = 1, \dots, m,</math></span> $(b_2) E(\mathbf{h}_t) = 0$ <span style="float: right;"><math>t = 1, \dots, n</math></span>



Model	Assumptions on $\mathbf{q}_{it}$
(c) Singletrait-Multistate	(c <sub>1</sub> ) $\mathbf{q}_{it} = \mathbf{x} + \mathbf{z}_t$ <span style="float: right;"><math>i = 1, \dots, m,</math> <math>t = 1, \dots, n</math></span>
	(c <sub>2</sub> ) $E(\mathbf{x}) = 0$
	(c <sub>3</sub> ) $E(\mathbf{z}_t) = 0$ <span style="float: right;"><math>t = 1, \dots, n</math></span>
	(c <sub>4</sub> ) $Cov(\mathbf{x}, \mathbf{z}_t) = 0$ <span style="float: right;"><math>t = 1, \dots, n</math></span>
	(c <sub>5</sub> ) $Cov(\mathbf{z}_s, \mathbf{z}_t) = 0$ <span style="float: right;"><math>s \neq t \quad s, t = 1, \dots, n</math></span>
(d) Multitrait-Multistate	(d <sub>1</sub> ) $\mathbf{q}_{it} = \mathbf{x}_i + \mathbf{z}_t$ <span style="float: right;"><math>i = 1, \dots, m, \quad t = 1, \dots, n</math></span>
	(d <sub>2</sub> ) $E(\mathbf{x}_i) = 0$ <span style="float: right;"><math>i = 1, \dots, m</math></span>
	(d <sub>3</sub> ) $E(\mathbf{z}_t) = 0$ <span style="float: right;"><math>t = 1, \dots, n</math></span>
	(d <sub>4</sub> ) $Cov(\mathbf{x}_i, \mathbf{z}_t) = 0$ <span style="float: right;"><math>i = 1, \dots, m, \quad t = 1, \dots, n</math></span>
	(d <sub>5</sub> ) $Cov(\mathbf{z}_s, \mathbf{z}_t) = 0$ <span style="float: right;"><math>s \neq t \quad s, t = 1, \dots, n</math></span>



In the partial credit model, we have

$$a := \sum_{i=1}^m \sum_{t=1}^n (c_{it} + 1)$$

response categories. Furthermore, there are at most

$$b_{max} := \sum_{i=1}^m \sum_{t=1}^n c_{it}$$

independent parameters (locations  $\mu_{it}$  and thresholds  $\kappa_{ij}$ ), where  $m$  and  $n$  denote the number of items and occasions, respectively.



## Models for Item Parameters

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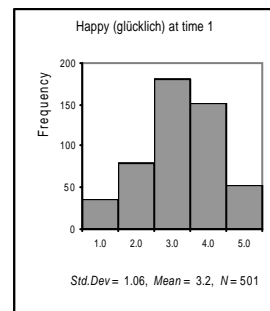
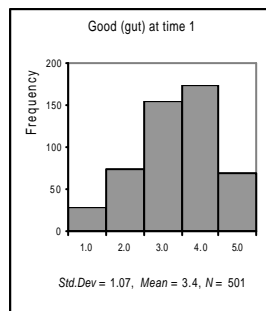
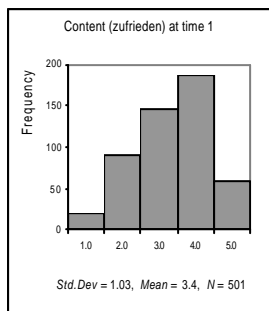
**Table 2.** Different models resulting from crossing five parameterizations of the locations parameters  $\mu_{it}$  with three models for the threshold parameters  $\kappa_{ij}$

		Models for the threshold parameters $\kappa_{ij}$			
		Equality of threshold distances within items			
		yes		no	
		Equality of thresholds across items			
		yes		no	
		Invariance of thresholds over time			
		yes	no	yes	no
		Models	$\mu$	P1111	P1112
for the	$\mu + \alpha_i$	P2111	P2112	P2121	P2122
location	$\mu + \beta_t$	P3111	P3112	P3121	P3122
parameters	$\mu + \alpha_i + \beta_t$	P4111	P4112	P4121	P4122
$\mu_{it}$	$\mu + \alpha_i + \beta_t + \gamma_{it}$	P5111	P5112	P5121	P5122



