The <u>NE</u>twork meta-analysis with <u>Missing Outcome data (NEMO)</u> project: a combination of a systematic overview, empirical and simulation study

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Why the NEMO project was proposed? (1)

- Missing outcome data (MOD) are ubiquitous in clinical trials...
- ... and in systematic reviews, inevitably!
- Implications of MOD on results:
 - loss of precision, if MOD are excluded;
 - o risk of bias, if MOD are handled inappropriately

(i.e. exclusion or imputation before analysis).

	Observed	<u>MOD</u>
		$\bigcirc\bigcirc\bigcirc\bigcirc\bigcirc$
ACTIVE		Θ
DI ACEDO		88880
PLACEBU		88800

%MOD: 40% vs 50% (active vs placebo)

Truth	14/20	7/20	4.3 (1.2, 16)
Exclude	8/ <u>12</u>	4/ <u>10</u>	3.0 (0.5, 17)
Impute 🙁	8/20	4/20	2.7 (0.6, 11)
Impute ©	16/20	14/20	1.7 (0.4, 7.3)

Why the NEMO project was proposed? (2)

- Several strategies to address MOD in systematic reviews* but without evidence on their performance:
 - exclude before analysis with & without uncertainty due to MOD;
 - impute before analysis & ignore uncertainty due to MOD;
 - model using pattern-mixture or selection model.

^{*}Higgins JPT, Deeks JJ, Altman DG on behalf of the Cochrane Statistical Methods. Chapter 16: Special topics in statistics. In: Higgins JPT, Green S, eds. Cochrane Handbook for Systematic Reviews of Interventions Version 5.1.0 (updated March 2011). The Cochrane Collaboration; 2011

Why the NEMO project was proposed? (3)

- > Intention-to-treat (ITT) principle preferred in the presence of MOD.
- Appropriate application of ITT in systematic reviews requires:
 - Distinction between observed outcomes and MOD;
 (observation-carried-forward methods hinder such an accurate distinction)
 - Information on the reasons for MOD;
 (transparent and detailed reporting of trials is required)
 - Sensitivity analyses that include all randomised individuals under plausible scenarios about MOD.
 - (to evaluate the robustness of the results of primary analysis)

Why the NEMO project was proposed? (4)

Reviewers fail to provide an accurate extraction of outcome data from the included trials.

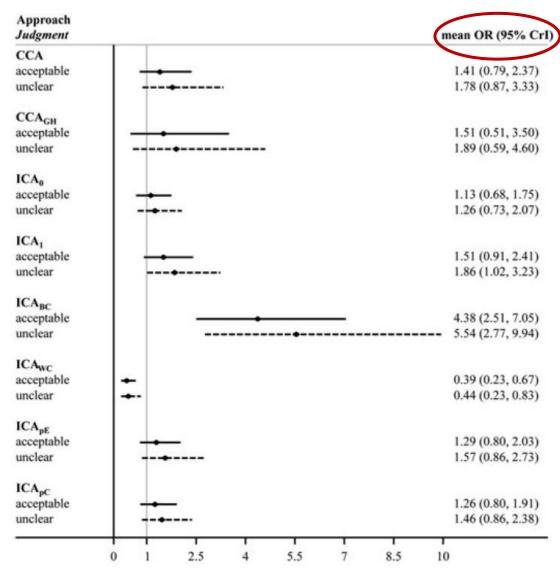
Extraction evaluation				
Item	Aspects reported in the systematic review	Acceptable	Unclear	Unacceptable ^a
i.	Information on outcome of completers for all included studies	✓		
ii.	Information on outcome "withdrawal" for all included studies	✓	✓	✓
iii.	Explicit description of how missing outcome data in meta-analyses is handled	✓	✓	
iv.	Information on outcome of completers and/or outcome "withdrawal" but only for a minority of included studies			✓
^a Meta-analyses that included at least one study with calculated negative non-events were immediately considered to be "unacceptably" extracted.				

