

Observed Score and True Score Equating for Multidimensional Item Response Theory under Nonequivalent Group Design

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What Is Test Equating?

- **Equating** is a *statistical process* that is used to adjust scores on different test forms so that scores on the forms are comparable (Kolen & Brennan, 2004).

Five Basic Requirements of Test Equating

- Equal Constructs
- Equal reliability
- Symmetry
- Equity
- Population Invariance

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- **Symmetry**
(A \rightarrow B transformation \longleftrightarrow B \rightarrow A transformation)
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Multidimensional Item Response Theory

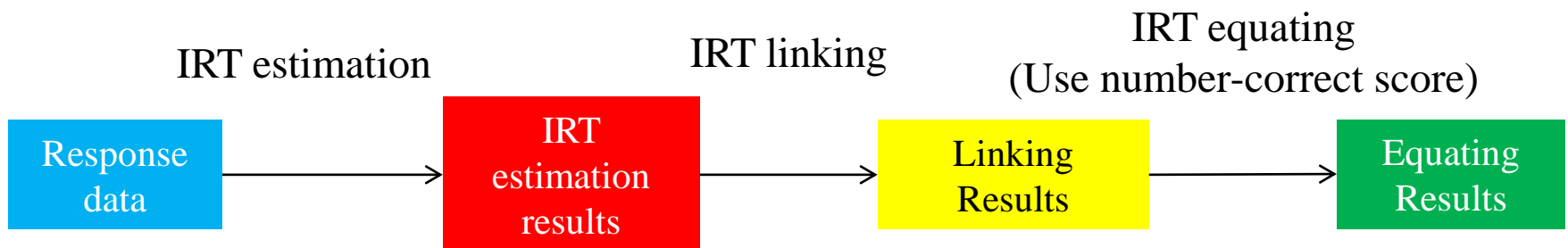
- Multidimensional Item Response Theory Model (MIRT)
 - Compensatory MIRT model (McKinley & Reckase, 1983)

$$P(x_{ij} = 1 | \boldsymbol{\theta}_j, \mathbf{a}_i, d_i) = \frac{e^{D(\mathbf{a}'_i \boldsymbol{\theta}_j + d_i)}}{1 + e^{D(\mathbf{a}'_i \boldsymbol{\theta}_j + d_i)}}$$

$\boldsymbol{\theta}_s$ represents **multiple** ability parameters associated with each respondent,
 \mathbf{a}_i represents **multiple** discrimination parameters associated with each item,
and d_i represents an item's location on an item response **surface**.

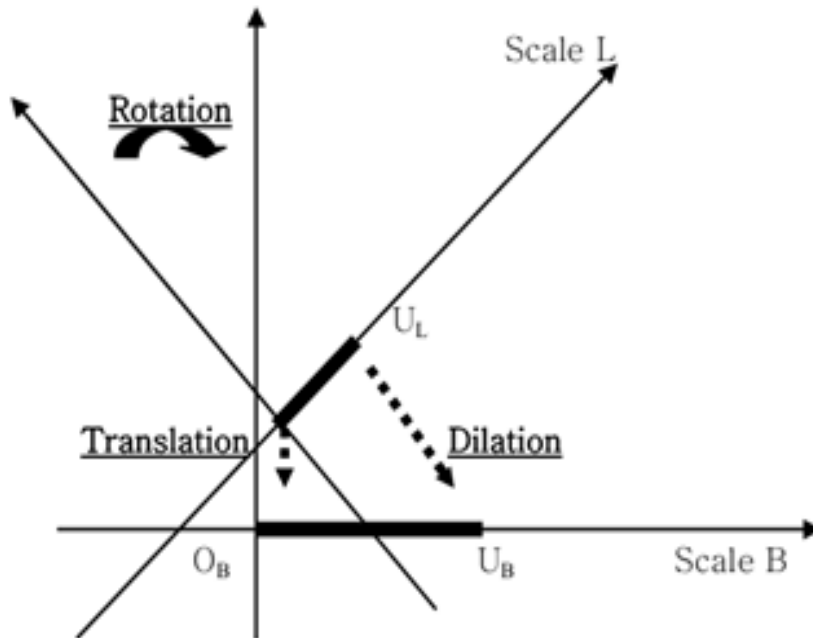
Common Procedure of Equating in IRT

- Step 1: IRT Estimation
- Step 2: IRT Linking/Scaling Aligning
- Step 3: IRT Equating (use number-correct scores, if necessary)



MIRT Linking/Scale Aligning

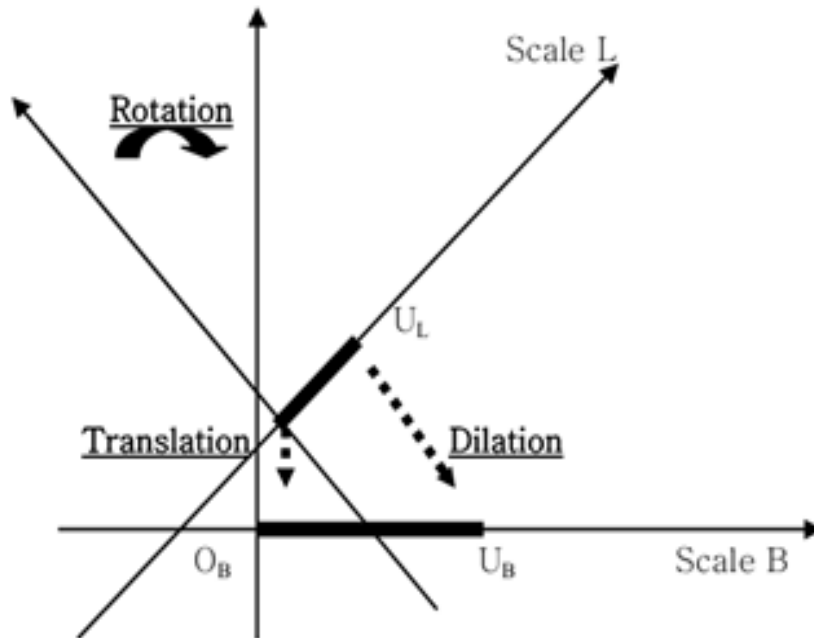
- **Dilation:** adjust unit



MIRT Linking (two-dimension case)

(Figure adapted from Min, 2003)

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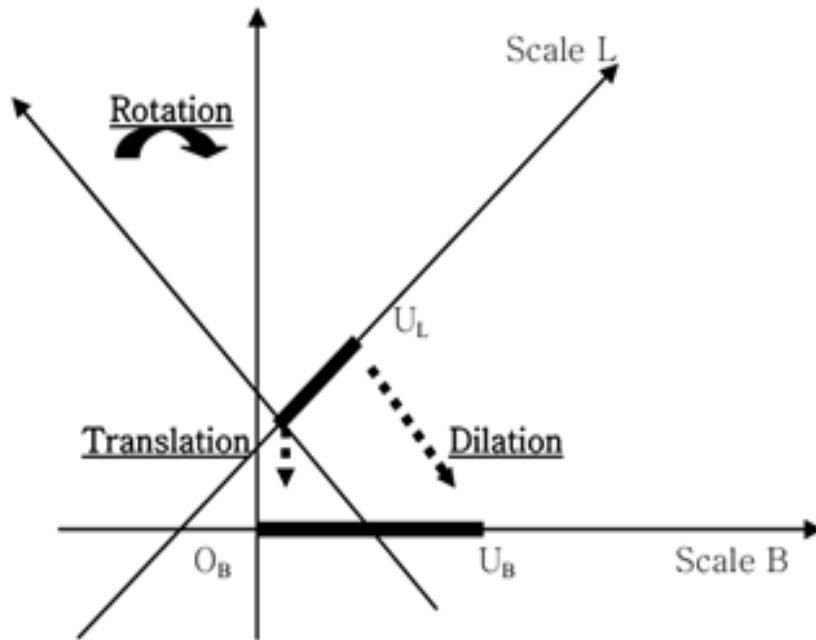


MIRT Linking (two-dimension case)

- Dilation: adjust unit
- **Translation:** adjust original zero point

(Figure adapted from Min, 2003)

MIRT Linking/Scale Aligning



MIRT Linking (two-dimension case)

- Dilation: adjust unit
- Translation: adjust original zero point
- **Rotation:** adjust the entire multidimensional axis systems so that both axis systems are in the same direction.

