

Comparison of exclusion, imputation and modelling of missing binary outcome data in frequentist network meta-analysis

Chrysostomos Kalyvas¹, Loukia M Spinelis²

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Hannover Medical School

¹Department of Health Technology Assessment Statistics, Merck Sharp & Dohme, Brussels, Belgium

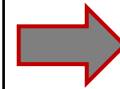
²Midwifery Research and Education Unit, Hannover Medical School, Hannover, Germany

What is already known

- **Missing outcome data (MOD)** are inevitable in systematic reviews.
- Reviewers tend to handle MOD as **missing at random (MAR)**.
- Available MAR methods:
 - exclusion and ignorance of MOD;
 - exclusion but accountability of MOD;
 - imputation of observed risks;
 - pattern-mixture model.

Exclusion and ignorance of MOD

	Completers	MOD
Active	😊😊😊😊😊😊😞	😞😊😊
	😊😊😊😊😊😊😞😞	😞😞😞
Placebo	😞😞😞😊😊	😞😞😞😞😞
	😞😞😞😞😊	😞😞😞😊😊

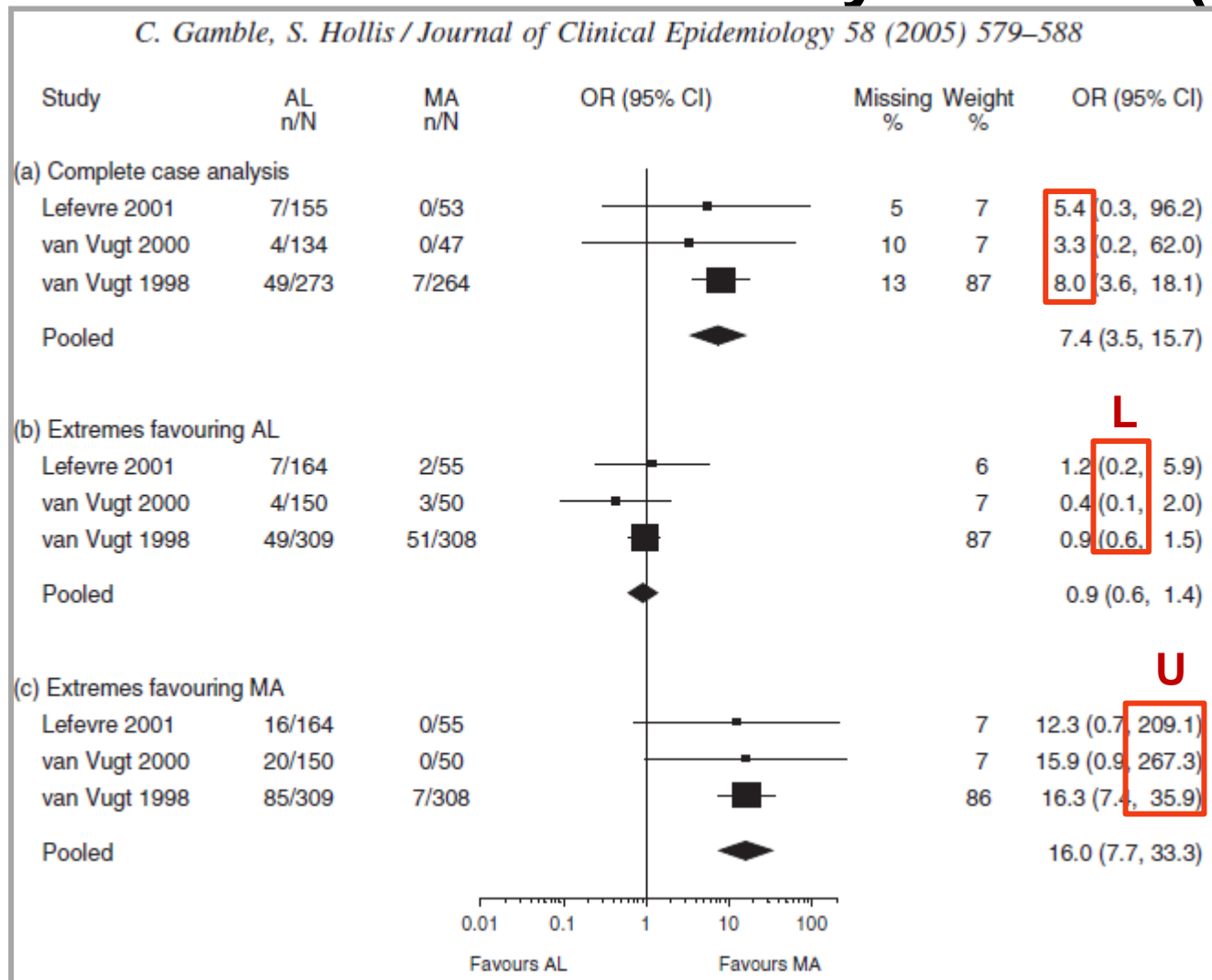


	Completers
Active	😊😊😊😊😊😊😞
	😊😊😊😊😊😊😞😞
Placebo	😞😞😞😊😊
	😞😞😞😞😊

Truth	Risk	Risk Ratio
Active	13/20	2.60 (1.14, 5.93)
Placebo	5/20	

Exclusion	Risk	Risk Ratio
Active	11/14	2.62 (0.98, 7.01)
Placebo	3/10	

Exclusion but accountability of MOD (1)