	Extraction Is Judged			
Cochrane Mental Health Group	Acceptable	Unclear	Unacceptable	
Depression, anxiety and neurosis	10	17 ^a	25	
Schizophrenia	0	42 ^a	16	
Developmental, psychosocial and learning problems	1	1	1	
Total	11	60	42	

^aThree systematic reviews in the Schizophrenia Group and one in the Depression, Anxiety and Neurosis Group did not provide adequate information for indicate "acceptable" extraction but reported attrition rates below 5%, which were too low to raise accuracy concerns.



Why the NEMO project was proposed? (5)

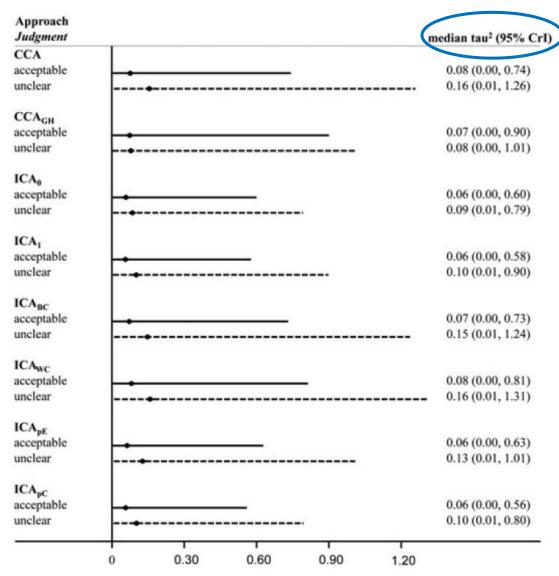


Inaccurate extraction of trial data may:

- lead to biased metaanalysis results;
- hinder the performance
 of methods to handle
 MOD.

From a systematic review with a meta-analysis with ,acceptable' extraction and the same meta-analysis on LOCF data (,unclear').

Why the NEMO project was proposed? (6)



Inaccurate extraction of trial data may:

- lead to biased metaanalysis results;
- hinder the performance
 of methods to handle
 MOD.

From a systematic review with a meta-analysis with ,acceptable' extraction and the same meta-analysis on LOCF data (,unclear').

'What is the reporting quality of systematic

reviews of multiple interventions in terms of

MOD?'

Search strategy

Published databases considered:

- Zarin et al. → inception 14/04/2015; ≥ 4 interventions
- Tan et al.
 → 1997 07/2012; ≥ 3 interventions
- Nikolakopoulou et al. → inception 12/2012; ≥ 4 interventions

Our own search:

08/2012 – 03/2017; ≥ 3 interventions; using Petropoulou et al.

search strategy in MEDLINE, EMBASE and the CDSRs.



Eligibility strategy (1)

Following Zarin et al.:

- ✓ Systematic reviews of randomized controlled trials (RCTs);
- ✓ No language restriction;
- Diagnostic test accuracy studies;
- Genetic studies;
- Observational studies;
- Mixture of RCTs and observational studies;
- Number of included trials < number of interventions.</p>

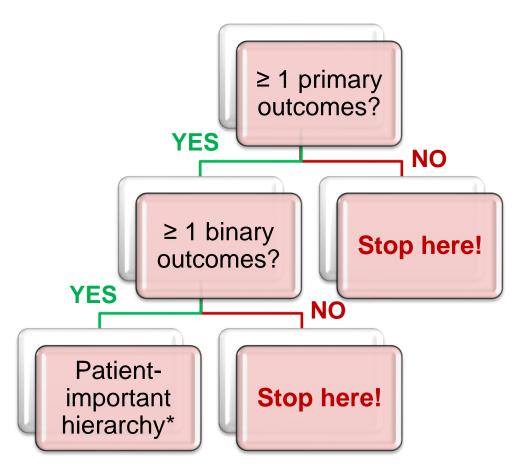
Eligibility strategy (2)