(Figure adapted from Min, 2003)

Symmetry Property and Unidimensionalization

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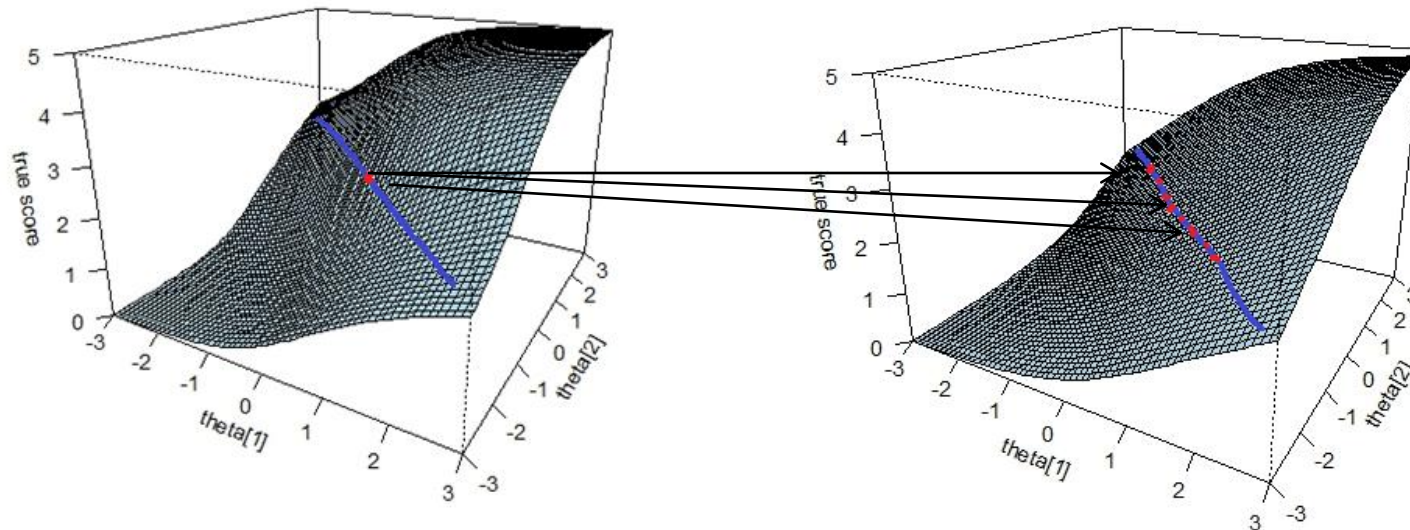
Comparability of the MIRT measure ?

Symmetry Property and Unidimensionalization (cont.)

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Symmetry Property and Unidimensionalization (cont.)

- Possible Violation of Test Equating's Symmetry Requirement
 - If we use the MIRT ability estimate **vector** as a measure of ability, a particular true score ($\tau(\theta) = \sum p(\theta)$) for one test form (i.e., test A) on the test characteristic surface (TCS), corresponds to infinite numbers of combinations of ability vectors on the other test form's TCS equiprobable contour (i.e., Test B) when both test forms are already in the same scale.



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- Unidimensionalization process devectorizes the vector or multidimensional features in the MIRT framework so that the ability measures from different test forms are comparable..
- Most importantly, through the process of unidimensionalization, the symmetry property of equating (Lord, 1980) for two test forms under MIRT framework is satisfied.

A Methodology Foundation of Unidimensionalization

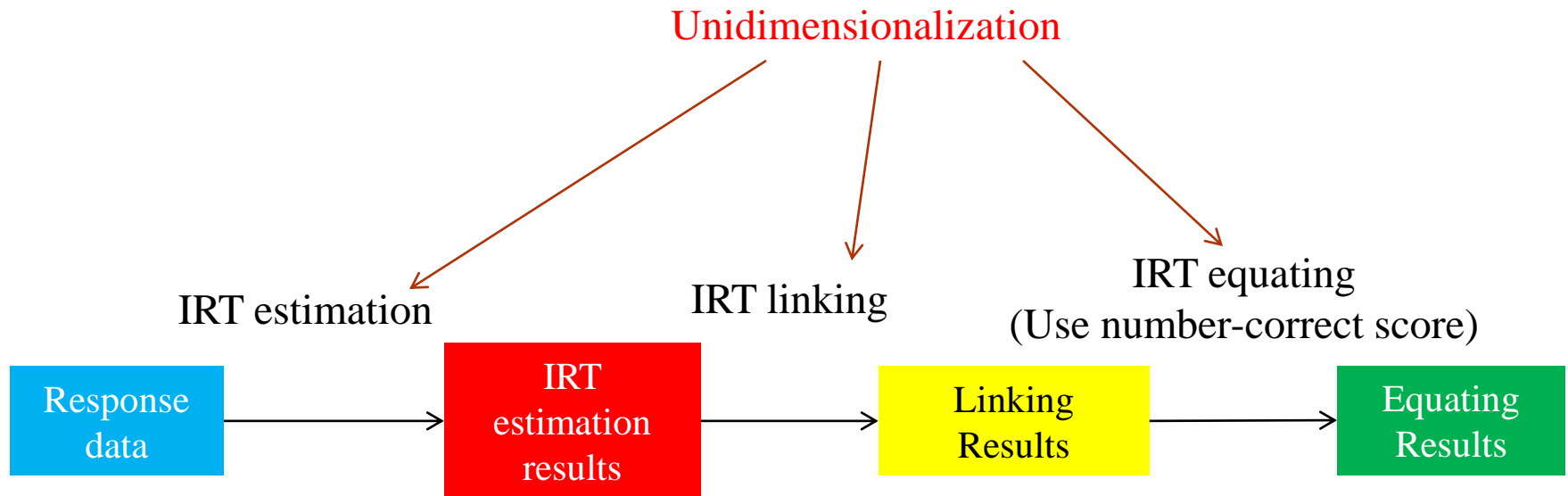
Unidimensional Approximation of MIRT (Zhang & Stout, 1999)

- Any set of item responses adequately modeled by a MIRT model, can be closely approximated by a **unidimensional IRT model** with estimated unidimensional ability composite (Θ_α) and estimated unidimensional item parameters ($\hat{a}_{\alpha j}$, $\hat{b}_{\alpha j}$, $\hat{T}_{\alpha j}$) (Zhang & Stout, 1999).
- The ability composite Θ_α of the multidimensional ability vector (i.e., $\Theta = [\theta_1, \theta_2, \dots, \theta_m]$) is defined as

$$\Theta_\alpha = \hat{\mathbf{a}}^T \hat{\boldsymbol{\theta}} = \boldsymbol{\alpha}^t \boldsymbol{\Theta} = \sum_{j=1}^d \alpha_j \theta_j$$

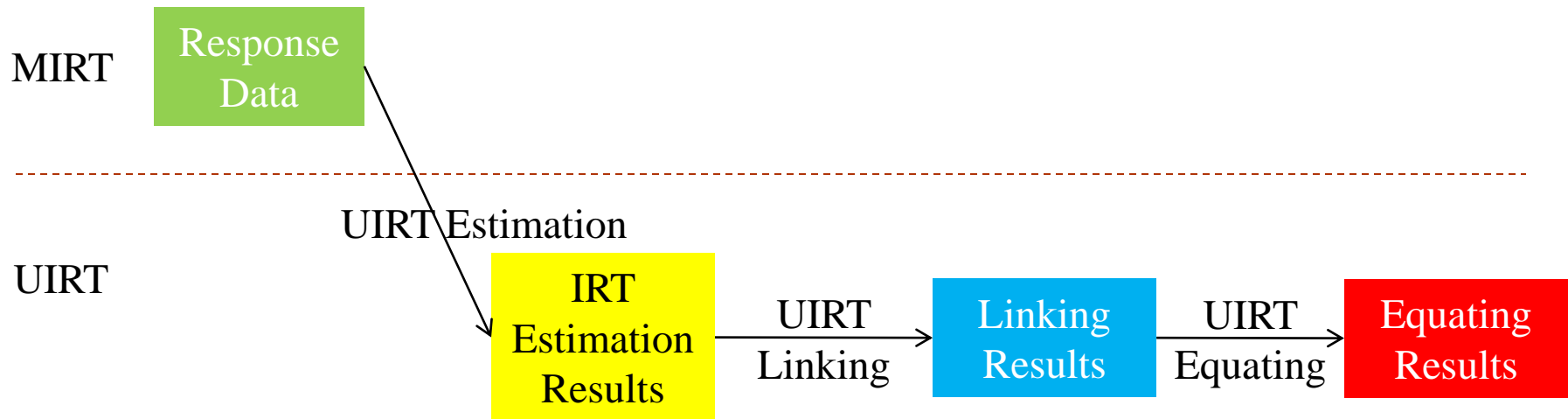
Unidimensionalization in MIRT Equating Procedures