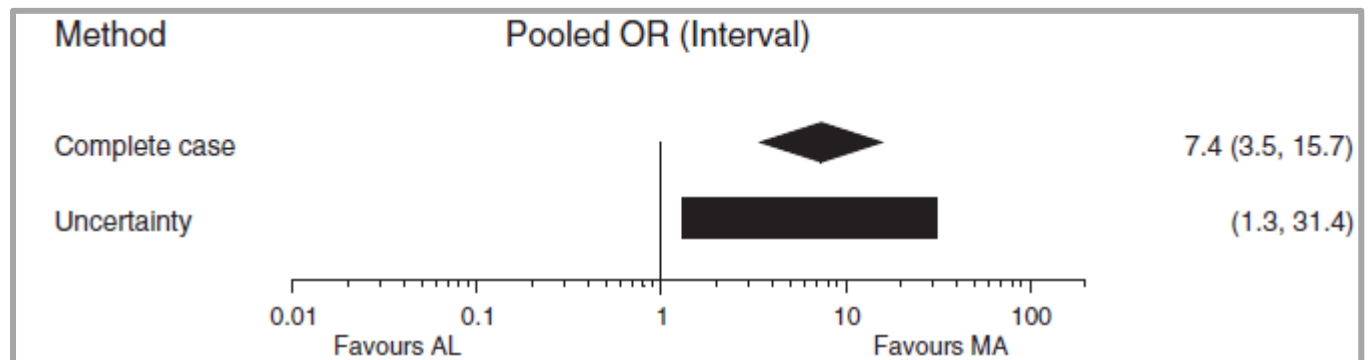
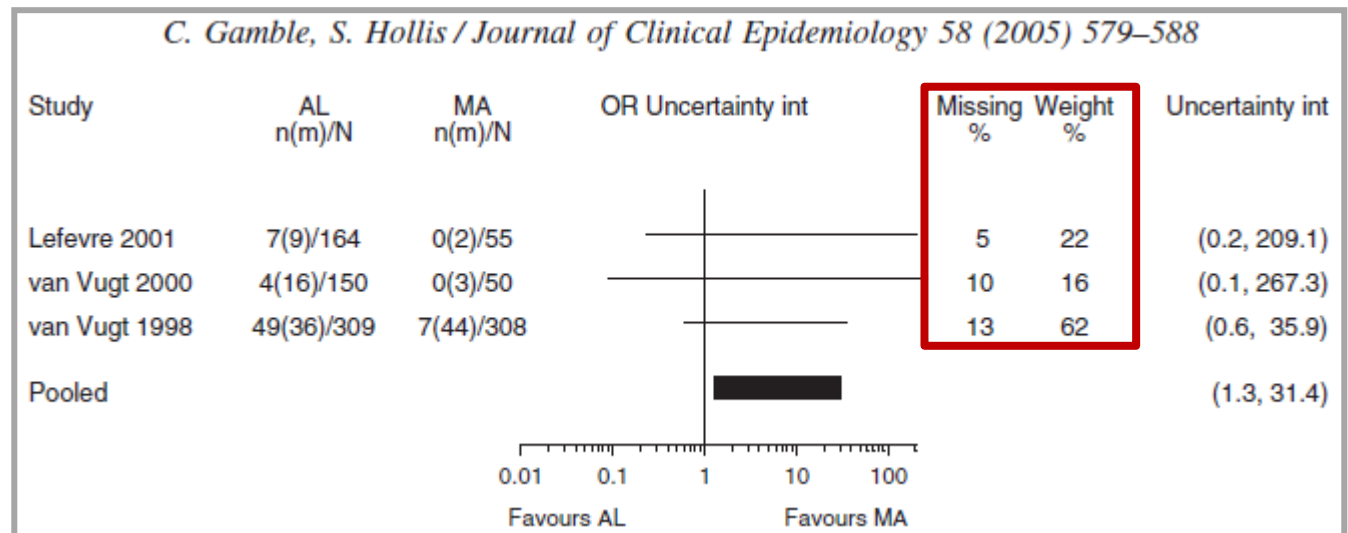


Gamble C, Hollis S. Uncertainty method improved on best-worst case analysis in a binary meta-analysis. *J Clin Epidemiol.* 2005;58(6):579-88.

Exclusion but accountability of MOD (2)

within-trial:

$$se(logOR) = \frac{U - L}{2 \cdot 1.96}$$



Gamble C, Hollis S. Uncertainty method improved on best-worst case analysis in a binary meta-analysis. *J Clin Epidemiol.* 2005;58(6):579-88.

