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# Part 6: Testing the Rasch Model (II)

Martin-Löf Test and some nonparametric ('exact') Tests

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Testing the Rasch Model

Martin-Löf Test

 $LR_{ML} = 2(\sum_{r} n_{r} \ln(\frac{n_{r}}{N}) - \sum_{r} n_{r} \ln(\frac{n_{r}}{N}) - \ln L_{c} + \ln L_{c}^{(1)} + \ln L_{c}^{(2)})$ 

 $LR_{ML}$  has an asymptotic  $\chi^2\text{-distribution}$  with  $d\!f$  =  $k_1k_2$  – 1

> MLoef(rmod, splitcr = "median")
Martin-Loef-Test (split criterion: median)
LR-value: 8.009
Chi-square df: 8
p-value: 0.433

> MLoef(rmod, splitcr = c(1, 1, 1, 1, 0, 0))
Martin-Loef-Test (split criterion: user-defined)
LR-value: 7.934
Chi-square df: 7
p-value: 0.338

Testing the Rasch Model

### Martin-Löf Test

evaluate unidimensionality test whether two sets of items form a Rasch scale

#### Idea:

similar to Andersen's LR Test – focus now on items the items are partitioned into two subsets of  $k_1$  and  $k_2$  items

$$LR_{ML} = 2\ln\left(\frac{L_c}{L_c^{(1)} \cdot L_c^{(2)}} \cdot q\right) \qquad q = \frac{\prod_r (\frac{n_r}{N})^{n_r}}{\prod_r (\frac{n_r}{N})^n}$$

product of likelihoods for subsets the same as for whole set

 $r = (r_1, r_2)'$ , score patterns on the two subtests  $(r_1 = 0, ..., k_1$  and  $r_2 = 0, ..., k_2$ ),  $n_r$  is number of persons obtaining pattern r $r = r_1 + r_2$  (sum score),  $n_r$  accordingly

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Nonparametric ('exact') Tests

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#### Nonparametric ('exact') Tests

Idea:

Parameter estimates depend only on the marginal totals r and s if the Rasch model fits the data, all binary matrices with the same marginals are equally likely

for any statistic of the data matrix, one can approximate the null distribution (i.e., the distribution if the Rasch model is valid)

take a random sample from the collection of equally likely data matrices construct the observed distribution of the statistic

one can then simply determine the exceedence probability of the statistic in the observed sample (its p-value) and thus construct a nonparametric test of the Rasch model