- 4 Possible Procedures of MIRT Equating



Unidimensionalization in MIRT Equating Procedures

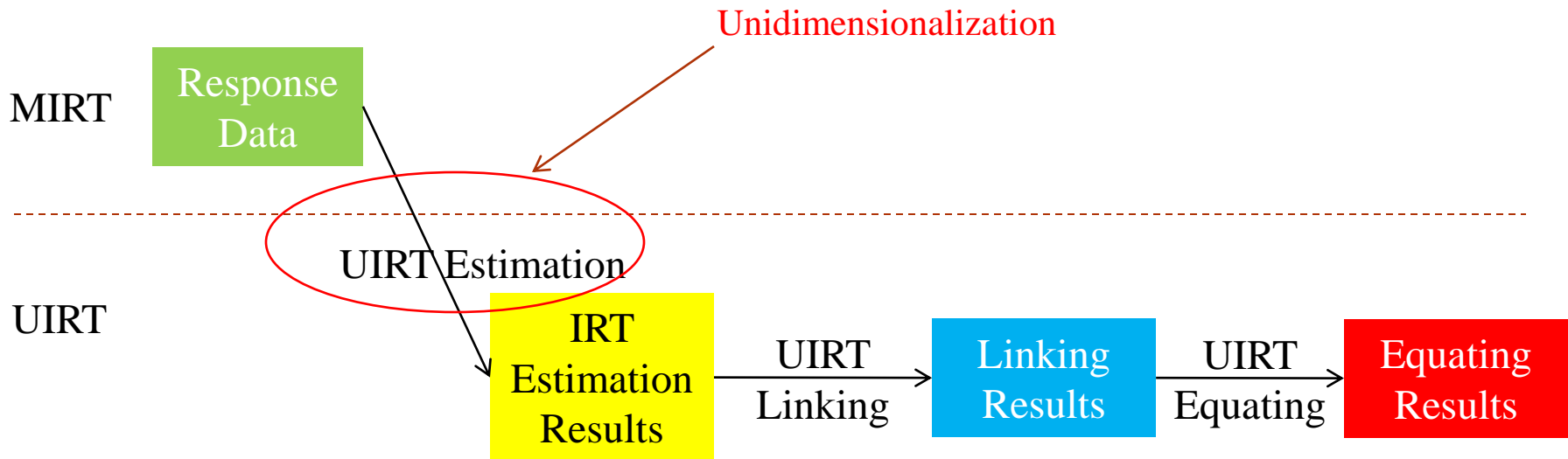
- Possible Procedure 1:
 - UIRT estimation - UIRT linking - UIRT equating



Unidimensionalization at IRT Estimation stage

Unidimensionalization in MIRT Equating Procedures

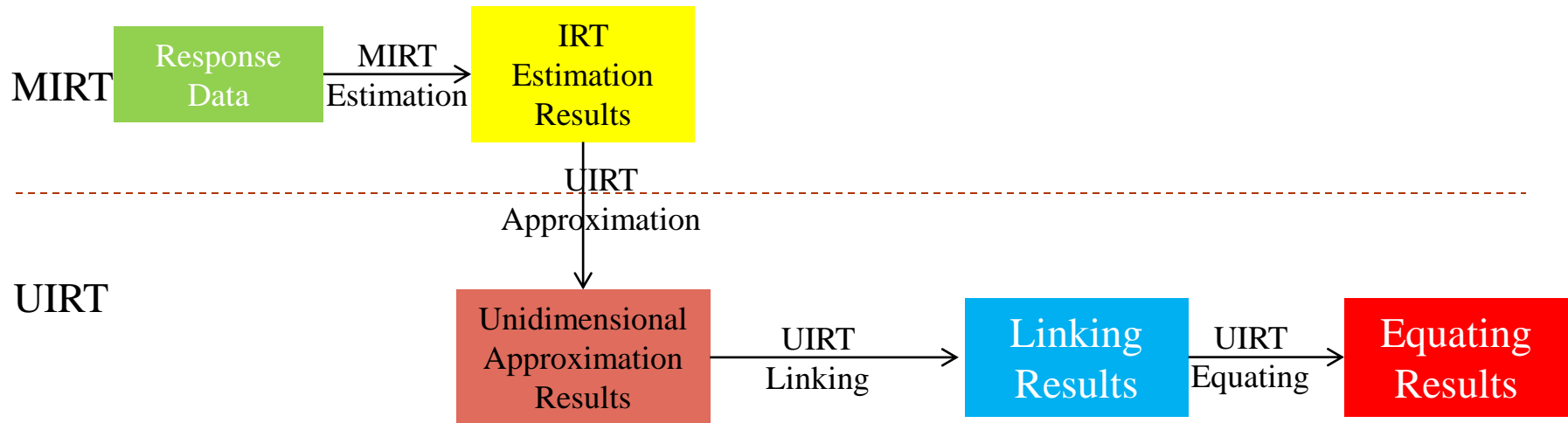
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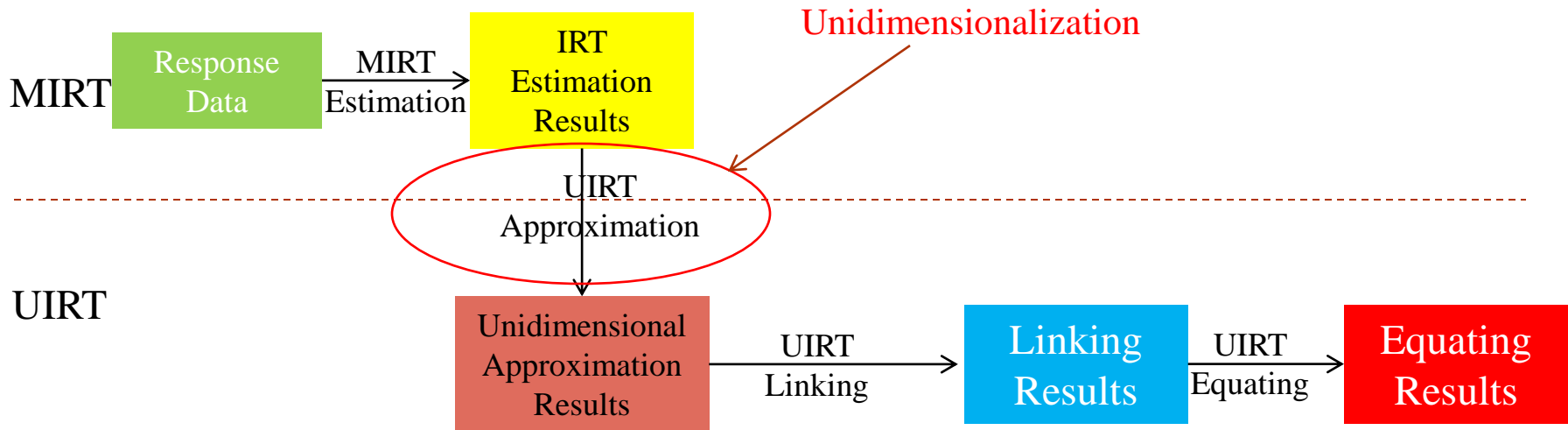
- Possible Procedure 2:
 - MIRT estimation - UIRT approximation - UIRT linking - UIRT equating



Unidimensionalization before IRT linking

Unidimensionalization in MIRT Equating Procedures

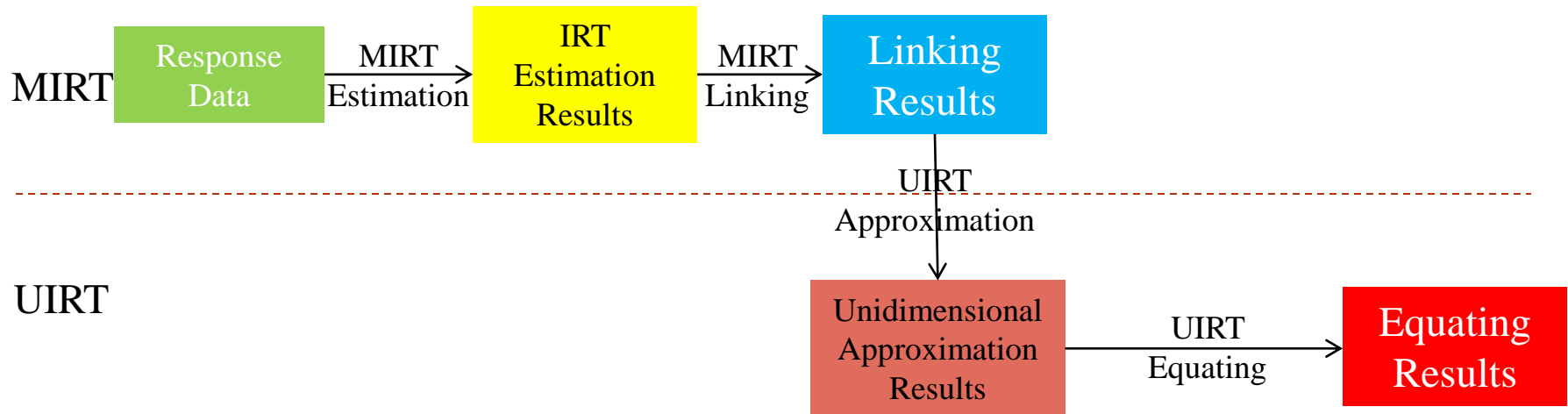
- Possible Procedure 2:
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Unidimensionalization before IRT linking