## Diagrams – Items at Time 1

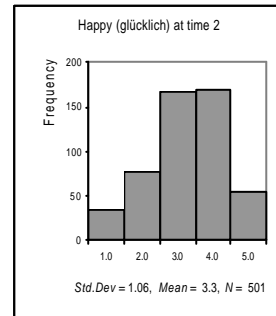
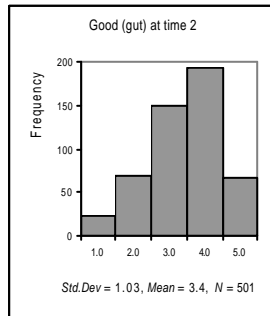
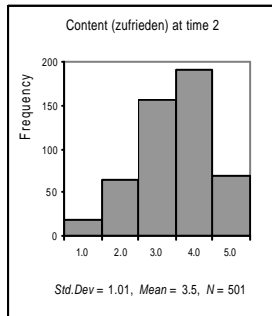
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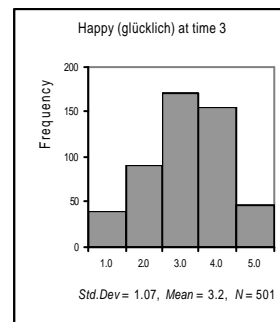
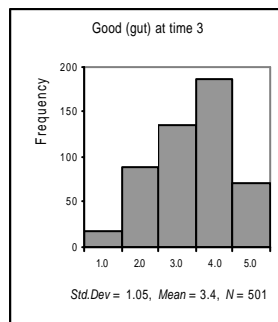
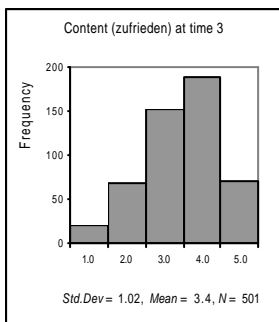
## Diagrams –Items at Time 2

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## Diagrams –Items at Time 3

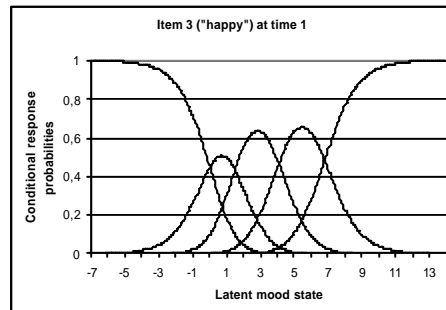
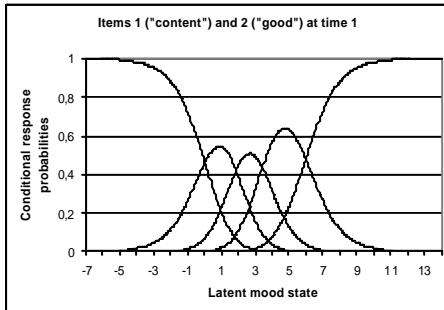
14





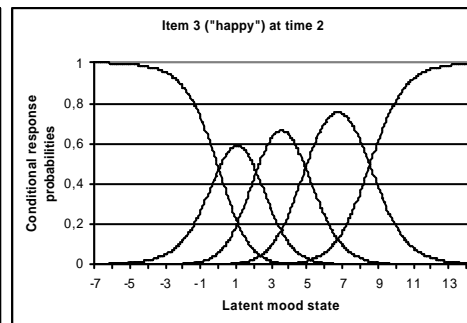
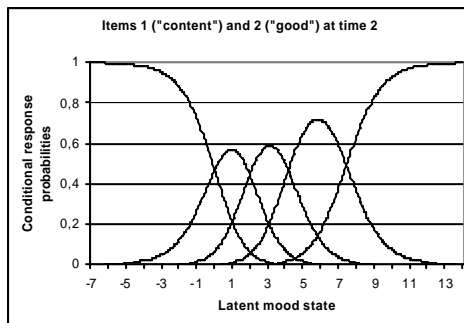
## Conditional Response Probabilities – Items at Time 1 –