Imputation of observed risks

Truth	Risk	%MOD	Risk Ratio
Active	13/20	30%	2.60
Placebo	5/20	50%	(1.14, 5.93)

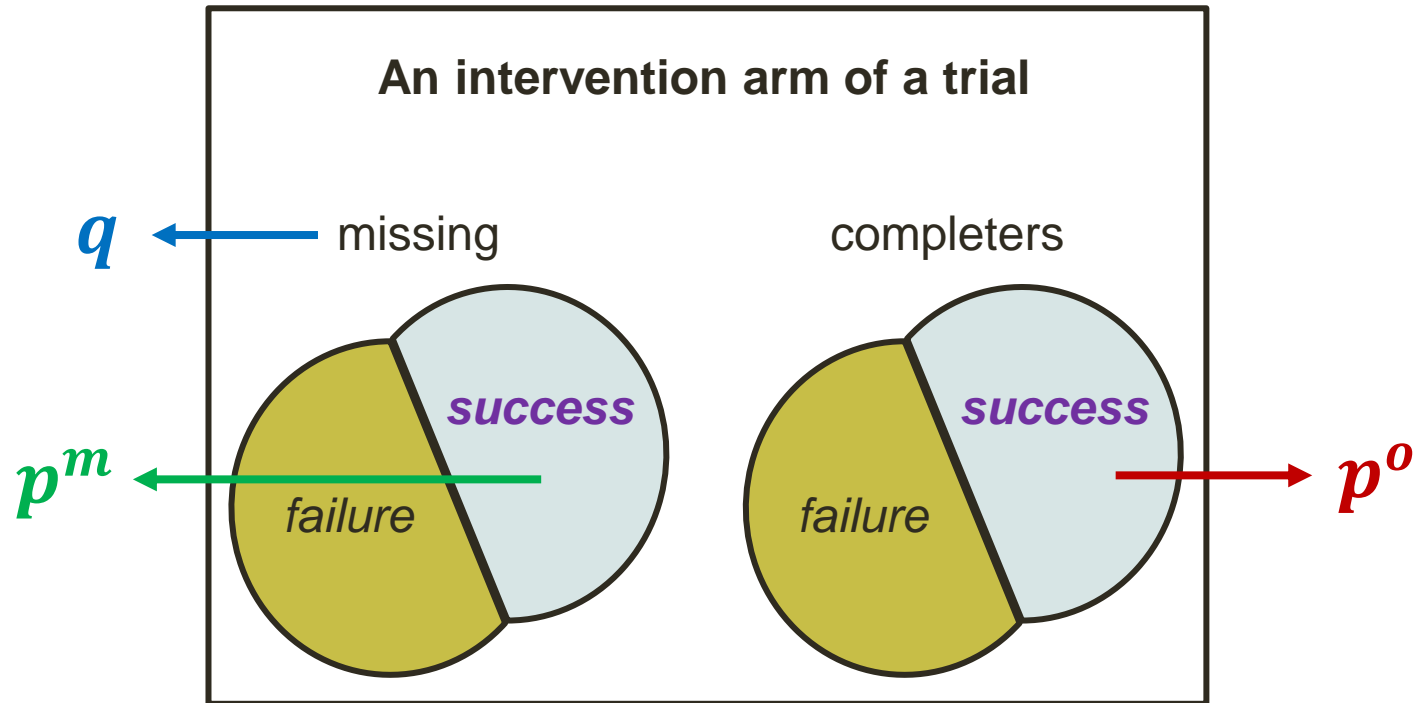
Observed	to impute
$\frac{11}{14} \cdot (1 - 0.3)$	$+$ $\frac{11}{14} \cdot 0.3$
$\frac{3}{10} \cdot (1 - 0.5)$	$+$ $\frac{3}{10} \cdot 0.5$



Imputation	Risk	Risk Ratio
Active	11/14	2.62
Placebo	3/10	(1.29, 5.31)

- **Same RR** with Exclusion (and ignorance of MOD)...
- but **more precise** for maintaining the randomised sample

Pattern-mixture model



$$p = p^o \cdot (1 - q) + p^m \cdot q$$

Informative Missingness Odds Ratio

Under pattern-mixture model

$$IMOR = \frac{p^m / (1 - p^m)}{p^o / (1 - p^o)} \quad \text{'Odds of an event being missing to odds of an event being observed'}$$

$$\log(IMOR) = \delta_{ik} \sim N(\Delta_{ik}, \sigma_{ik}^2) \quad \text{arm } k \text{ of trial } i$$

Δ	interpretation
$\Delta > 0$	more likely that a missing case to be an event
$\Delta < 0$	less likely that a missing case to be an event
$\Delta = 0$	Missing at random (on average)

Pattern-mixture model with IMOR

- Risk ratio (RR) same with Exclusion.