Unidimensionalization in MIRT Equating Procedures

- Possible Procedure 3:
 - MIRT Estimation - MIRT Linking - UIRT Approximation - UIRT Equating

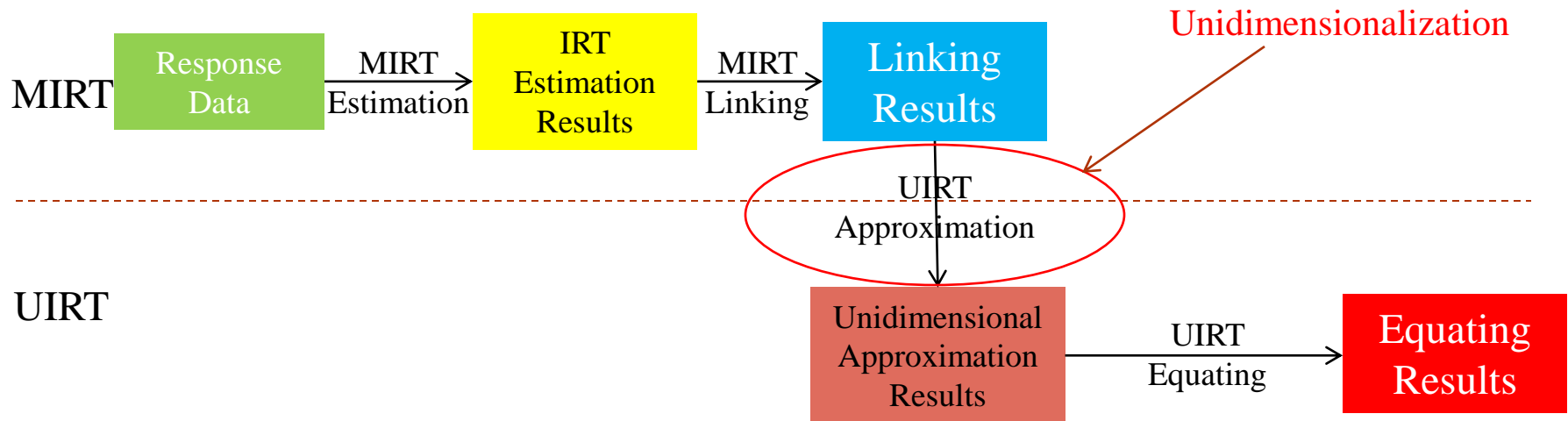


Unidimensionalization before Test Equating stage

Unidimensionalization in MIRT Equating Procedures

- Possible Procedure 3:

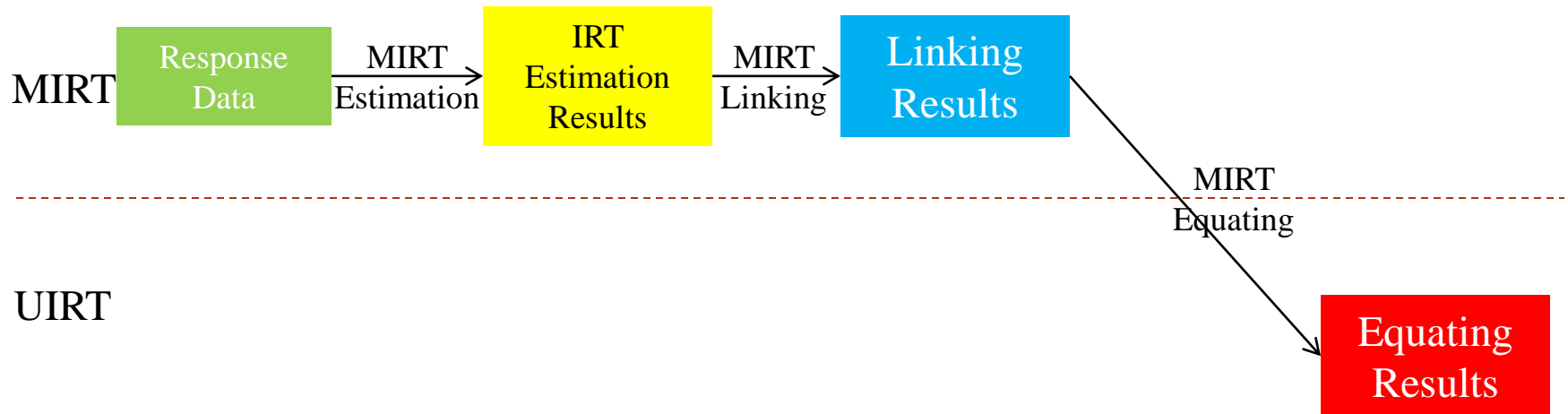
- MIRT Estimation - MIRT Linking - **UIRT Approximation** - UIRT Equating



Unidimensionalization before Test Equating stage

Unidimensionalization in MIRT Equating Procedures

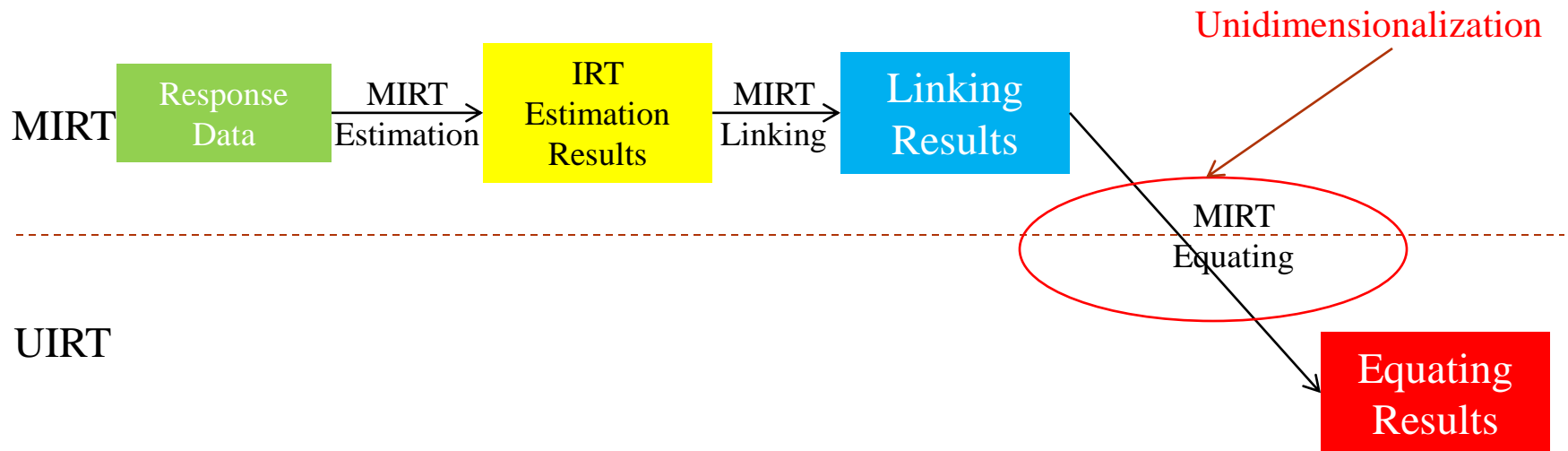
- Possible Procedure 4:
 - MIRT estimation - MIRT linking - MIRT Equating



Unidimensionalization at MIRT Equating stage

Unidimensionalization in MIRT Equating Procedures

- Possible Procedure 4:
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Unidimensionalization at MIRT Equating stage

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- To evaluate the performance of the MIRT equating procedures under NEAT design.
- To explore how different MIRT linking methods interacting with MIRT equating procedures (Brossman, 2010) impact on the equating results, under various testing conditions.
- To provide a possible guidance to educational practitioners for their future MIRT equating application.

