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Response shift analysis at the item level among multiple groups using Rasch models: a simulation study

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Introduction

Integrating covariates in item-level response shift (RS) analysis could help investigating whether clinical (pathology, treatment...) or psychological (anxiety, perceived stress,...) characteristics of patients are associated with RS and changes in Patient-Reported Outcomes (PRO). However, few methods are available to analyze RS between more than two groups of patients, in particular when using Rasch Measurement Theory (RMT).

Objectives: A simulation study was performed to evaluate the performance of a method for item-level RS analysis using RMT among multiple groups of patients.

Methods: Data simulation

Longitudinal Partial-Credit model
2 times of measurement (T1;T2)
3 groups of patients: G0 (reference group), G1 & G2 (focal groups)

SIMULATION PARAMETERS

- Sample size (N): 200 or 300 (balanced groups)
- Number of items (J): 4 or 7
- Number of response categories (M): 4 or 7
- Normal distribution of the latent variable:
 - No group effect or positive group effect ($G2 > G1 > G0$)
 - No time effect or positive time effect (same PRO improvement for G1 & G2)

RS SCENARIOS

- No RS
- RS affected one focal group (G2)
- RS affected both focal groups (G1 & G2)
 - Similarly: same RS effects
 - Differentially: different magnitudes of RS effects ($G2 > G1$)
- RS affected all groups the same way

TYPE OF RS

One type per scenario, one affected item

- **Uniform recalibration (UR):**

Same shift of all item difficulties of one item

- **Non-uniform recalibration (NUR):**

Various shifts on some item difficulties of one item

336 combinations - 500 replications

Methods: Data analysis replicated on each simulated dataset

"RespOnse Shift Algorithm at the Item level"

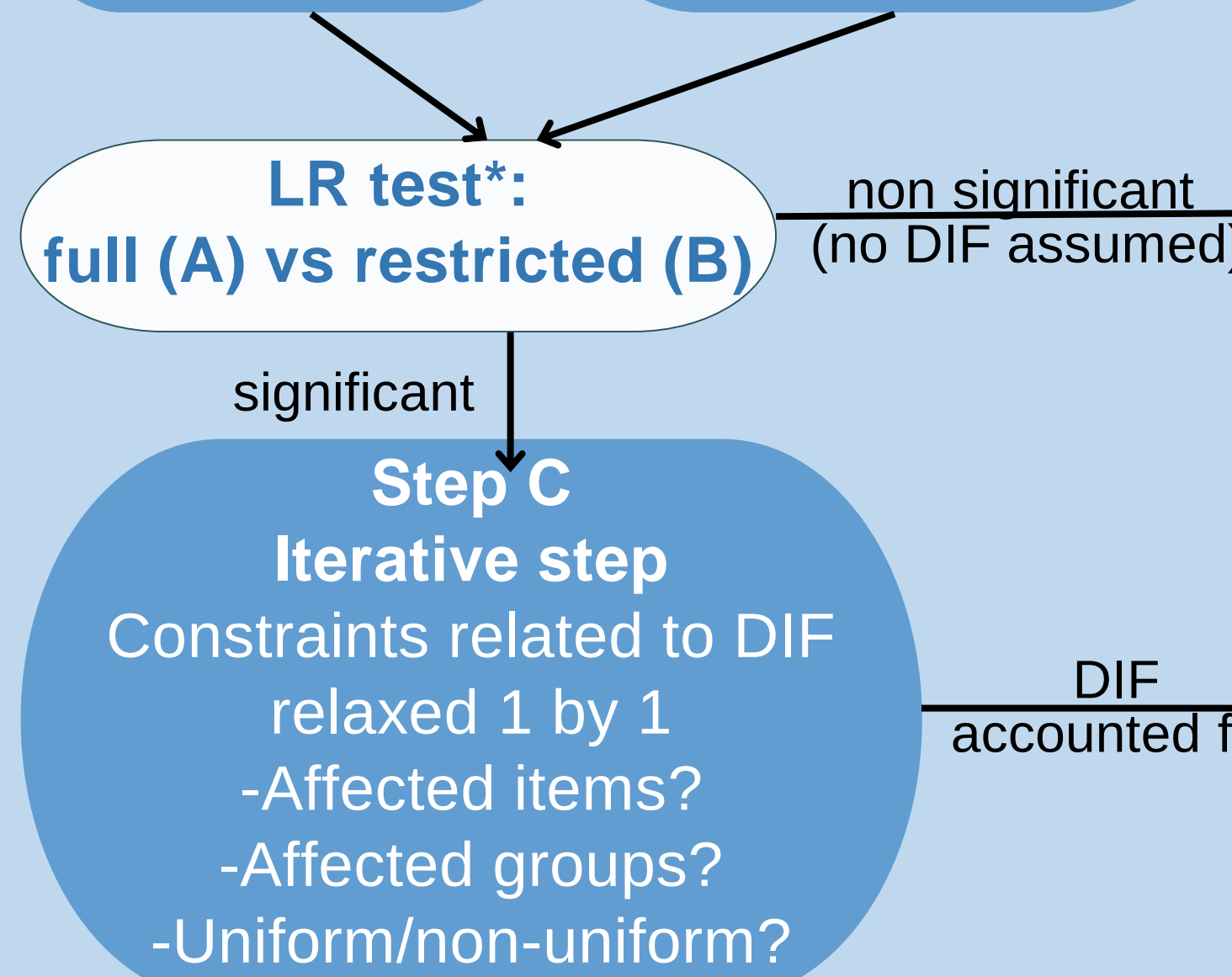
ROSALI (Hammas et al., 2020)

