



## Basic Concepts

1

The set of possible outcomes  
of the random experiment

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

Test-score variables

$$Y_{it}: \Omega \rightarrow \mathbb{R}$$

Projections

$$U: \Omega \rightarrow \Omega_U \text{ person projection}$$

$$S_t: \Omega \rightarrow \Omega_{S_t} \text{ situation projection}$$



## Theoretical Variables

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$$\mathbf{t}_{it} := E(Y_{it} | U, S_t) \quad \text{Latent state variable}$$

$$\mathbf{e}_{it} := Y_{it} - \mathbf{t}_{it} \quad \text{Measurement error variable}$$

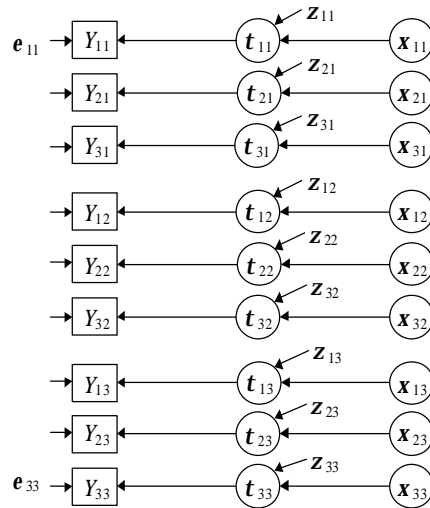
$$\mathbf{x}_{it} := E(Y_{it} | U) \quad \text{Latent trait variable}$$

$$\mathbf{z}_{it} := \mathbf{t}_{it} - \mathbf{x}_{it} \quad \text{Latent state residual}$$



## Path Diagram

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## Properties of Latent Variables

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Decomposition of variables

$$Y_{it} = t_{it} + e_{it}$$

$$t_{it} = x_{it} + z_{it}$$

Decomposition of variances

$$\text{Var}(Y_{it}) = \text{Var}(t_{it}) + \text{Var}(e_{it})$$

$$\text{Var}(t_{it}) = \text{Var}(x_{it}) + \text{Var}(z_{it})$$



## Expected Values and Covariances of the Residuals

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$$E(\mathbf{e}_{it}) = 0$$

$$E(\mathbf{z}_{it}) = 0$$

$$\text{Cov}(\mathbf{e}_{it}, \mathbf{z}_{js}) = 0$$

$$\text{Cov}(\mathbf{e}_{it}, \mathbf{t}_{js}) = 0$$

$$\text{Cov}(\mathbf{e}_{it}, \mathbf{x}_{js}) = 0$$

$$\text{Cov}(\mathbf{z}_{it}, \mathbf{x}_{js}) = 0$$



## Expected Values and Covariances of the Residuals

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*Proof of:*  $E(\mathbf{z}_{it}) = 0$

$$E(\mathbf{z}_{it}) = E(\mathbf{t}_{it} - \mathbf{x}_{it}) \quad (\text{Inserting the definition})$$

$$= E(\mathbf{t}_{it}) - E(\mathbf{x}_{it}) \quad (\text{Rule iii of Rule Box 5.1})$$

$$= E[E(Y_{it} | U, S_i)] - E[E(Y_{it} | U)] \quad (\text{Inserting the definitions})$$

$$= E(Y_{it}) - E(Y_{it}) = 0 \quad (\text{Rule iv of Rule Box 6.2})$$



## Expected Values and Covariances of the Residuals

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*Proof of:*  $Cov(\mathbf{z}_{it}, \mathbf{x}_{js}) = 0$

First we proof  $E(\mathbf{z}_{it} | \mathbf{x}_{js}) = 0$ , which implies  $Cov(\mathbf{z}_{it}, \mathbf{x}_{js}) = 0$ .

$$\begin{aligned}\mathbf{z}_{it} &= \mathbf{t}_{it} - \mathbf{x}_{it} && \text{(Definition of } \mathbf{z}_{it}\text{)} \\ &= \mathbf{t}_{it} - E(Y_{it} | U) && \text{(Definition of } \mathbf{x}_{it}\text{)} \\ &= \mathbf{t}_{it} - E[E(Y_{it} | U, S_t) | U] && \text{(Rule vi of Rule Box 6.2)} \\ &= \mathbf{t}_{it} - E(\mathbf{t}_{it} | U) && \text{(Definition of } \mathbf{t}_{it}\text{)}\end{aligned}$$

This shows that  $\mathbf{z}_{it}$  is a *residual* with respect to the regressor  $U$ . Hence, its regression on all functions of  $U$  is zero and  $\mathbf{x}_{js} := E(Y_{js} | U)$  is such a function of  $U$ . Furthermore, its covariances with all functions of  $U$  are zero, because the covariances of a residual with all functions of its regressor are zero.