MIRT Linking Methods used in the Study

- Min's (M) Method (2003)
- Oshima, Davey and Lee's (ODL) Method (2000)
 - The direct method (OD)
 - The Test Characteristic Function method (TCF)
 - The Item Characteristic Function method (ICF)
- Reckase and Martineau (NOP) Method (2004)
- Coefficients Obtained from These MIRT Linking Methods
 - Rotation Matrix - \mathbf{T}
 - Translation Vector - \mathbf{m}
 - Dilation Vector - \mathbf{K}

MIRT Equating Methods used in the Study

- MIRT Equating Methods (Brossman, 2010)
 - Full MIRT observed score equating method (MOSE)
 - (Possible procedure 4)
 - Unidimensional approximation of MIRT true score equating (ATSE)
 - (Possible procedure 3)
 - Unidimensional approximation of MIRT observed score equating (AOSE)
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So, only Procedure 3 and Procedure 4 were applied in this study.

MIRT Equating Methods for This Study (cont.)

- Full MIRT Observed Score Equating Procedure
 - The full MIRT observed score equating method is a straightforward extension of UIRT observed score equating through the compound binomial recursion formula.

$$f(x) = \sum_1 \sum_2 \dots \sum_m f(x | \boldsymbol{\theta}) \psi(\boldsymbol{\theta})$$

- or

$$f(x) = \int \int \dots \int f(x | \boldsymbol{\theta}) \psi(\boldsymbol{\theta}) d\boldsymbol{\theta}$$

- where m is defined as the number of dimensions.

MIRT Equating Methods for This Study (cont.)

- Unidimensional Approximation of MIRT True Score Equating
- The UIRT true score equating procedure is utilized to equated **composite true scores** (T_α) on both multidimensional test forms. Thus,

$$irt_B(\tau_{\alpha Bi}) = \tau_B(\tau_{\alpha Ei}^{-1})$$

- and

$$func(\theta_{\alpha i}) = \tau_{\alpha A} - \sum_{j:A} p_{ij}(\theta_{\alpha i} | a_{\alpha j}, b_{\alpha j}, c_j)$$

- Finally, the composite true score on the base form $\tau_{\alpha B}(\theta_\alpha)$ associated with the composite true score on the equated form $\tau_{\alpha E}(\theta_\alpha)$ can be computed as

$$\tau_{\alpha B} = \sum_{j:B} p_{ij}(\theta_{\alpha i} | a_{\alpha j}, b_{\alpha j}, c_j)$$

MIRT Equating Methods for This Study (cont.)

- Unidimensional Approximation of MIRT Observed Score Equating
 - The conditional distributions for the **unidimensional ability composite** $f(x|\theta_\alpha)$ is determined at each composite ability level (θ_α) through the compound binomial recursion formula.

$$f(x) = \sum_{\theta_\alpha} f(x|\theta_{\alpha i})\psi(\theta_{\alpha i})$$

- Then,

$$f(x) = \int_{\theta_\alpha} f(x|\theta_{\alpha i})\psi(\theta_{\alpha i})d\theta_\alpha$$

Simulation Design

- MIRT model used: M2PL (With $D=1.7$)
- Test length: total 40 items, 20 anchor items
- Test structure: Approximate simple structure (APSS) and complex structure (CS)
- Sample size: 2000
- Replication time: 200
- Population Design:
 - Null condition
 - Mean-difference
 - SD-difference
 - Correlation-difference
- MIRT estimation software: TESTFACT
- MIRT linking and MIRT equating: R

Evaluation Criteria

- Weighted average equating bias ($Bias_w$)

$$Bias_i = \frac{\sum_{k=1}^N [\hat{e}_{base_k}(x_i) - e_{base}(x_i)]}{N}$$

For the entire test: $Bias_w = \sum_{x=1}^{39} Bias[\hat{e}_{base}(x_i)]P(x_i)$

- Weighted Average Root Mean Square Deviation (ARMSDw)

$$RMSD_i = \sqrt{\frac{1}{N} \sum_{k=1}^N [\hat{e}_{base_k}(x_i) - e_{base}(x_i)]^2}$$

For the entire test: $ARMSD_w = \sum_{x=1}^{39} RMSD[\hat{e}_{base}(x_i)]P(x_i)$

Results

Repeated ANOVA Analysis ($BIAS_w$ and $ARMSD_w$)

| Statistic | Factors | Source | Partial ω^2 | Statistic | Factors | Source | Partial ω^2 |
|-----------|---------------------------|----------------------|--------------------|---------------------------|---------|----------------------|--------------------|
| $ARMSD_w$ | Between | test_str | 0.0067 | $Bias_w$ | Between | test_str | 0.02067 |
| | Between | group | 0.91944 | | Between | group | 0.92557 |
| | Between | test_str*group | 0.02128 | | Between | test_str*group | 0.00458 |
| | Within | link | 0.94089 | | Within | link | 0.8497 |
| | Within | link*test_str | 0.03362 | | Within | link*test_str | 0.00641 |
| | Within | link*group | 0.94122 | | Within | link*group | 0.88045 |
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- **The 2nd largest effect size: equating method * group distribution**

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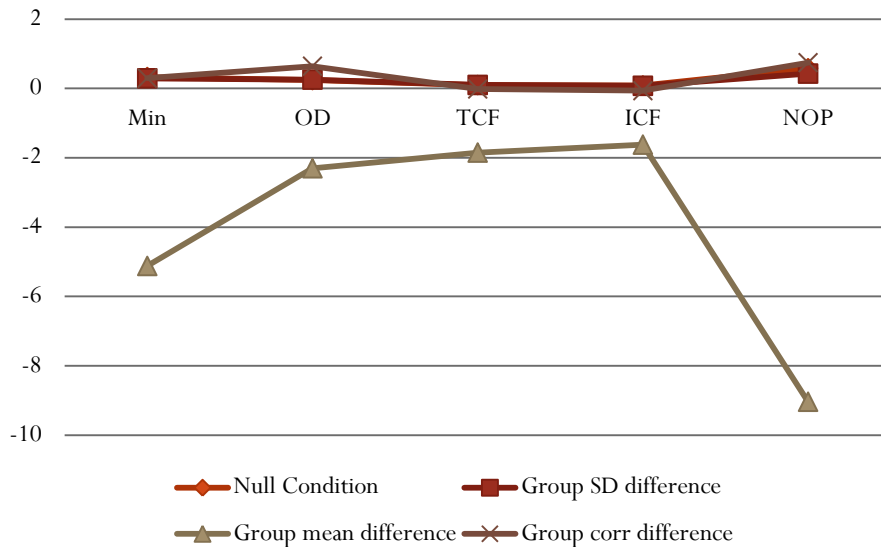
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- The 2nd largest effect size: equating method * group distribution
- **Test structure and all the interactions including test structure-very small effect size**

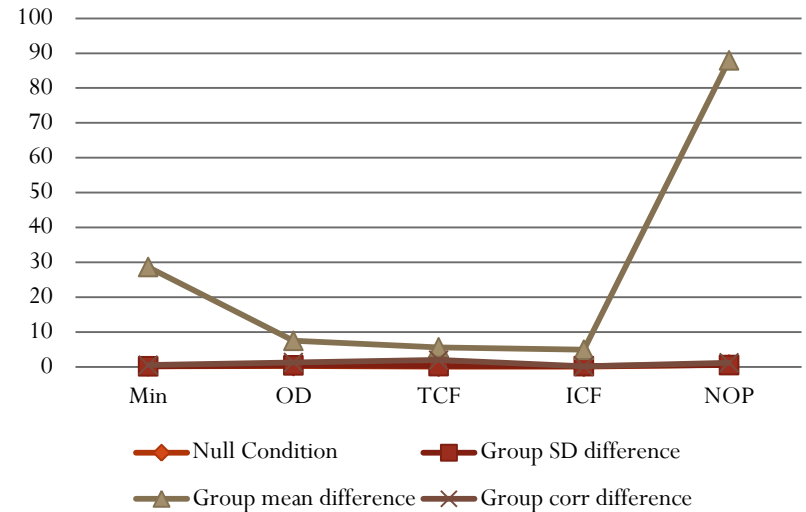
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- Comparison for the Linking Method x Group Distribution Interaction