Nonparametric ('exact') Tests	Nonparametric ('exact') Tests	ų
Nonparametric Tests (cont'd)	<pre>using eRm: rsampler()</pre>	
sample space: $\Sigma_{rs}$ (all possible matrices with fixed $r$ and $s$ )	(from package RaschSampler)	
the distribution is uniform over of $\Sigma_{rs}$ .	<pre>&gt; rmat &lt;- rsampler(stress, rsctrl(burn_in = 100,</pre>	n_eff = 100, seed = 123
for each data matrix $X$ , $X \in \Sigma_{rs}$ : $p(X) = \frac{1}{\#\Sigma_{rs}}$	<pre>&gt; summary(rmat) Status of object rmat after call to RSampler:</pre>	
$\#\Sigma_{rs}$ can be huge	k = 6 burn_in = 100	
for a $12 \times 12$ table with all $r = s = 2$ this is 21,959,547,410,077,2		
algorithmic difficulty is to draw uniform random samples	tfixed = FALSE n_tot = 101 outvec contains 10100 elements	
three approaches:	outvec contains foroo elements	
<ul> <li>Ponocny(2001) (simple Monte Carlo)</li> </ul>		
<ul> <li>Chen(2005) (importance sampling)</li> </ul>		
– Verhelst, Hatzinger & Mair (2007) (MCMC, in eRm)		
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Nonparametric ('exact') Tests	Nonparametric ('exact') Tests	1
Nonparametric Tests	Local Dependence	
Procedure:	$\mathbf{T_1}$ : checks for local dependence via increa	
• $X_0$ observed data matrix	tions for item pairs: cases are counted w both items.	ith equal responses o
• simulate $n$ Rasch-conform matrices $X_1,\ldots,X_n$ with fin	here here here here here here here here	m. O otherwise
margins as in $X_0$	$T_1(X) = \sum_{v} \delta_{ij}(x_{vi}, x_{vj}),  \delta() = 1 \text{ if } x_{vi} = 1$	
margins as in $X_0$ • calculate a test-statistics $T^{(0)}$ for $X_0$ • calculate $T^{(1)}, \ldots, T^{(2)}$ for $X_1, \ldots, X_n$	$\mathbf{T_2}$ : checks for local dependence to detect scales via increased dispersion of subscale	t model deviating sul
• calculate a test-statistics $T^{(0)}$ for $X_0$	$\mathbf{T_2}$ : checks for local dependence to detec	t model deviating sul
• calculate a test-statistics $T^{(0)}$ for $X_0$ • calculate $T^{(1)}, \ldots, T^{(2)}$ for $X_1, \ldots, X_n$	$ \begin{array}{c} \mathbf{T_2:} \ \text{checks for local dependence to detec} \\ \text{scales via increased dispersion of subscale} \\ T_2(X) = Var_v(r_v^{(\mathcal{S})}), \ \text{where } r_v^{(\mathcal{S})} = \sum_{i \in \mathcal{S}} x_{vi} \end{array} $	t model deviating sul
• calculate a test-statistics $T^{(0)}$ for $X_0$ • calculate $T^{(1)}, \ldots, T^{(2)}$ for $X_1, \ldots, X_n$ Test:	$ \begin{array}{c} \mathbf{T_2:} \ \text{checks for local dependence to detec} \\ \text{scales via increased dispersion of subscale} \\ T_2(X) = Var_v(r_v^{(\mathcal{S})}), \ \text{where } r_v^{(\mathcal{S})} = \sum_{i \in \mathcal{S}} x_{vi} \end{array} $	t model deviating sul
• calculate a test-statistics $T^{(0)}$ for $X_0$ • calculate $T^{(1)}, \ldots, T^{(2)}$ for $X_1, \ldots, X_n$ Test: • count number $x$ of the $T^{(.)}$ 's which exceed $T_0$ giving $p = x$ • if $p \le \alpha$ reject $H_0$	$T_2: \text{ checks for local dependence to detect scales via increased dispersion of subscales T_2(X) = \operatorname{Var}_v(r_v^{(\mathcal{S})}), \text{ where } r_v^{(\mathcal{S})} = \sum_{i \in \mathcal{S}} x_{vi}$	t model deviating sub person rawscores. pattern $r'$ 10100 2
• calculate a test-statistics $T^{(0)}$ for $X_0$ • calculate $T^{(1)}, \ldots, T^{(2)}$ for $X_1, \ldots, X_n$ Test: • count number $x$ of the $T^{(.)}$ 's which exceed $T_0$ giving $p = x$ • if $p \le \alpha$ reject $H_0$ Advantage:	$T_{2}: \text{ checks for local dependence to detect scales via increased dispersion of subscales T_{2}(X) = \operatorname{Var}_{v}(r_{v}^{(\mathcal{S})}), \text{ where } r_{v}^{(\mathcal{S})} = \sum_{i \in \mathcal{S}} x_{vi}$	t model deviating sub person rawscores. pattern $r'$ 10100 2 10011 3
• calculate a test-statistics $T^{(0)}$ for $X_0$ • calculate $T^{(1)}, \ldots, T^{(2)}$ for $X_1, \ldots, X_n$ Test: • count number $x$ of the $T^{(\cdot)}$ 's which exceed $T_0$ giving $p = x$ • if $p \le \alpha$ reject $H_0$ Advantage: - arbitrary specific tests statistics can be constructed	$T_{2}: \text{ checks for local dependence to detect scales via increased dispersion of subscale} T_{2}(X) = \operatorname{Var}_{v}(r_{v}^{(\mathcal{S})}), \text{ where } r_{v}^{(\mathcal{S})} = \sum_{i \in \mathcal{S}} x_{vi}$	t model deviating sul person rawscores. pattern $r'$ 10100 2 10011 3 01100 2
• calculate a test-statistics $T^{(0)}$ for $X_0$ • calculate $T^{(1)}, \ldots, T^{(2)}$ for $X_1, \ldots, X_n$ Test: • count number $x$ of the $T^{(.)}$ 's which exceed $T_0$ giving $p = x$ • if $p \le \alpha$ reject $H_0$ Advantage:	$T_{2}: \text{ checks for local dependence to detect scales via increased dispersion of subscales T_{2}(X) = \operatorname{Var}_{v}(r_{v}^{(\mathcal{S})}), \text{ where } r_{v}^{(\mathcal{S})} = \sum_{i \in \mathcal{S}} x_{vi}$	t model deviating sub person rawscores. pattern $r'$ 10100 2 10011 3
• calculate a test-statistics $T^{(0)}$ for $X_0$ • calculate $T^{(1)}, \ldots, T^{(2)}$ for $X_1, \ldots, X_n$ Test: • count number $x$ of the $T^{(\cdot)}$ 's which exceed $T_0$ giving $p = x$ • if $p \le \alpha$ reject $H_0$ Advantage: - arbitrary specific tests statistics can be constructed - valid and powerful, even in small samples	$T_{2}: \text{ checks for local dependence to detect scales via increased dispersion of subscales} T_{2}(X) = \operatorname{Var}_{v}(r_{v}^{(\mathcal{S})}), \text{ where } r_{v}^{(\mathcal{S})} = \sum_{i \in \mathcal{S}} x_{vi}$	t model deviating sub person rawscores. pattern $r'$ 10100 2 10011 3 01100 2 01101 3