## Important Coefficients

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Reliability

$$Rel(Y_{it}) = \frac{Var(\mathbf{t}_{it})}{Var(Y_{it})} = Con(Y_{it}) + Spe(Y_{it})$$

Consistency

$$Con(Y_{it}) = \frac{Var(\mathbf{x}_{it})}{Var(Y_{it})}$$

Occasion Specificity

$$Spe(Y_{it}) = \frac{Var(\mathbf{z}_{it})}{Var(Y_{it})}$$

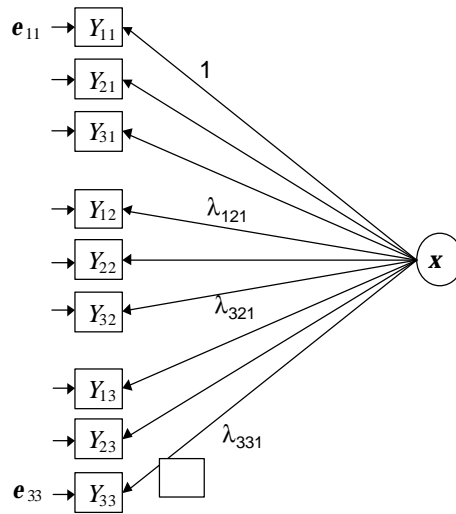
Stability of the latent state variable  $Kor(\mathbf{t}_{it}, \mathbf{t}_{is})$

Stability of the latent trait variable  $Kor(\mathbf{x}_{it}, \mathbf{x}_{is})$



## Singletrait Model Path Diagram

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## Singletrait Model Definition

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$$Y_{it} = \mathbf{t}_{it} + \mathbf{e}_{it}$$
$$= \lambda_{it0} + \lambda_{it1} \mathbf{x} + \mathbf{e}_{it}$$
$$\text{Cov}(\mathbf{e}_{it}, \mathbf{e}_{js}) = 0 \quad (i, t) \neq (j, s)$$



## Singletrait Model

### *Fixing the scale and Identification*

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Fixing the scale of the latent trait variable  $\mathbf{x}$ :

$$E(\mathbf{x}) = 0 \quad \text{and} \quad \text{Var}(\mathbf{x}) = 1$$

or 
$$\lambda_{110} = 0 \quad \text{and} \quad \lambda_{111} = 1$$

Identification in the case  $\lambda_{110} = 0$  and  $\lambda_{111} = 1$

$$E(\mathbf{x}) = E(Y_{11})$$

$$\lambda_{it1}^2 \text{Var}(\mathbf{x}) = \frac{\text{Cov}(Y_{it}, Y_{js}) \text{Cov}(Y_{it}, Y_{ku})}{\text{Cov}(Y_{js}, Y_{ku})}, \quad (i, t) \neq (j, s) \neq (k, u)$$

$$\text{Var}(\mathbf{t}_{it}) = \lambda_{it1}^2 \text{Var}(\mathbf{x})$$

$$\text{Con}(Y_{it}) = \text{Rel}(Y_{it}) = \frac{\lambda_{it1}^2 \text{Var}(\mathbf{x})}{\text{Var}(Y_{it})}$$



## Singletrait Model

### *Testability*

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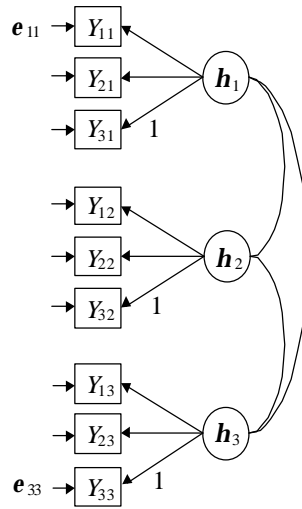
$$\begin{aligned} \text{Var}(Y_{it}) &= \text{Var}(\mathbf{t}_{it}) + \text{Var}(\mathbf{e}_{it}) \\ &= \lambda_{it1}^2 \text{Var}(\mathbf{x}) + \text{Var}(\mathbf{e}_{it}) \end{aligned}$$

$$\text{Cov}(Y_{it}, Y_{js}) = \text{Cov}(\mathbf{t}_{it}, \mathbf{t}_{js}) = \lambda_{it1} \lambda_{js1} \text{Var}(\mathbf{x}), \quad (i, t) \neq (j, s)$$



## Multistate Model *path diagram*

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## Multistate Model *Definition*