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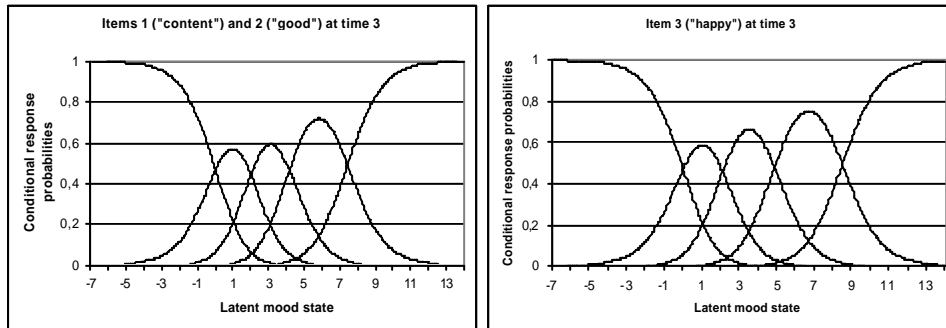
## Conditional Response Probabilities – Items at Time 2 –

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## Conditional Response Probabilities – Items at Time 3 –



## Item Parameter Estimates I

**Table 5.** Item parameter estimates with standard errors and fit statistics

Parameter	Estimate	SE	Unweighed Fit		Weighed Fit	
			MNSQ	T	MNSQ	T
$\mu$	3.476	0.02	1.06	0.92	1.03	0.48
$\beta_2$	0.961	0.04	1.11	1.74	1.13	1.98
$\beta_3$	0.838	0.04	1.13	1.97	1.15	2.20
$\kappa_{112} = \kappa_{212}$	1.848	0.10	0.94	-0.92	1.09	1.12
$\kappa_{113} = \kappa_{213}$	3.396	0.08	0.97	-0.51	1.05	0.76
$\kappa_{114} = \kappa_{214}$	5.967	0.12	0.81	-3.20	1.06	0.61
$\kappa_{312}$	1.489	0.15	0.67	-6.00	0.97	-0.36
$\kappa_{313}$	4.076	0.13	0.80	-3.31	1.01	0.19
$\kappa_{314}$	6.780	0.19	0.97	-0.49	1.22	1.67



## Deviance and Number of Parameters

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**Table 3.** Models with their deviance and number of parameters

Model	Deviance	Convergence	Number of parameters in the model for the		
			Latent variables	Expectations	Thresholds
<b>0</b>	11013.34	no	45	9	27
<b>A</b>	11017.07	no	9	9	27
<b>B</b>	10993.67	yes	6	9	27
<b>C</b>	11089.00	yes	4	9	27
<b>D</b>	13233.31	yes	6	9	1
<b>E</b>	11037.84	yes	6	9	9
<b>F</b>	11027.24	yes	6	9	9
<b>G</b>	11005.71	yes	6	9	12
<b>H</b>	11034.98	yes	6	1	12
<b>I</b>	11005.20	yes	6	7	12
<b>J</b>	11024.54	yes	6	7	12
<b>K</b>	11013.13	yes	6	5	12
<b>L</b>	11017.64	yes	6	3	12



## Covariances and Correlations

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**Table 4.** Covariances and correlations of the traits and state residuals

	$z_1$	$z_2$	$z_3$	$x_1$	$x_2$
$z_1$	2.161 (0.137)	0	0	0	0
$z_2$	0	4.088 (0.258)	0	0	0
$z_3$	0	0	4.631 (0.293)	0	0
$x_1$	0	0	0	0.872 (0.055)	.814
$x_2$	0	0	0	1.301 (0.101)	2.928 (0.185)