#### ٩ (A) Nonparametric ('exact') Tests Nonparametric ('exact') Tests using eRm: NPtest() $T_2$ : > t21 <- NPtest(stress, n = 100, method = "T2", idx = 1:3) $T_1$ : > t21Nonparametric RM model test: T2 (local dependence - model deviating subscales) > t1 <- NPtest(stress, n = 100, method = "T1") (dispersion of subscale person rawscores) > t1 Nonparametric RM model test: T1 (local dependence - increased Number of sampled matrices: 100 inter-item correlations) Items in subscale: 1 2 3 (counting cases with equal responses on both items) Statistic: variance one-sided p-value: 0.09 Number of sampled matrices: 100 > t22 <- NPtest(stress, n = 100, method = "T2", idx = c(1, 2, 6), stat = "mad1") Number of Item-Pairs tested: 15 > t22 Nonparametric RM model test: T2 (local dependence - model deviating Item-Pairs with one-sided p < 0.05subscales) (dispersion of subscale person rawscores) (1,3) (3,5)0.01 0.04 Number of sampled matrices: 100 Items in subscale: 1 2 6 Statistic: mean absolute deviation one-sided p-value: 0.77 Psychometric Methods 2010/11 9 Psychometric Methods 2010/11 10 ٢ ۲ Nonparametric ('exact') Tests Nonparametric ('exact') Tests DIF using eRm: NPtest() $T_4$ : checks for group anomalies (DIF) via too high (low) raw $T_{4}$ : scores on item(s) for specified group > age <- sample(20:90, 100, replace = TRUE)</pre> $T_{\mathbf{4}}(X)$ = $\sum\limits_{v \in \mathcal{G}} x_{vi}$ where $\mathcal{G}$ is any subgroup > age <- age < 30 > t41 <- NPtest(stress, n = 100, method = "T4", idx = 1:3, group = age) > t41 Nonparametric RM model test: T4 (Group anomalies - DIF) **Unequal Discrimination** (counting high raw scores on item(s) for specified group) Number of sampled matrices: 100 $T_7$ : checks for lower discrimination (2PL) in item subscale via Items in Subscale: 1 2 3 Group: age n = 10counting cases with response 1 on more difficult and 0 on easier one-sided p-value: 0.72 items. The test is global (a single statistic) for the subscale $T_{7}(X) = -\sum_{i,j \in S} \delta(x_{i} < x_{j} \land s_{i} > s_{j})$

Nonparametric ('exact') Tests

## $T_7$ :

```
> t7 <- NPtest(stress, n = 100, method = "T7", idx = 1:3)
> t7
Nonparametric RM model test: T7 (different discrimination - 2PL)
    (counting cases with response 1 on more difficult and 0 on easier item)
Number of sampled matrices: 100
Item Scores:
123
47 42 41
one-sided p-value: 0.95
these are just a few examples (we will see others later)
Ponocny (2001) gives a systematic and various more examples
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