- Variance of RR \equiv Variance due to sampling error

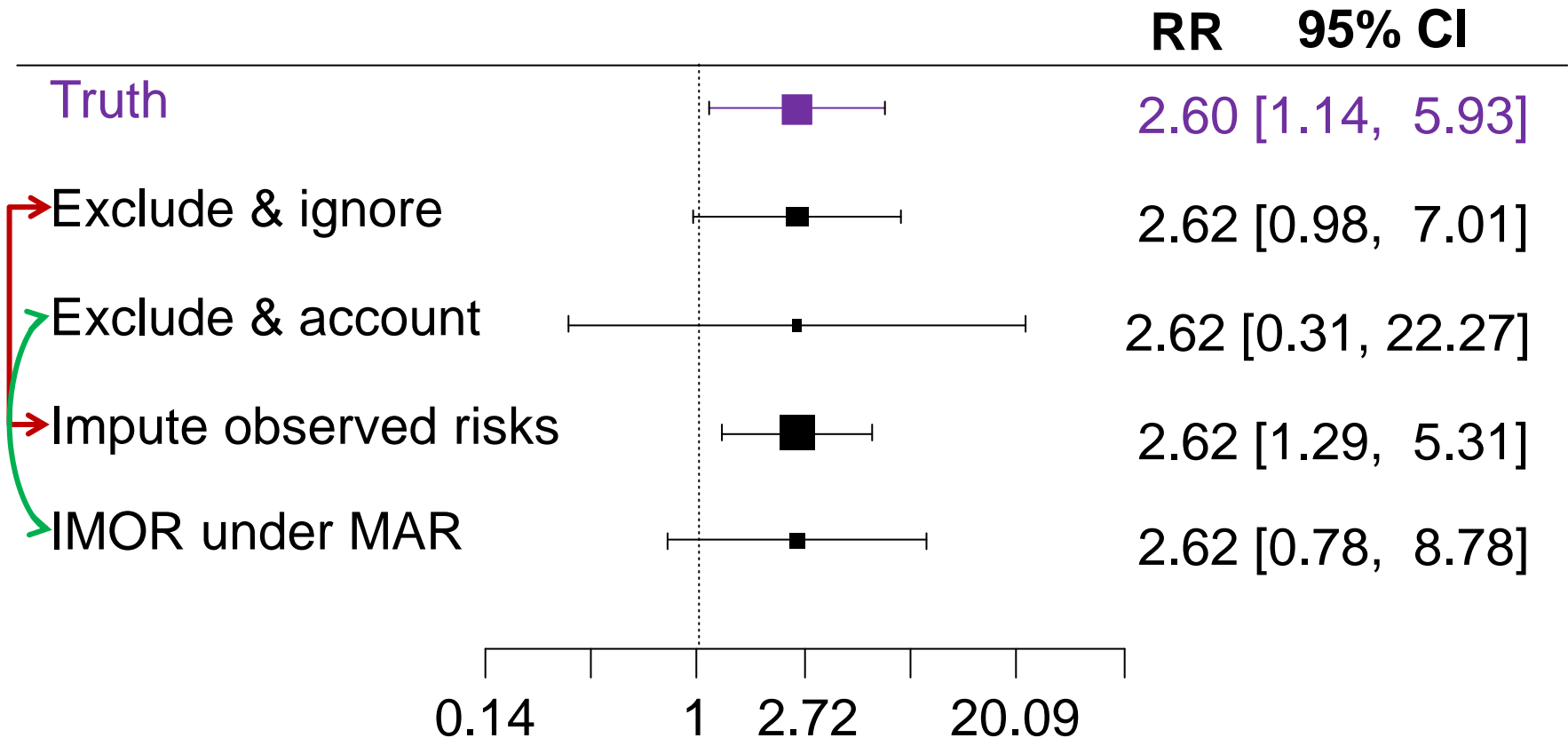


Variance arising from IMOR | **MAR**



IMOR	Risk	Risk Ratio
Active	11/14	2.62
Placebo	3/10	(0.78, 8.76)

Bring all methods together



— ignores uncertainty due to MOD

— accounts uncertainty due to MOD

Aims of the study

Empirical study

- Agreement of the methods regarding **network meta-analysis**

(NMA) parameters:

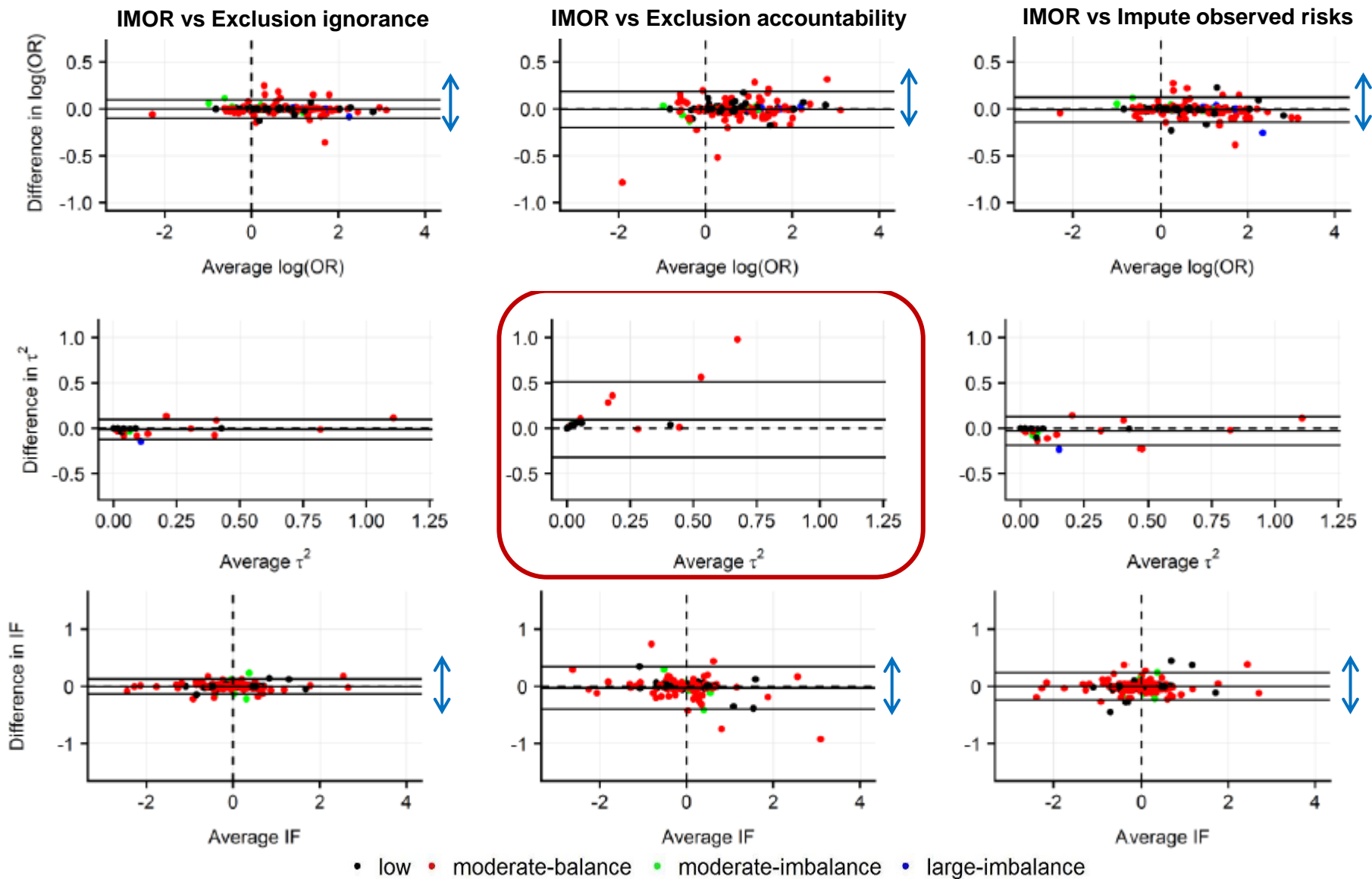
- Exclude and ignore MOD
 - Exclude but account MOD
 - Impute observed risks
- } vs. IMOR under MAR