Our own criteria:

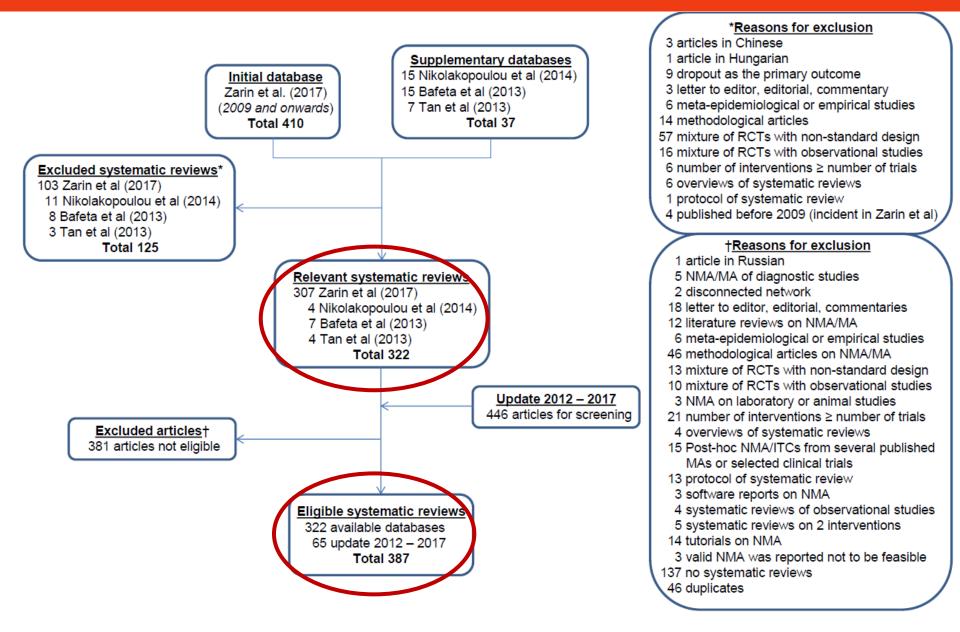
- ✓ At least 3 interventions are investigated;
- ✓ Systematic reviews published from 2009 and onwards;
 - * the new Cochrane risk of bias tool was published during 2009
- RCTs with non-standard design;
 - e.g. quasi, crossover, factorial, cluster, split-mouth.
- Overviews of systematic reviews;
- MOD investigated as single primary outcome.

Eligibility strategy (4)

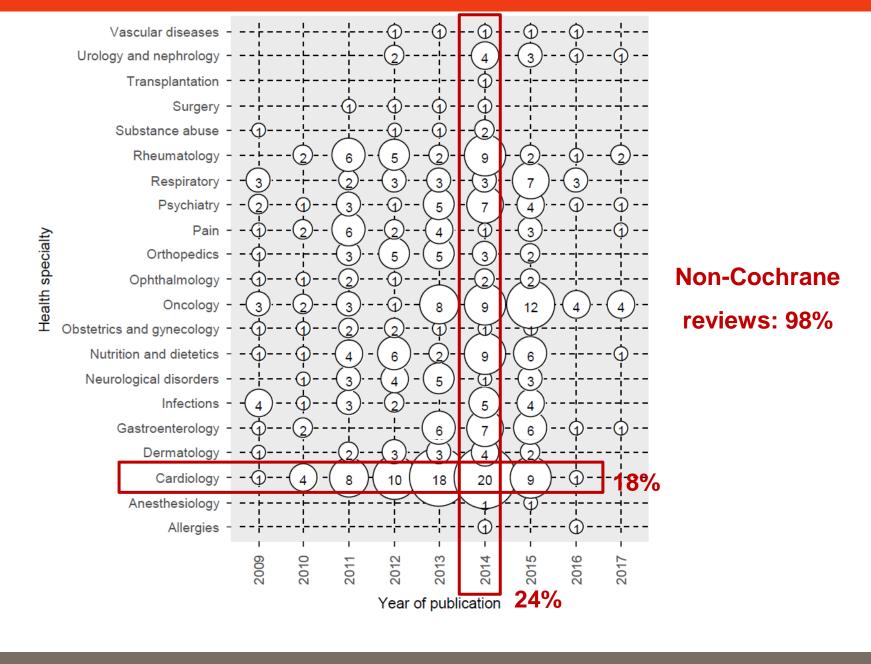
For each eligible Systematic review:



*Akl EA, Kahale LA, Agarwal A, et al. Impact of missing participant data for dichotomous outcomes on pooled effect estimates in systematic reviews: a protocol for a methodological study. Syst Rev 2014;3:137

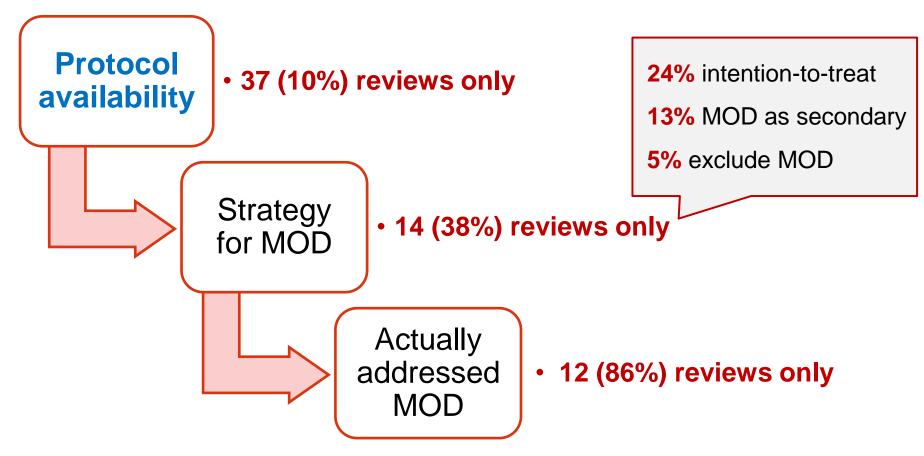


387 selected Systematic Reviews

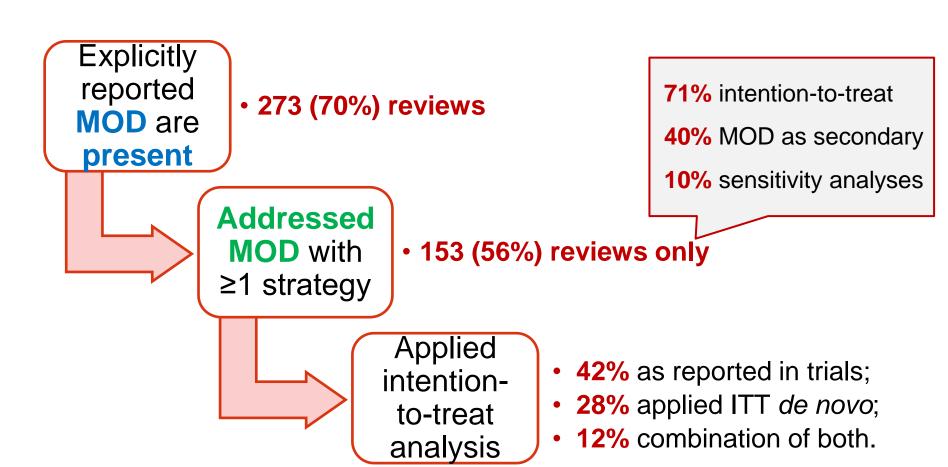


Addressing MOD in the promocol

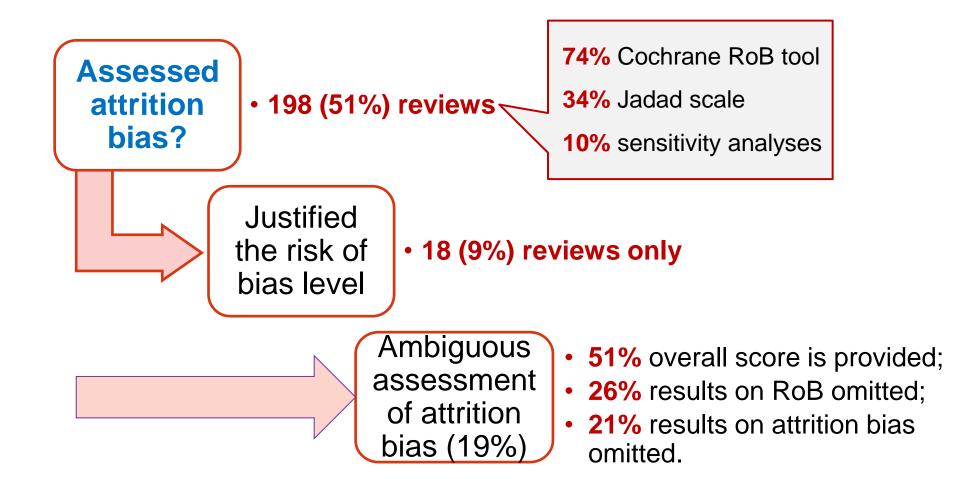




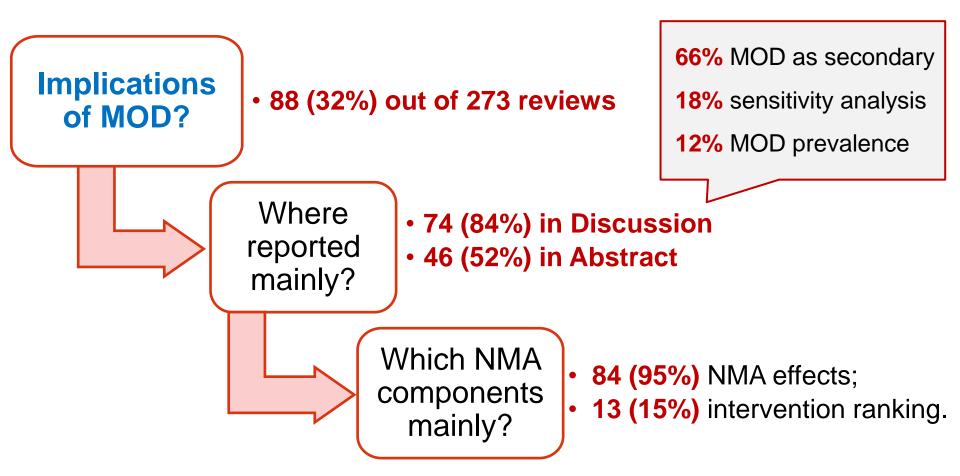
Addressing MOD in the REVIEW



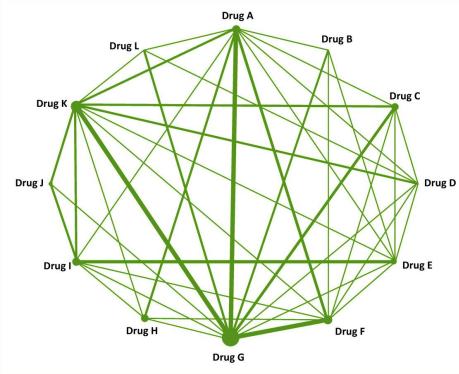
Addressing MOD in the REVIEW



Addressing MOD in the REVIEW



273 NMAs with present MOD



https://www.quantics.co.uk/blog/software-bayesian-network-meta-analyses/

Addressing MOD in the NMA

Strategy explicitly reported

- in 113 (40%) reviews only
 - 92% claimed intention-to-treat

60% as reported in trials

22% exclusion of MOD

17% intention-to-treat

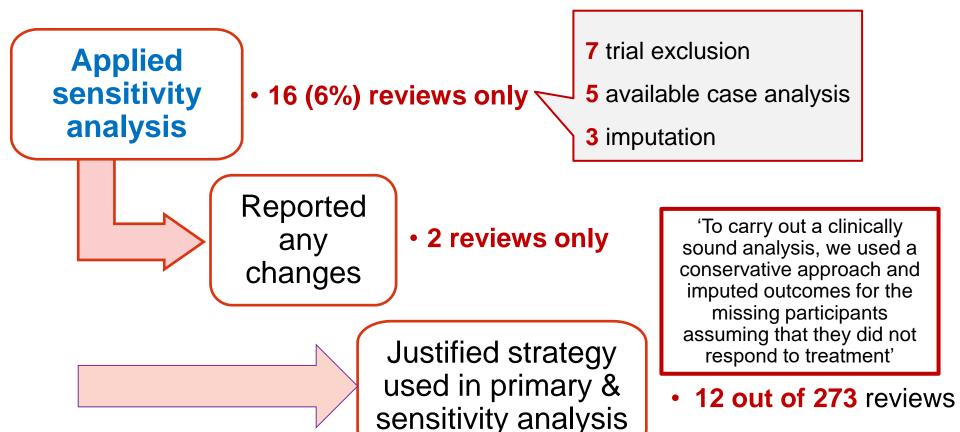
Actual strategy judged

- agreed with 14 (12%) reviews
- able to judge in 95 (35%) reviews

Intention-totreat analysis (16; 17%)

