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Solutions to the exercises of Chapter 10

- Exercise 10.1
The challenge in this exercise is that you must see that a Rasch model has to be fit to the replicated data sets too in order to be able to compute the Q_3 index.
- Exercise 10.2
Call δ_2 the pairwise interaction parameter and δ_3 the three-way interaction parameter. Then,

$$W_p = \begin{pmatrix} 0 & 0 \\ y_{p1} & 0 \\ y_{p1} + y_{p2} & y_{p1}y_{p2} \\ y_{p1} + y_{p2} + y_{p3} & y_{p1}y_{p2} + y_{p1}y_{p3} + y_{p2}y_{p3} \\ y_{p1} + y_{p2} + y_{p3} + y_{p4} & y_{p1}y_{p2} + y_{p1}y_{p3} + y_{p1}y_{p4} \\ & + y_{p2}y_{p3} + y_{p2}y_{p4} + y_{p3}y_{p4} \end{pmatrix}.$$

Define $t_1 = \sum_{j=1}^{i-1} y_{pj}$ and $t_2 = \sum_{j=1}^{i-2} \sum_{k=1}^{i-1} y_{pj}y_{pk}$. Interpretation:

Holding everything else constant, the odds of giving a 1-response change with factor $e^{\delta_2(t_1 - t_1^*)}$ if t_1 instead of t_1^* previous successes are achieved. For δ_3 a similar interpretation holds, but then for an increase in previous number of pairs of successes. Note however that the “holding everything constant” requirement is somewhat artificial here because if t_1 changes, t_2 is likely to change as well (but not automatically).

- Exercise 10.3
For two items:

$$\begin{aligned} \Pr(Y_1 = 1) &= \sum_{y=0}^1 \Pr(Y_1 = 1, Y_2 = y) \\ &= \frac{\exp(\theta_p - \beta_1) + \exp(2\theta_p - \beta_1 - \beta_2 + \delta)}{1 + \exp(\theta_p - \beta_1) + \exp(\theta_p - \beta_2) + \exp(2\theta_p - \beta_1 - \beta_2 + \delta)} \end{aligned}$$

- Exercise 10.4

$$\begin{aligned}
 \Pr(Y_1 = 1, Y_2 = y) &= \int \Pr(Y_1 = 1, Y_2 = y | \theta_p) dF(\theta_p) \\
 &= \int \Pr(Y_1 = 1 | \theta_p) \Pr(Y_2 = y | \theta_p) dF(\theta_p) \\
 \Pr(Y_1 = 1) &= \sum_{y=0}^1 \Pr(Y_1 = 1, Y_2 = y) \\
 &= \int \Pr(Y_1 = 1 | \theta_p) dF(\theta_p)
 \end{aligned}$$

- Exercise 10.5

$$R = \left(\begin{array}{ccc|ccc}
 1 & \cdots & \rho_1 & \rho_2 & \cdots & \rho_2 \\
 \vdots & \ddots & \vdots & \vdots & \ddots & \vdots \\
 \rho_1 & \cdots & 1 & \rho_2 & \cdots & \rho_2 \\
 \rho_2 & \cdots & \rho_2 & 1 & \cdots & \rho_3 \\
 \vdots & \ddots & \vdots & \vdots & \ddots & \vdots \\
 \rho_2 & \cdots & \rho_2 & \rho_3 & \cdots & 1
 \end{array} \right),$$

with $\rho_1 = \frac{\sigma^2 + \tau^2}{\sigma^2 + \tau^2 + 1}$, $\rho_2 = \frac{\sigma^2}{\sqrt{\sigma^2 + \tau^2 + 1} \sqrt{\sigma^2 + 1}}$, and $\rho_3 = \frac{\sigma^2}{\sigma^2 + 1}$

- Exercise 10.6

$$\begin{aligned}
 \Pr(Y_1 = y_1) &= \frac{\exp(y_1(\theta - \beta_1))}{1 + \exp(\theta - \beta_1)} \\
 \Pr(Y_2 = y_2 | y_1) &= \frac{\exp(y_2(\theta - \beta_2 + y_1 \delta_1))}{1 + \exp(\theta - \beta_2 + y_1 \delta_1)}
 \end{aligned}$$

and

$$\begin{aligned}
 \Pr(Y_3 = y_3, Y_4 = y_4 | y_1, y_2) &= \\
 &= \frac{\exp(y_3(\theta - \beta_3 + s \delta_1) + y_4(\theta - \beta_4 + s \delta_1) + y_3 y_4 \delta_2)}{1 + \exp(\theta - \beta_3 + s \delta_1) + \exp(\theta - \beta_4 + s \delta_1) + \exp(2\theta - \beta_3 - \beta_4 + 2s \delta_1 + \delta_2)}
 \end{aligned}$$

where $s = y_1 + y_2$.

- Exercise 10.7

Design matrix for the normal child:

$$\mathbf{X} = \begin{pmatrix}
 1 & 0 & 0 & 0 & 0 & 0 & 0 \\
 0 & 1 & 0 & 0 & 0 & y_1 & 0 \\
 0 & 0 & 1 & 0 & 0 & y_1 + y_2 & 0 \\
 0 & 0 & 0 & 1 & 0 & y_1 + y_2 + 2 + y_3 & 0 \\
 0 & 0 & 0 & 0 & 1 & y_1 + y_2 + y_3 + y_4 & 0
 \end{pmatrix}$$

Design matrix for the disabled child:

$$\mathbf{X} = \begin{pmatrix} 1 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 & 0 & y_1 \\ 0 & 0 & 1 & 0 & 0 & 0 & y_1 + y_2 \\ 0 & 0 & 0 & 1 & 0 & 0 & y_1 + y_2 + y_3 \\ 0 & 0 & 0 & 0 & 1 & 0 & y_1 + y_2 + y_3 + y_4 \end{pmatrix}$$

- Exercise 10.8

The model can be constructed by simply adding θ_{p1} to the exponents in all cumulative probability equations pertaining to the items involved in the testlet.