Stata module –rosali- on ssc

Freely accessible on <https://pro-online.net>

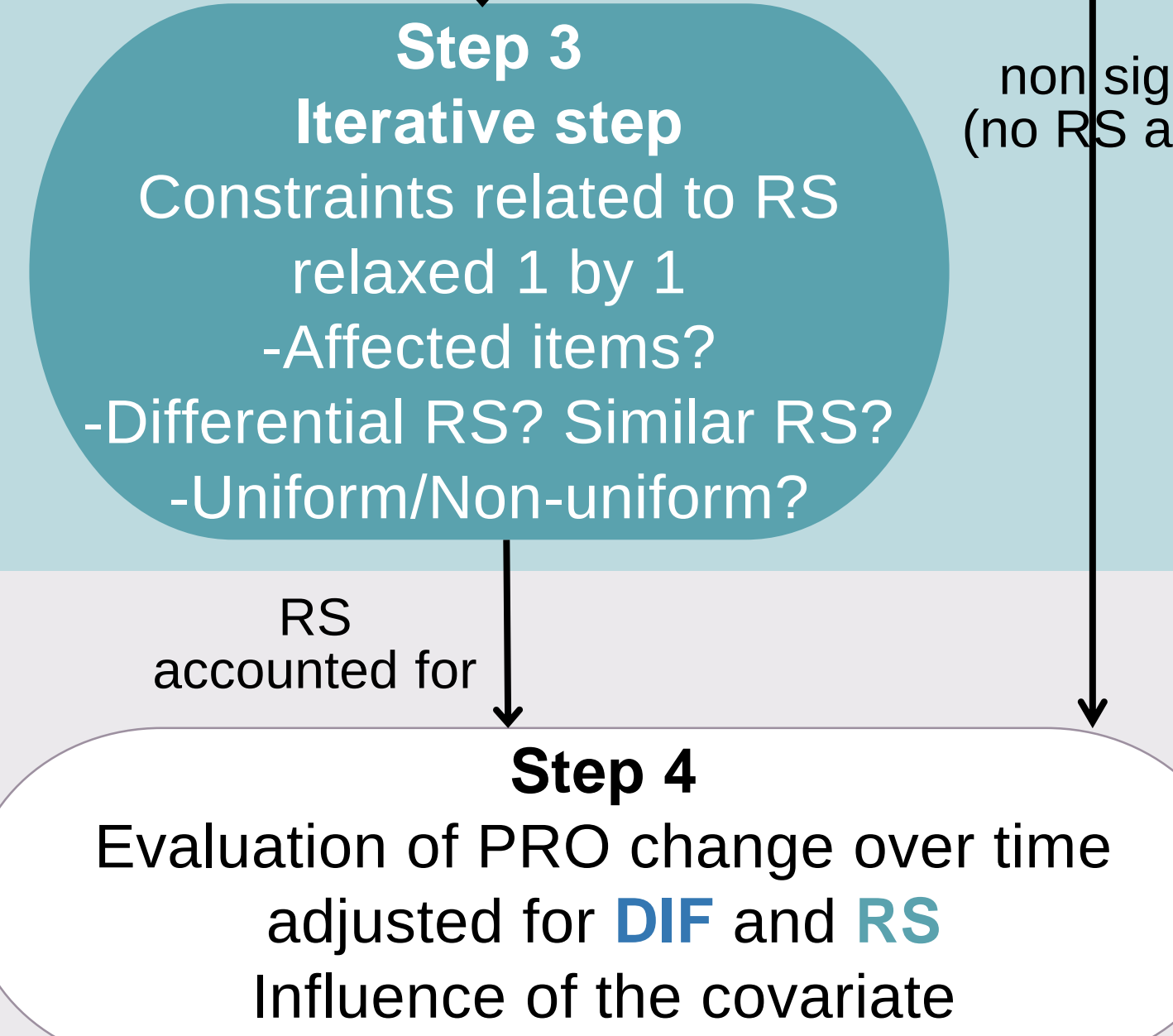