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$$Y_{it} = \mathbf{t}_{it} + \mathbf{e}_{it}$$

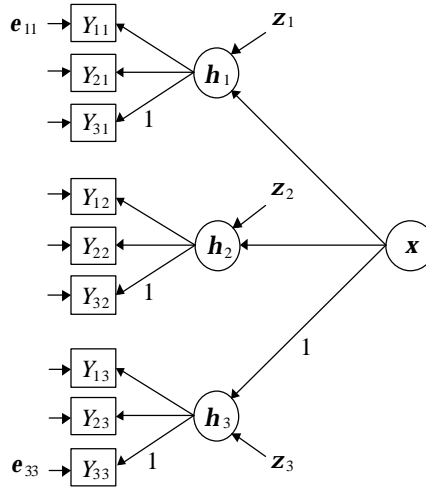
$$\mathbf{t}_{it} = \lambda_{i0} + \lambda_{i1} \mathbf{h}_t + \mathbf{e}_{it}$$

$$\text{Cov}(\mathbf{e}_{it}, \mathbf{e}_{js}) = 0 \quad (i, t) \neq (j, s)$$



## Multistate-Singletrait Model *path diagram*

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## Multistate-Singletrait Model *Definition*

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$$Y_{it} = \mathbf{t}_{it} + \mathbf{e}_{it}$$
$$= \lambda_{i0} + \lambda_{it1} \mathbf{h}_t + \mathbf{e}_{it}$$

$$\mathbf{h}_t = \gamma_{t0} + \gamma_{t1} \mathbf{x} + \mathbf{z}_t$$

$$\text{Cov}(\mathbf{e}_{it}, \mathbf{e}_{js}) = 0 \quad (i, t) \neq (j, s)$$

$$\text{Cov}(\mathbf{e}_{it}, \mathbf{h}_s) = 0$$

$$\text{Cov}(\mathbf{z}_t, \mathbf{z}_s) = 0$$

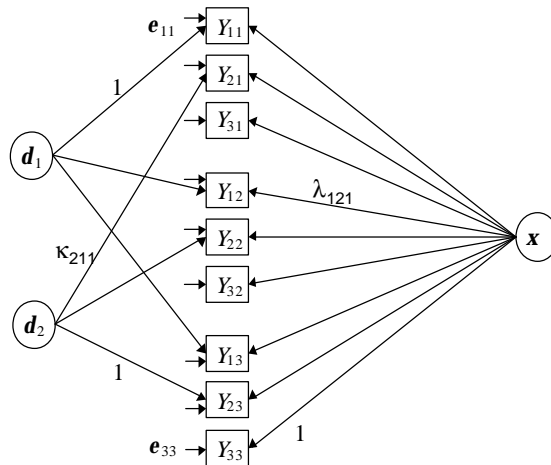
$$\text{Cov}(\mathbf{z}_t, \mathbf{e}_{js}) = 0$$

$$\text{Cov}(\mathbf{z}_t, \mathbf{x}) = 0$$



## Singletrait Model with Method Factors *path diagram*

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## Singletrait Model with Method Factor *Definition*

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$$Y_{it} = \mathbf{t}_{it} + \mathbf{e}_{it}$$
$$= \lambda_{it0} + \lambda_{it1}\mathbf{x} + \mathbf{d}_i + \mathbf{e}_{it}$$

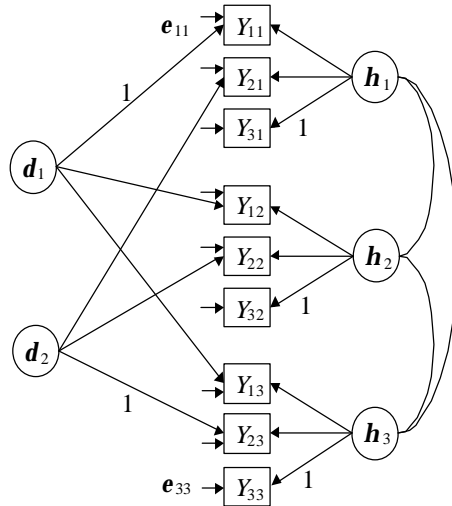
$$Y_{33} = \mathbf{t}_{33} + \mathbf{e}_{33}$$
$$= \mathbf{x} + \mathbf{e}_{33}$$

$$\text{Cov}(\mathbf{e}_{it}, \mathbf{e}_{js}) = 0 \quad (i, t) \neq (j, s)$$



Multistate Model with Method Factors  
*path diagram*

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Singletrait-Multistate Model with Method Factors  
*path diagram*

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