Simulation study

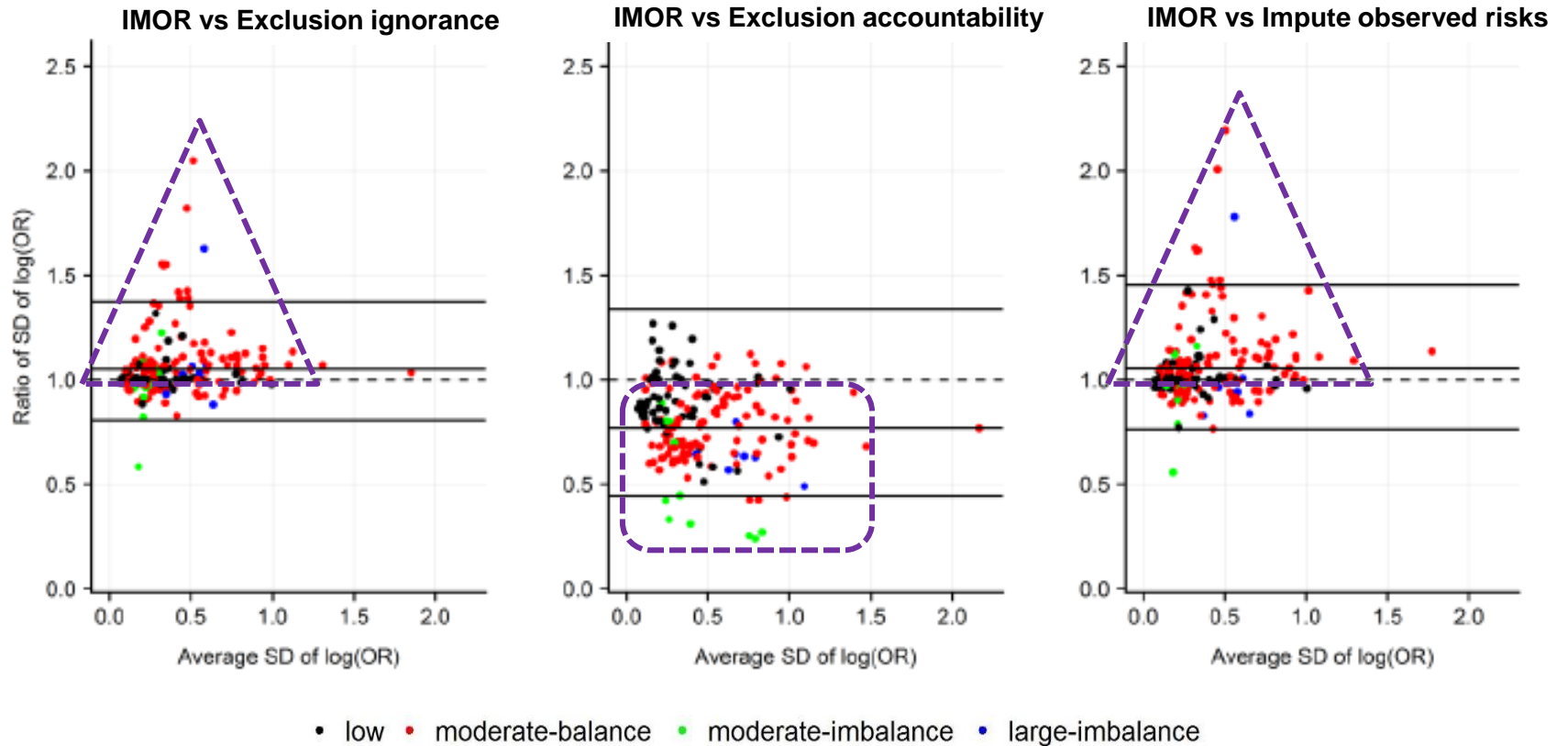
- Performance of methods in terms of:
 - Bias (NMA log odds ratio, τ^2)
 - Coverage (NMA log odds ratio)