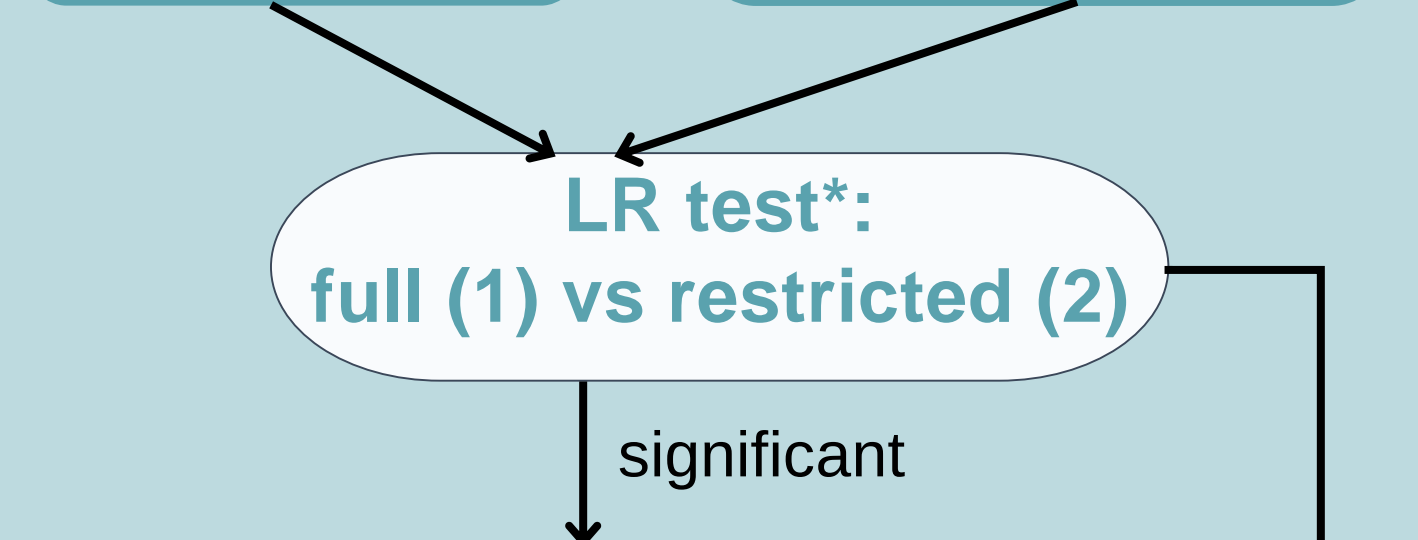
Extended to three groups of patients