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$ARMSD_w$



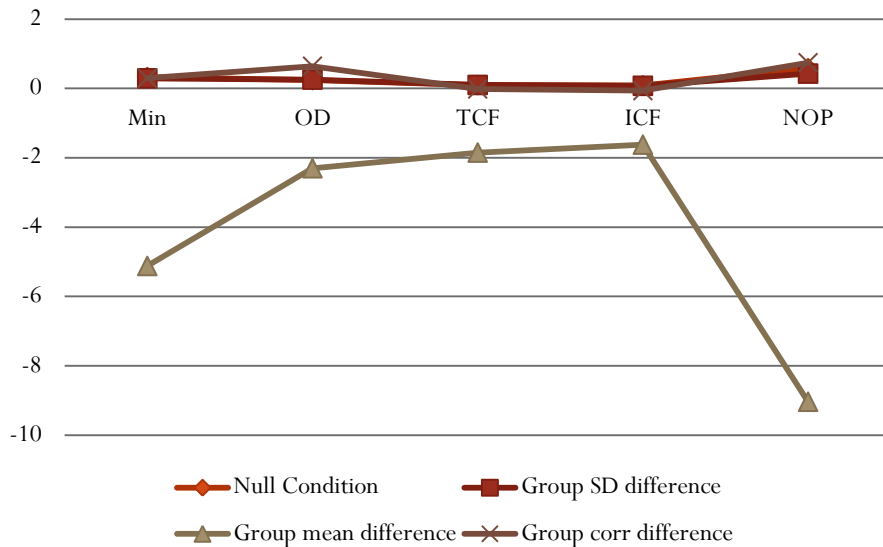
| Bias | Min | OD | TCF | ICF | NOP | ARMSD | Min | OD | TCF | ICF | NOP |
|-----------------------|---------|---------|---------|---------|---------|---------|---------|---------|---------|---------|-----|
| Null Condition | 0.32926 | 0.24153 | 0.09908 | 0.08588 | 0.58308 | 0.23089 | 0.28789 | 0.11543 | 0.11071 | 0.69485 | |
| Group SD difference | 0.29248 | 0.25659 | 0.10186 | 0.07712 | 0.42949 | 0.20884 | 0.48575 | 0.26935 | 0.26063 | 0.55671 | |
| Group mean difference | -5.1191 | -2.3007 | -1.8481 | -1.6189 | -9.0351 | 28.6865 | 7.52435 | 5.64556 | 4.9614 | 87.9546 | |
| Group corr difference | 0.29875 | 0.64432 | -0.015 | -0.0637 | 0.74563 | 0.54088 | 1.33343 | 2.00157 | 0.19166 | 1.24462 | |

- Overall: TCF and ICF performed best across all group distribution conditions;

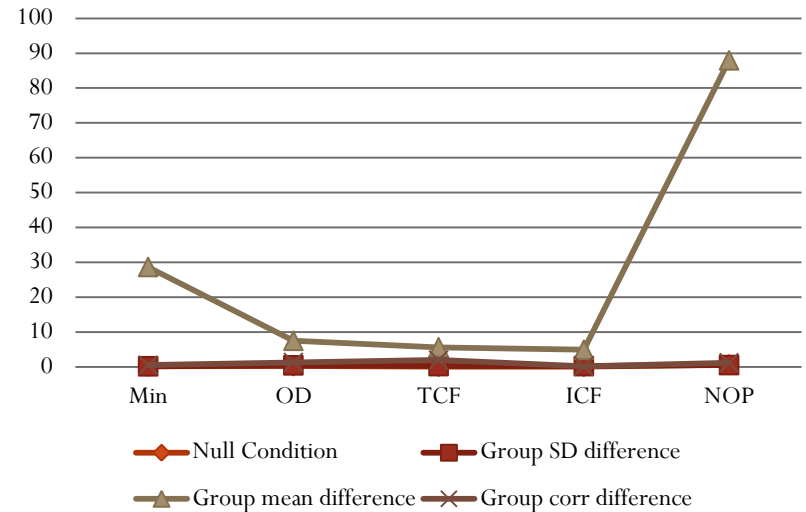
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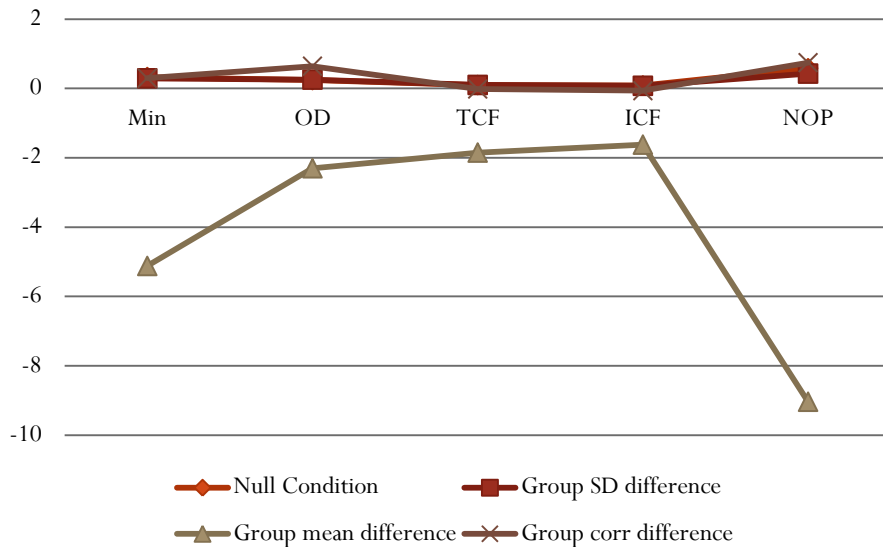
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- Overall: TCF and ICF performed best across all group distribution conditions; **OD and M methods' performances are next;**

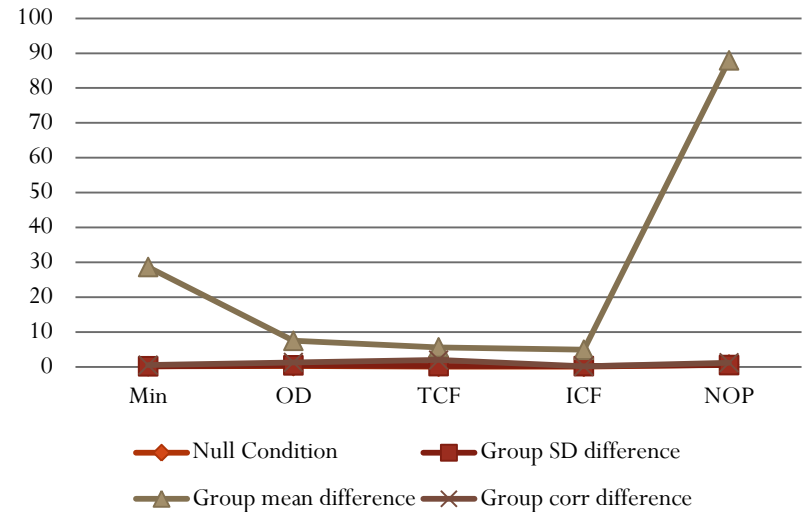
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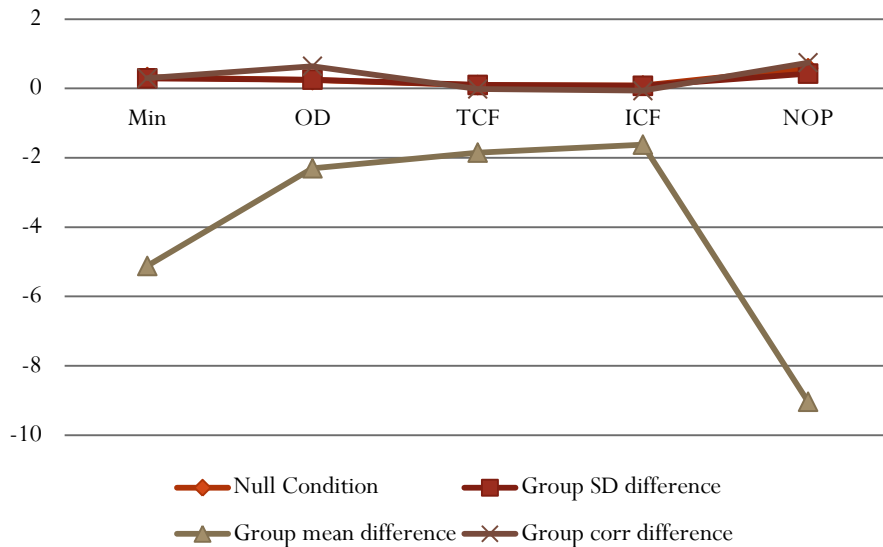
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|-----------------------|---------|---------|---------|---------|---------|---------|---------|---------|---------|---------|-----|
| Null Condition | 0.32926 | 0.24153 | 0.09908 | 0.08588 | 0.58308 | 0.23089 | 0.28789 | 0.11543 | 0.11071 | 0.69485 | |
| Group SD difference | 0.29248 | 0.25659 | 0.10186 | 0.07712 | 0.42949 | 0.20884 | 0.48575 | 0.26935 | 0.26063 | 0.55671 | |
| Group mean difference | -5.1191 | -2.3007 | -1.8481 | -1.6189 | -9.0351 | 28.6865 | 7.52435 | 5.64556 | 4.9614 | 87.9546 | |
| Group corr difference | 0.29875 | 0.64432 | -0.015 | -0.0637 | 0.74563 | 0.54088 | 1.33343 | 2.00157 | 0.19166 | 1.24462 | |