Empirical study: Methods used

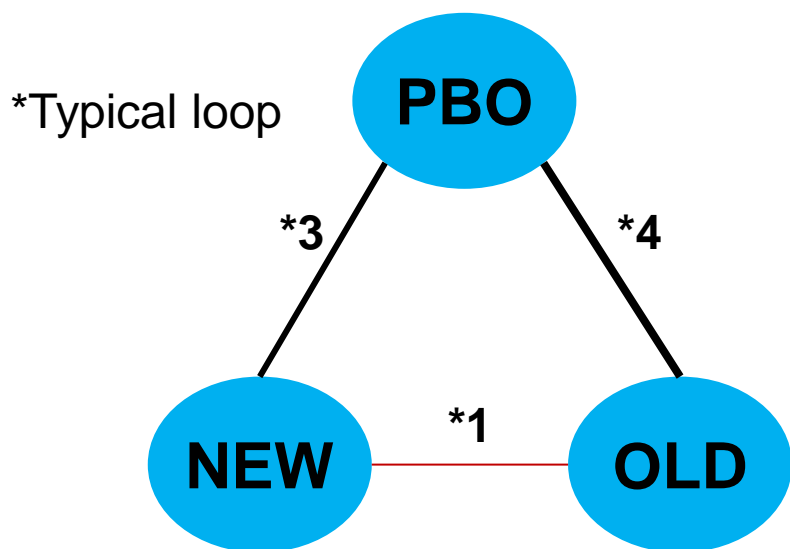
- Database: 29 networks from various health fields.
- MOD distinction: **low**, **moderate** and **large** w/o balanced arms.
- Bland-Altman scatterplots to assess agreement.
- Random-effects NMA model based on electrical network theory.
- All analyses were ran using the *netmeta* R package.



Posterior standard deviation of NMA log OR



Simulation study: Methods used



Scenarios based on 29 networks

- PBO-ctrl studies have less patients
- Baseline risk higher in OLD than PBO

- IMOR = 2 for NEW and OLD
- IMOR = 0.5 for PBO

Comparison		Moderate %MOD		Large %MOD	
NEW	PBO	<i>U</i> (0.05,0.10)	<i>U</i> (0.11,0.20)	<i>U</i> (0.21,0.30)	<i>U</i> (0.31,0.40)
OLD	PBO				
NEW	OLD				

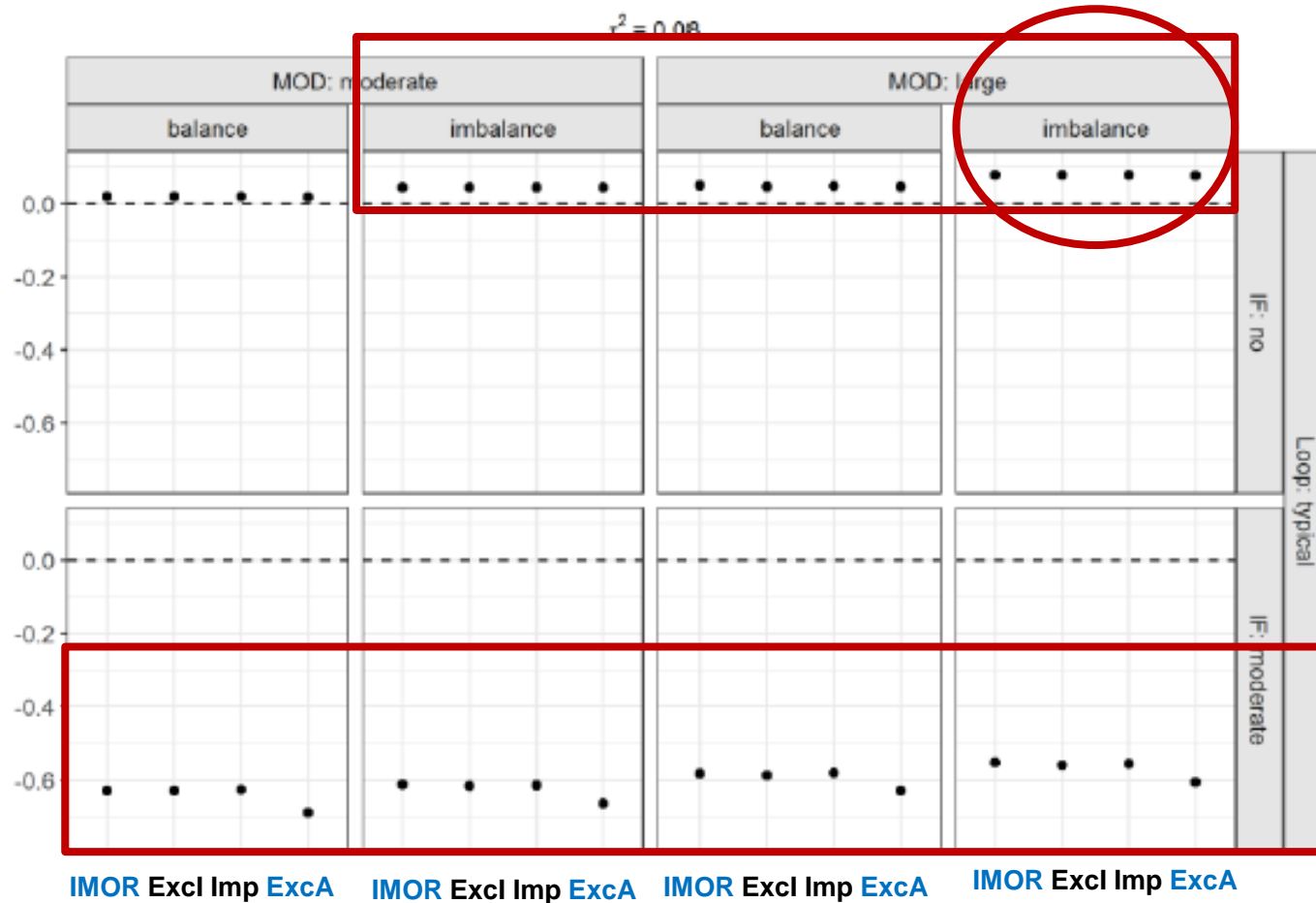
Veroniki AA, Mavridis D, Higgins JPT, Salanti G. Characteristics of a loop of evidence that affect detection and estimation of inconsistency: a simulation study. BMC Med Res Methodol. 2014;14:106.