## Item Parameter Estimates II

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$\kappa_{122} = \kappa_{222}$	1.979	0.08	1.36	5.19	1.23	2.71
$\kappa_{123} = \kappa_{223}$	4.146	0.06	1.02	0.35	1.07	1.01
$\kappa_{124} = \kappa_{224}$	7.425	0.09	0.89	-1.74	1.17	1.93
$\kappa_{332}$	2.118	0.11	1.07	1.10	0.98	-0.20
$\kappa_{333}$	4.889	0.10	0.87	-2.09	0.95	-0.71
$\kappa_{334}$	8.513	0.14	0.60	-7.42	1.30	2.60

*Note:* The standard errors are computed with the quick option described in Wu et al. (1998, p. 137). These approximations are computed ignoring the covariances of the parameter estimates.



## Expectations, Variances, Consistency

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**Table 6.** *Expectations, variances, consistency, and occasion specificity coefficients of the item-and-time-specific latent variables*

	Expectations	Variances	Consistency	Occasion Specificity
$q_{11}$	3.476	3.033	0.288	0.712
$q_{21}$	3.476	3.033	0.288	0.712
$q_{31}$	3.476	5.089	0.575	0.425
$q_{12}$	4.437	4.960	0.176	0.824
$q_{22}$	4.437	4.960	0.176	0.824
$q_{32}$	4.437	7.015	0.417	0.583
$q_{13}$	4.314	5.504	0.159	0.841
$q_{23}$	4.314	5.504	0.159	0.841
$q_{33}$	4.314	7.559	0.387	0.613



## *Estimated, Centered, Standardized Thresholds I*

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**Table 7.** *Estimated, centered, and standardized thresholds for all items at each occasion of measurement*

$\kappa_{111}$	0.000	-3.476	-1.996	
$\kappa_{112}$	1.848	-1.628	-0.935	
$\kappa_{113}$	3.396	-0.080	-0.046	
$\kappa_{114}$	5.967	2.491	1.430	-0.387
$\kappa_{211}$	0.000	-3.476	-1.996	
$\kappa_{212}$	1.848	-1.628	-0.935	
$\kappa_{213}$	3.396	-0.080	-0.046	
$\kappa_{214}$	5.967	2.491	1.430	-0.387
$\kappa_{311}$	0.000	-3.476	-1.541	
$\kappa_{312}$	1.489	-1.987	-0.881	
$\kappa_{313}$	4.076	0.600	0.266	
$\kappa_{314}$	6.780	3.304	1.465	-0.173



## *Estimated, Centered, Standardized Thresholds II*

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$\kappa_{121}$	0.000	-4.437	-1.992	
$\kappa_{122}$	1.979	-2.458	-1.104	
$\kappa_{123}$	4.146	-0.291	-0.131	
$\kappa_{124}$	7.425	2.988	1.342	-0.471
$\kappa_{221}$	0.000	-4.437	-1.992	
$\kappa_{222}$	1.979	-2.458	-1.104	
$\kappa_{223}$	4.146	-0.291	-0.131	
$\kappa_{224}$	7.425	2.988	1.342	-0.471
$\kappa_{321}$	0.000	-4.437	-1.675	
$\kappa_{322}$	2.118	-2.319	-0.876	
$\kappa_{323}$	4.889	0.452	0.171	
$\kappa_{324}$	8.513	4.076	1.539	-0.210



*Estimated, Centered, Standardized Thresholds III* 25

$\kappa_{131}$	0.000	-4.314	-1.839	
$\kappa_{132}$	1.979	-2.335	-0.995	
$\kappa_{133}$	4.146	-0.168	-0.072	
$\kappa_{134}$	7.425	3.111	1.326	-0.395
$\kappa_{231}$	0.000	-4.314	-1.839	
$\kappa_{232}$	1.979	-2.335	-0.995	
$\kappa_{233}$	4.146	-0.168	-0.072	
$\kappa_{234}$	7.425	3.111	1.326	-0.395
$\kappa_{331}$	0.000	-4.314	-1.569	
$\kappa_{332}$	2.118	-2.196	-0.799	
$\kappa_{333}$	4.889	0.575	0.209	
$\kappa_{134}$	8.513	4.199	1.527	-0.158