- 8 all MOD as non-events;
- 1 all MOD as events;
- 7 no scenario is provided.

Addressing MOD in the NMA



Cipriani A, Furukawa TA, Salanti G, et al. Comparative efficacy and acceptability of 12 new-generation antidepressants: a multiple-treatments meta-analysis. Lancet. 2009;373(9665):746-58.

Conclusions

- ☐ The quality of reporting and handling MOD in systematic reviews with NMAs is particularly inadequate!
- □ Reviewers remain unaware of the presence and importance of MOD in systematic reviews of multiple interventions.
- □ Poor handling of MOD attests to limited knowledge of the reviewers regarding the existing relevant methodology.
- □ Education amongst reviewers, peer reviewers and journal editors is deemed necessary!

'Do different models for binary MOD agree in

terms of core components of NMA?'

Informative Missingness Odds Ratio (IMOR)

Pattern-mixture model (White at al., 2008)

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

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

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

$$\sigma_{ik}^2 = \sigma^2 = 1$$



Prior structures of logIMOR (1)

Identical structure*

... log IMORs are considered identical depending on further assumptions on whether missingness mechanisms are:

- common in the whole network
- intervention-related
- trial-related

Assumption	Prior specification
Common-within-network	$\delta_{ik} = \delta, \delta \sim N(0,1)$
Intervention-specific	$\delta_{ik} = \delta_{t_{ik}}, \delta_{t_{ik}} \sim N(0,1)$
Trial-specific	$\delta_{ik} = \delta_i, \delta_i \sim N(0,1)$

^{*}Turner NL, Dias S, Ades AE, Welton NJ. A Bayesian framework to account for uncertainty due to missing binary outcome data in pairwise meta-analysis. Stat Med 2015;34:2062-80

Prior structures of logIMOR (2)

Hierarchical structure (Turner et al. 2015)

... log IMORs are considered different yet related to each other by borrowing information either across:

- the whole network
- different trials for the same intervention
- different interventions in the same trial