DIF assessment at T1



*LR test: likelihood ratio test

RS assessment between T1 and T2



Methods: Performance criteria

Criteria assessed at the end of ROSALI

Rate of DIF detection

Proportion of datasets for which DIF effects were detected and accounted for
→ Expected to be close to 5% as no DIF was simulated in any scenario

Rate of RS detection

Proportion of datasets for which RS effects were detected and accounted for
→ Expected to be close to 5% when RS was not simulated (false detection)
→ Expected to be high when RS was simulated

False detection of DIF and RS

- All rates are lower than 3%
- Higher rates of false detection based on LR Tests: [3.6%-7.8%] for DIF, [7.2%-10.8%] for RS
→ Bonferroni correction applied in step C and step 3
→ part of false detection corrected: lower rates of false detection at the end of ROSALI (table opposite)

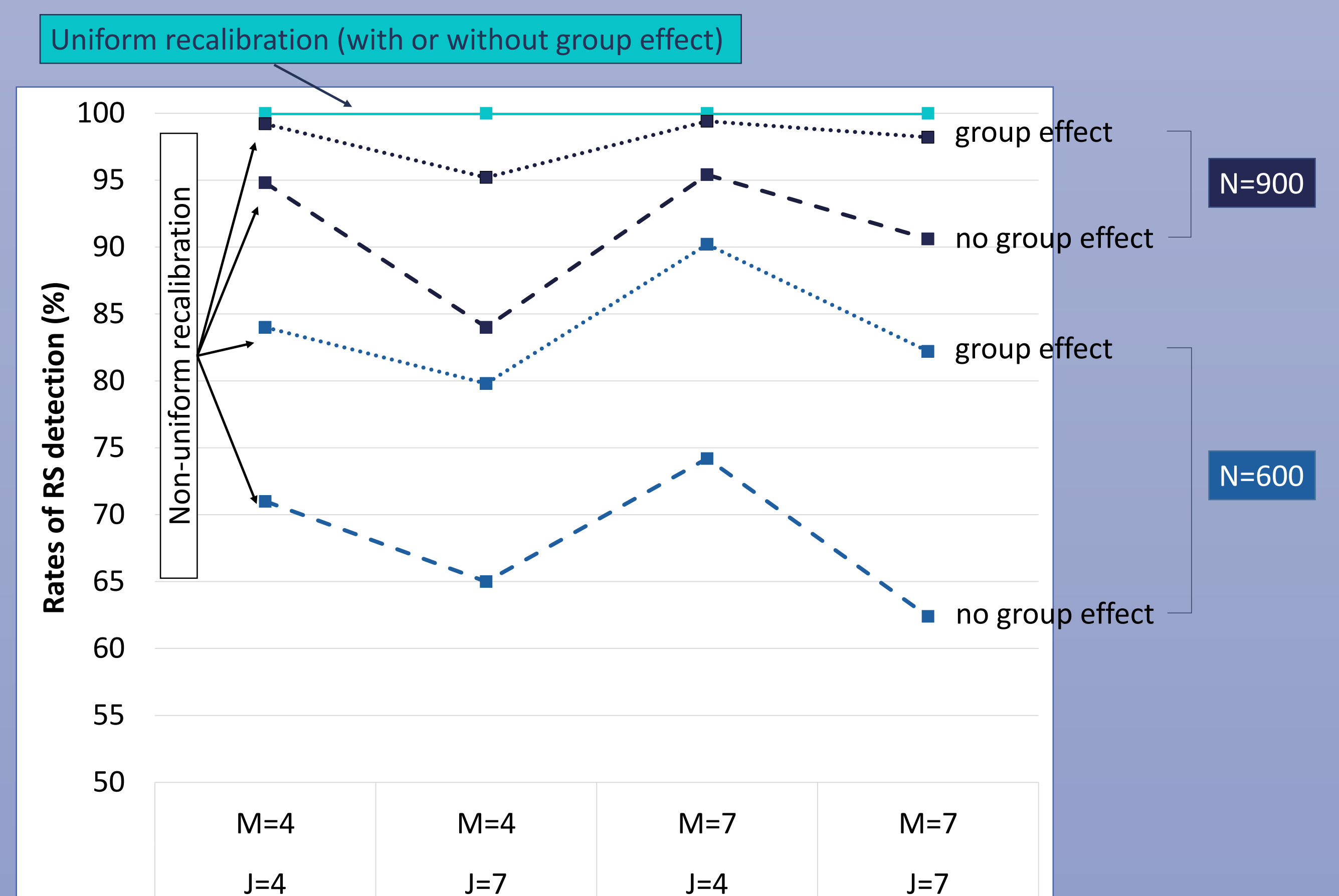
Ranges of rates of DIF and RS detection

No DIF and RS simulated

N	J	M	DIF	RS
600	4	4	1.8-2.8%	1.0-2.2%
900	4	4	1.4-2.4%	0.4-2.4%
600	7	4	0.8-2.4%	0.6-2.4%
900	7	4	0.8-2.0%	1.4-2.6%
600	4	7	1.2-2.0%	0.8-1.6%
900	4	7	1.2-2.4%	1.4-2.6%
600	7	7	1.4-2.2%	0.2-2.0%
900	7	7	0.6-1.8%	0.6-2.2%

Example of scenarios with simulated RS

RS simulated in one group – Improvement of the PRO over time



- RS detected in 100% of datasets if RS is uniform (whatever the values of the other simulated parameters)
- RS detection decreases as J increases
- RS detection increases as N or M increases or when a group effect has been simulated

Discussion

Low rates of false detection of DIF and RS

Scenarios with simulated RS

- Uniform recalibration: good performances
- Non-uniform recalibration
Best performances: RS affecting all groups similarly
Worst performances: RS affecting one group or two groups in a different way

A total sample size of 600 patients (300 per group) may be insufficient in some contexts to reach a reasonable rate of detection

Iterative steps C and 3: adjustment of p-values for multiple testing is valuable

The low rates of false detection indicates that ROSALI satisfactorily prevents from concluding erroneously to DIF or RS. The different patterns of RS inform about the required sample size to correctly detect RS depending on the expected number of affected groups.

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