

Data Example

A simple example is fraction subtraction,
analyzed in K. Tatsuoka (1990) and C.
Tatsuoka (2002).

Let's consider what one must master to
subtract fractions.

Attributes

According to Mislavy (1996).

- (1) performing basic fraction subtraction operation
- (2) simplifying/reducing
- (3) separating whole number from fraction
- (4) borrowing one from whole number to fraction
- (5) converting whole number to fraction

Latent Variable Models

Notation:

\mathbf{Y} denotes a vector of binary responses for the J items of an examination.

$\boldsymbol{\alpha} = (\alpha_1, \alpha_2, \dots, \alpha_K)'$. The k th element, α_k , of $\boldsymbol{\alpha}$ is a binary indicator of an examinee's classification with regard to the k th attribute.

Components of \mathbf{Y} are independent given $\boldsymbol{\alpha}$.

$\mathbf{Q} = [q_{jk}]$, where element q_{jk} of matrix \mathbf{Q} indicates whether k th skill is required for j th item.

NIDA Model

Let η_{ijk} indicate whether the i th subject correctly applied the k th attribute in completing the j th item.

$$s_k = P(\eta_{ijk} = 0 \mid \alpha_{ik} = 1, q_{jk} = 1)$$

$$g_k = P(\eta_{ijk} = 1 \mid \alpha_{ik} = 0, q_{jk} = 1)$$

$$Y_{ij} = \prod_{k=1}^K \eta_{ijk}$$

$$P(Y_{ij} = 1 \mid \boldsymbol{\alpha}_i, \boldsymbol{s}, \boldsymbol{g}) = \prod_{k=1}^K P(\eta_{ijk} = 1 \mid \alpha_{ik}, s_k, g_k)$$

$$= \prod_{k=1}^K \left[(1 - s_k)^{\alpha_{ik}} g_k^{1-\alpha_{ik}} \right]^{q_{jk}}.$$

$$L(\boldsymbol{s}, \boldsymbol{g}; \boldsymbol{\alpha}) =$$

$$\prod_{i=1}^N \prod_{j=1}^J \left\{ \prod_{k=1}^K \left[(1 - s_k)^{\alpha_{ik}} g_k^{1-\alpha_{ik}} \right]^{q_{jk}} \right\}^{Y_{ij}} \left\{ 1 - \prod_{k=1}^K \left[(1 - s_k)^{\alpha_{ik}} g_k^{1-\alpha_{ik}} \right]^{q_{jk}} \right\}^{1-Y_{ij}}.$$

The DINA Model

Let η_{ij} denote whether the i th examinee possesses the attributes required for the j th item.

$$\eta_{ij} = \prod_{k=1}^K \alpha_{ik}^{q_{jk}}.$$

The parameter s_j refers to the probability of slipping and incorrectly answering the item when $\eta_{ij} = 1$, and g_j is the probability of correctly guessing the answer when $\eta_{ij} = 0$.

$$P(Y_{ij} = 1 \mid \boldsymbol{\alpha}) = (1 - s_j)^{\eta_{ij}} g_j^{1 - \eta_{ij}}.$$

$$L(\boldsymbol{s}, \boldsymbol{g}; \boldsymbol{\alpha}) =$$

$$\prod_{i=1}^N \prod_{j=1}^J \left[(1 - s_j)^{y_{ij}} s_j^{1 - y_{ij}} \right]^{\eta_{ij}} \left[g_j^{y_{ij}} (1 - g_j)^{1 - y_{ij}} \right]^{1 - \eta_{ij}}.$$

The Fusion Model

$$P(Y_{ij} = 1 | \boldsymbol{\alpha}, \pi_j^*, r_{jk}^*, c_j, \eta_i)$$

$$= \pi_j^* P_{c_j}(\eta_i) \prod_{k=1}^K r_{jk}^{*(1-\alpha_{ik})q_{jk}}$$

$P_{c_j}(\eta)$ is a Rasch IRF with parameter c_j

Let Y_{ijk} be a latent response denoting whether examinee i correctly applies a required attribute k to item j .

$$\pi_j^* = \prod_{k=1}^K P[Y_{ijk} = 1 | \alpha_{ik} = 1]$$

$$r_{jk}^* = P[Y_{ijk} = 1 | \alpha_{ik} = 0] / P[Y_{ijk} = 1 | \alpha_{ik} = 1]$$