Hartung J, Knapp G. A refined method for the meta-analysis of controlled clinical trials with binary outcome. Stat Med. 2001;20:3875–89.

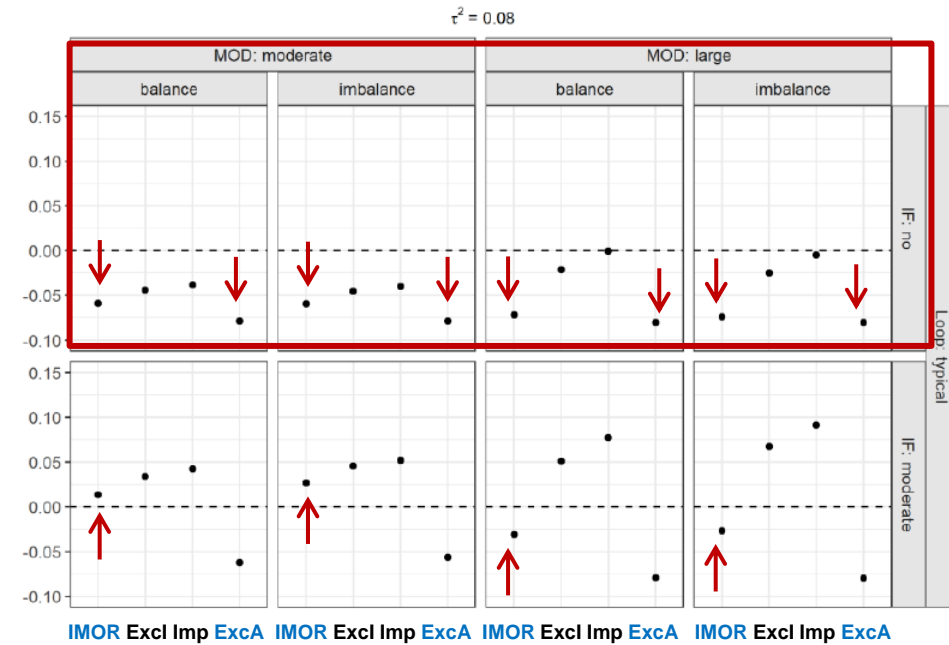
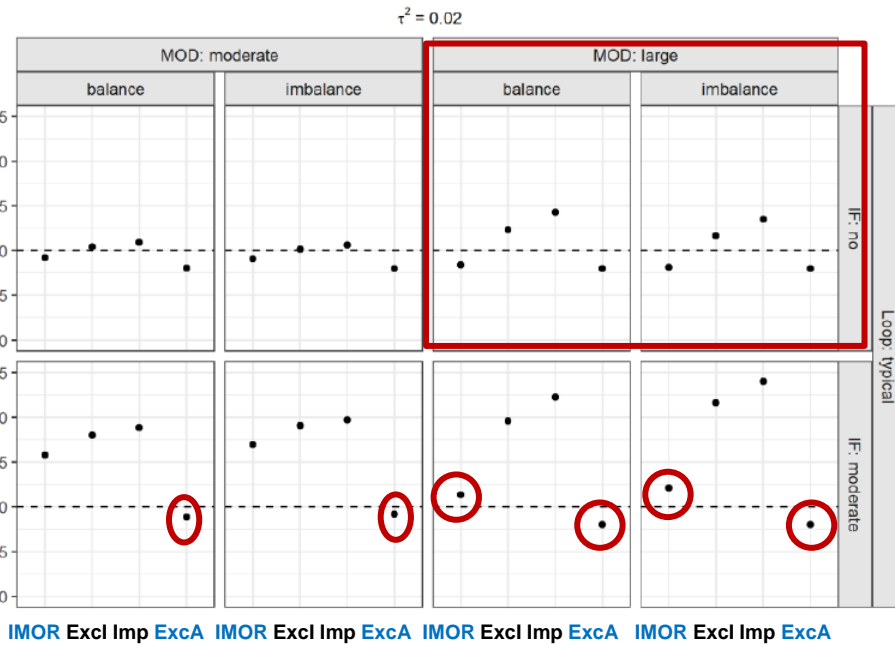