- Overall: TCF and ICF performed best across all group distribution conditions; OD and M methods' performances are next; **NOP method performed worst among all 5 Linking methods.**

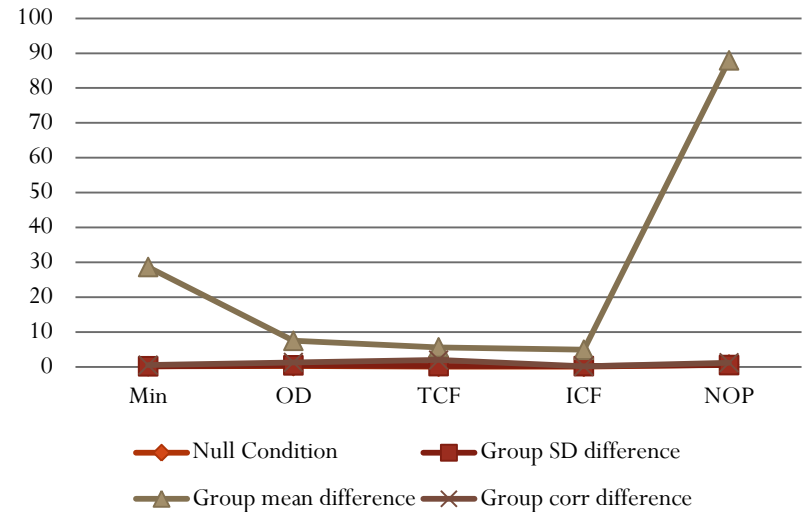
Results (cont.)

- Comparison for the Linking Method x Group Distribution Interaction

$Bias_w$



$ARMSD_w$



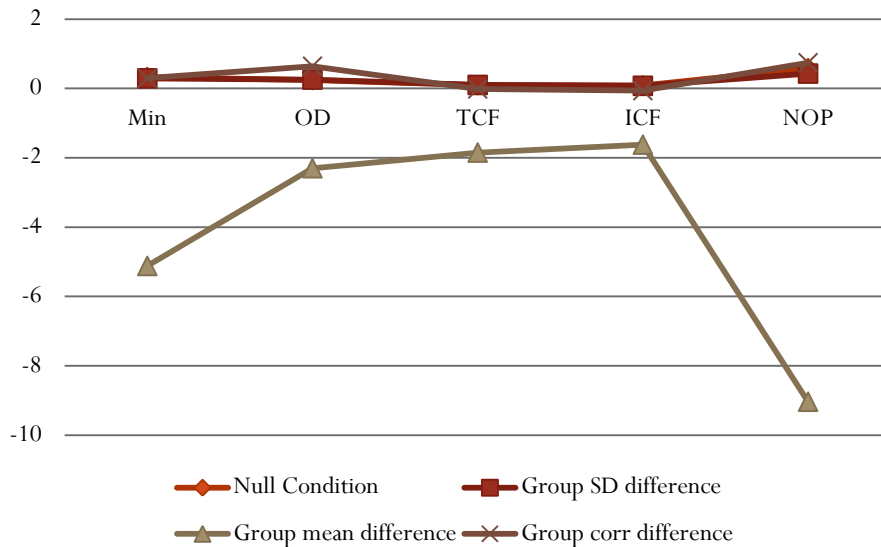
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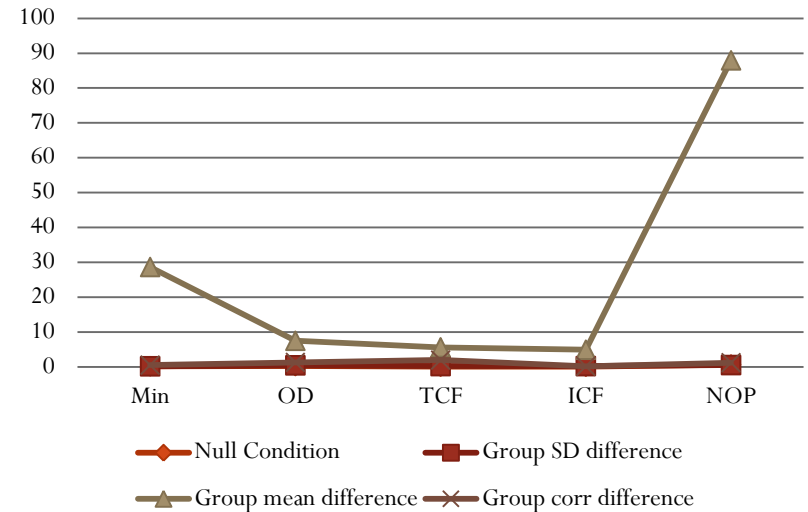
Results (cont.)

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$ARMSD_w$



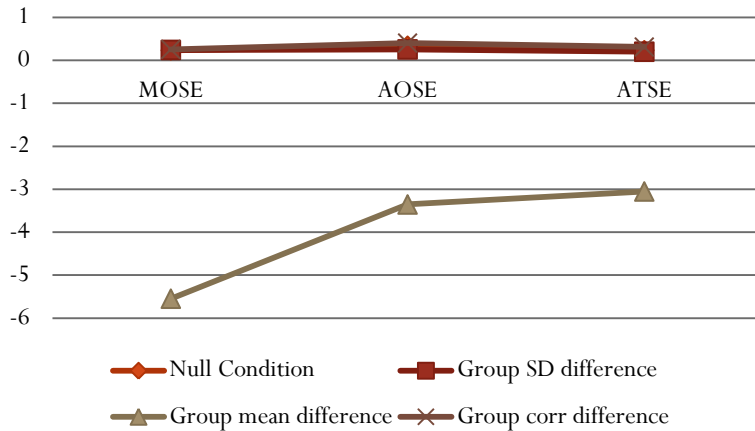
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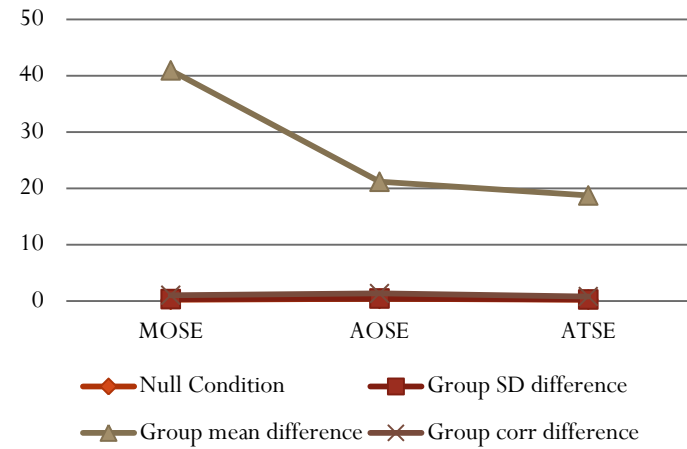
Results (cont.)

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$ARMSD_w$



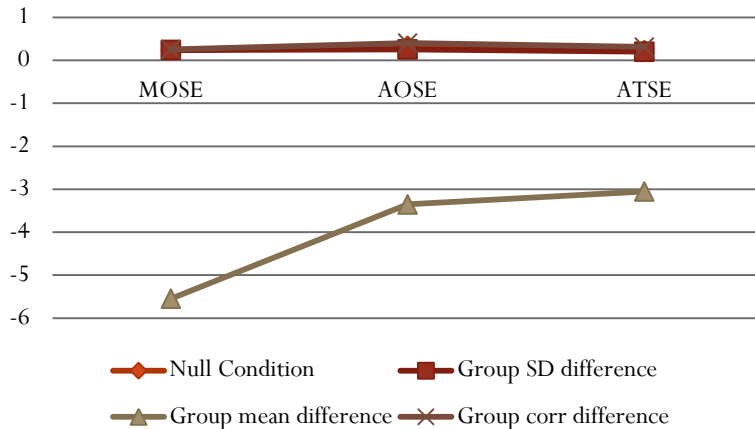
| $BIAS_w$ | MOSE | AOSE | ATSE | $ARMSD_w$ | MOSE | AOSE | ATSE |
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| Null Condition | 0.23211 | 0.33764 | 0.23354 | 0.23195 | 0.39008 | 0.24183 | |
| Group SD difference | 0.24126 | 0.25349 | 0.19976 | 0.34226 | 0.45045 | 0.27605 | |
| Group mean difference | -5.5469 | -3.3541 | -3.0522 | 40.9469 | 21.1754 | 18.7411 | |
| Group corr difference | 0.25488 | 0.40193 | 0.30917 | 1.0374 | 1.34202 | 0.80787 | |

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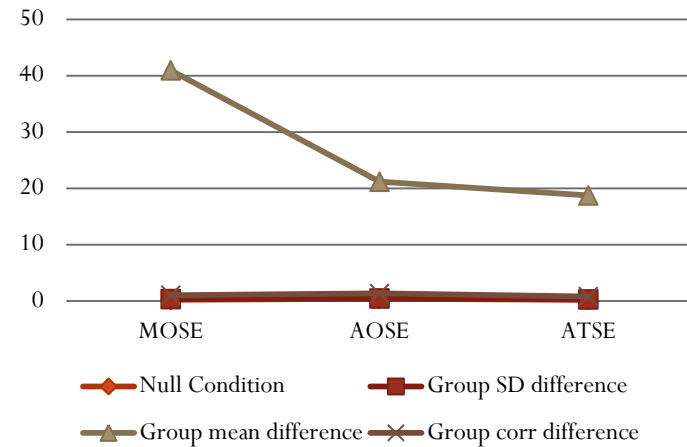
Results (cont.)

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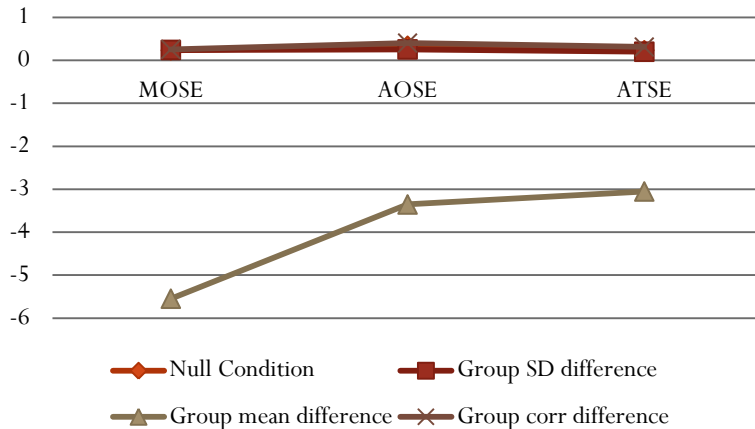
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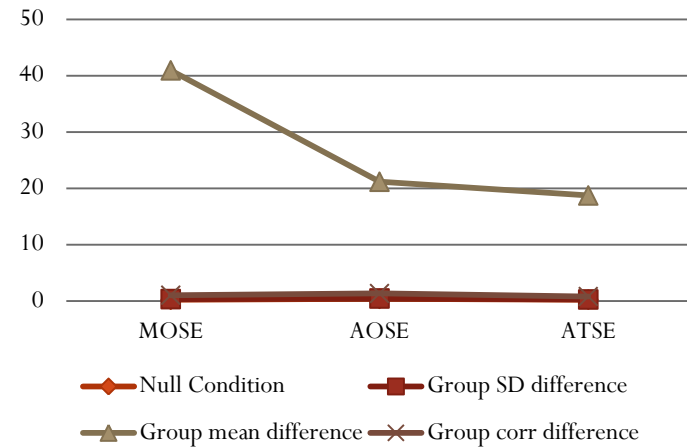
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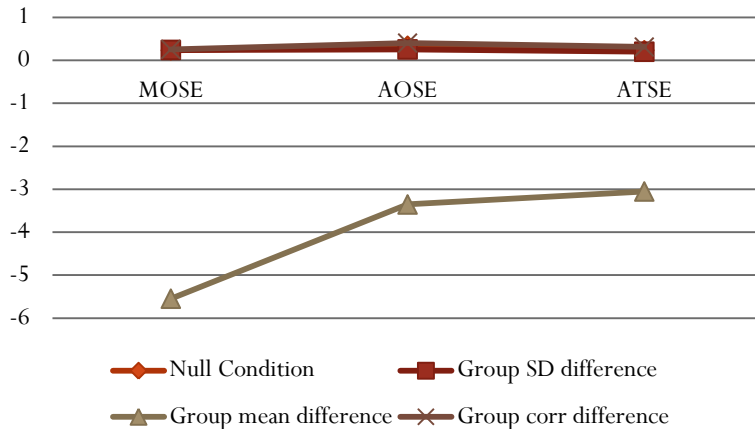
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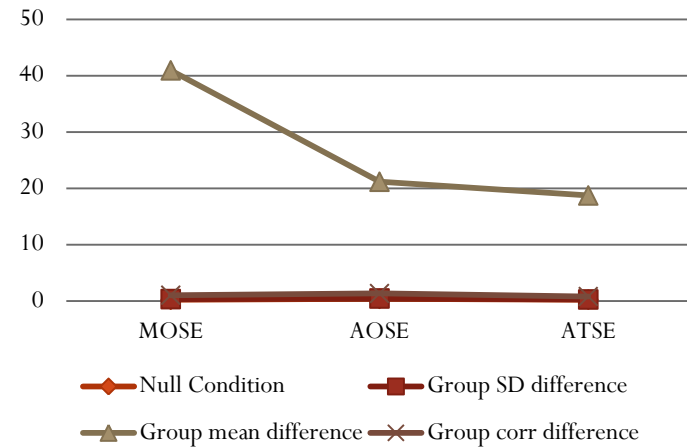
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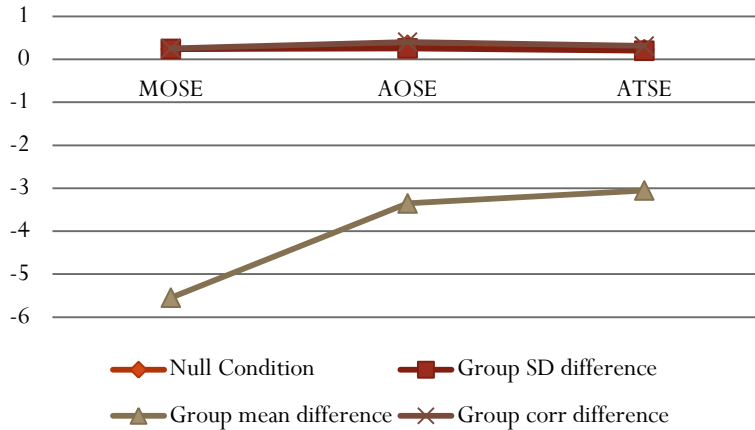
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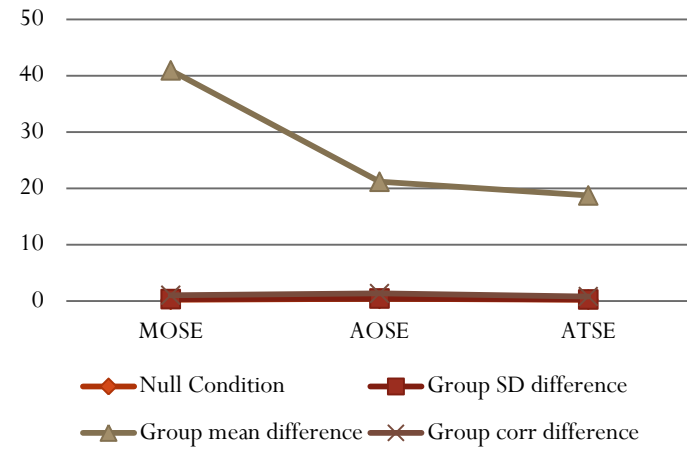
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 - ATSE procedure demonstrated, overall, the best equating performance as compared with the other two equating procedures (i.e., MOSE and AOSE) across all group distribution conditions.

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- Orthogonal rotation vs. oblique rotation in MIRT linking influencing MIRT equating results needs further investigation

Key References

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Thank you!

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