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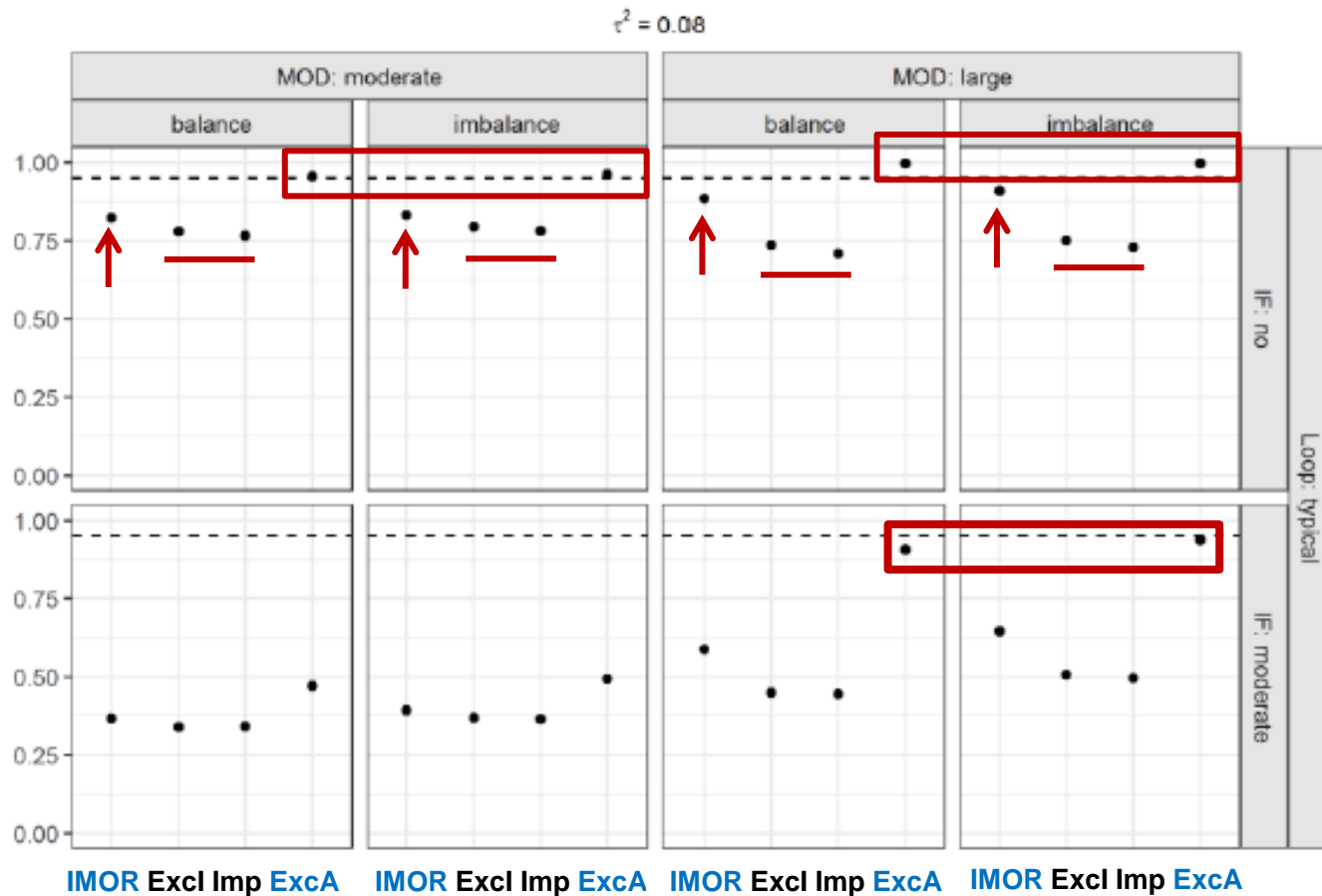
Mean bias for log OR (new vs. old): substantial τ^2



Mean bias for τ^2 : typical loop



95% coverage for log OR (new vs. old): substantial τ^2



Take home message

- ✗ Methods that **ignore uncertainty** due to MOD are **simple, but suboptimal** overall.
- ✓ Methods that **account for uncertainty** due to MOD should **be preferred**.
- ? 'Exclusion with accountability' **shares same shortcoming** with 'Exclusion but ignorance'.
- IMOR is computationally intensive but **adjusts results for MOD** while **maintaining randomised sample**.
- ? Interpret IMOR with **caution** when: **large MOD, substantial τ^2** and/or **inconsistency**.

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<https://theculturetrip.com/europe/belgium/articles/10-things-to-know-when-traveling-in-brussels-belgium/>

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Thank you for your attention!