Assumption	Prior specification
Common-within-network	$\delta_{ik} \sim N(\Delta, \sigma^2)$
	$\Delta \sim N(0,1), \ \sigma \sim U(0,1)$
Intervention-specific	$\delta_{ik} \sim N(\Delta_{t_{ik}}, \sigma_{t_{ik}}^2)$
•	$\Delta_{t_{ik}} \sim N(0,1), \ \sigma_{t_{ik}} \sim U(0,1)$
Trial-specific	$\delta_{ik} \sim N(\Delta_i, \sigma_i^2)$
	$\Delta_i \sim N(0,1), \ \sigma_i \sim U(0,1)$

Missingness scenarios

$$\delta_{t_{ik}} \sim N(\Delta_{t_{ik}}, 1)$$

Scenario*	Interpretation	Abbrev.
$e^{\Delta_{ik}}=1$	Missing at random more likely	MAR
$e^{\Delta_{t_{ik}}}=2$	More missing are events	MME
$e^{\Delta_{t_{ik}}}=1/2$	More missing are non-events	MMNE
$e^{\Delta_{t_{ik}}}=2\ \&\ e^{\Delta_R}=1/2$	MME for non-references; MMNE for reference	ВС
$e^{\Delta_{t_{ik}}}=1/2 \& e^{\Delta_R}=2$	MMNE for non-references; MME for reference	WC

^{*}Spineli LM, Higgins JP, Cipriani A, Leucht S, Salanti G. Evaluating the impact of imputations for missing participant outcome data in a network meta-analysis. Clin Trials 2013;10:378-88



Classification of networks based on MOD (1)

Prevalence: low, moderate, large

- Low MOD → 12/29 (41%) networks
- Moderate and balance MOD → 14/29 (48%) networks
- Moderate and imbalance MOD → 2/29 networks
- Large and imbalance MOD → 1/29 networks

Characterising networks

Use of the 'five to twenty' rule based on Sackett et al (1997) in each network:

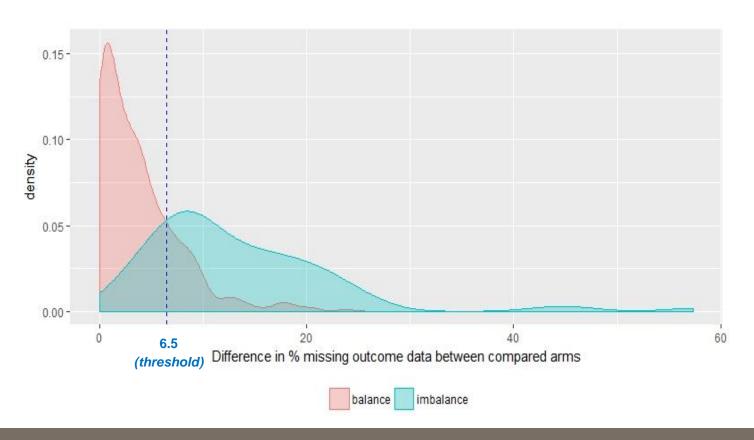
- low → median of total %MOD across trials < 5%
- high → median of total %MOD across trials > 20%
- moderate → otherwise

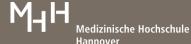


Classification of networks based on MOD (2)

Balance vs. Imbalance in the compared arms

Calculate difference in %MOD between compared arms within each trial:





Model specification & Presentation of results



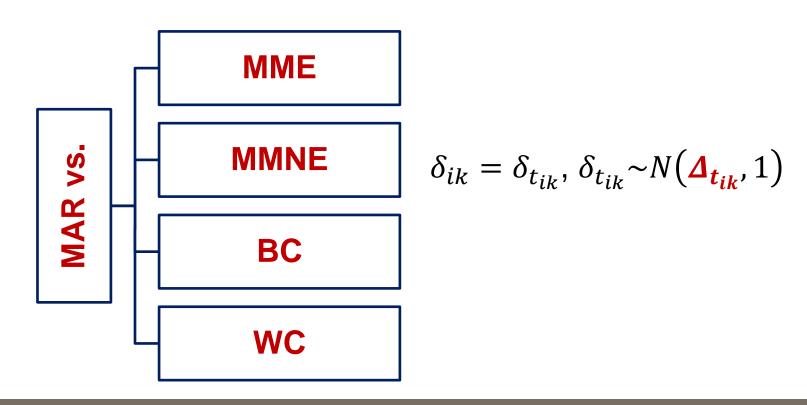


- Incorporation of log IMOR as described in Turner et al (2015);
- > For location parameters use of N(0, 10 000);
- Use of empirical priors on between-trial variance based on Turner et al (2015);
- > 3 chains for 10 000 updates and 1 000 burn-in.
- Bland-Altman plots to investigate the agreements in the compared methods

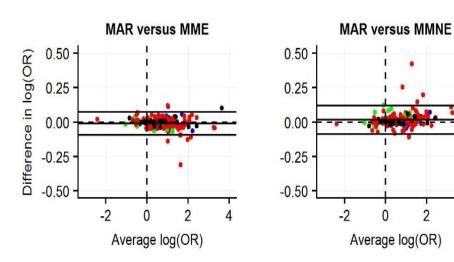
Research questions (1)

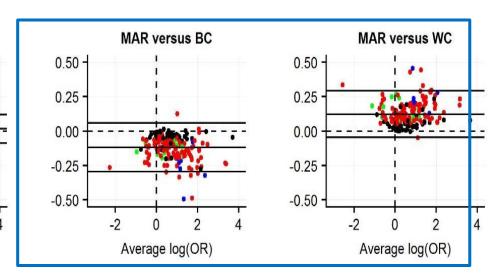
Agreement between on average MAR and extreme scenarios

Identical log IMORs with intervention-specific normal prior distribution



Posterior mean





Research questions (2)

Agreement between accountability & ignorance of uncertainty due to MOD

➤ Identical log IMORs with intervention-specific normal prior distribution

$$\delta_{ik} = \delta_{t_{ik}}, \, \delta_{t_{ik}} \sim N(\Delta_{t_{ik}}, 1)$$

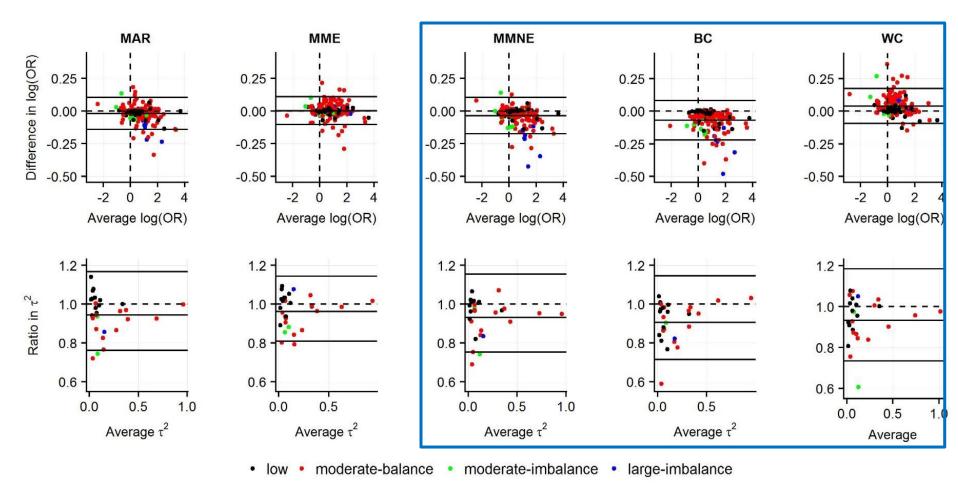
Scenarios considered: MAR, MME, MMNE, BC and WC

uncertainty about the scenario considered

Accounted for vs.	ignored
$\delta_{t_{ik}} \sim N(\Delta_{t_{ik}}, 1)$	$\delta_{t_{ik}} = \Delta_{t_{ik}}$

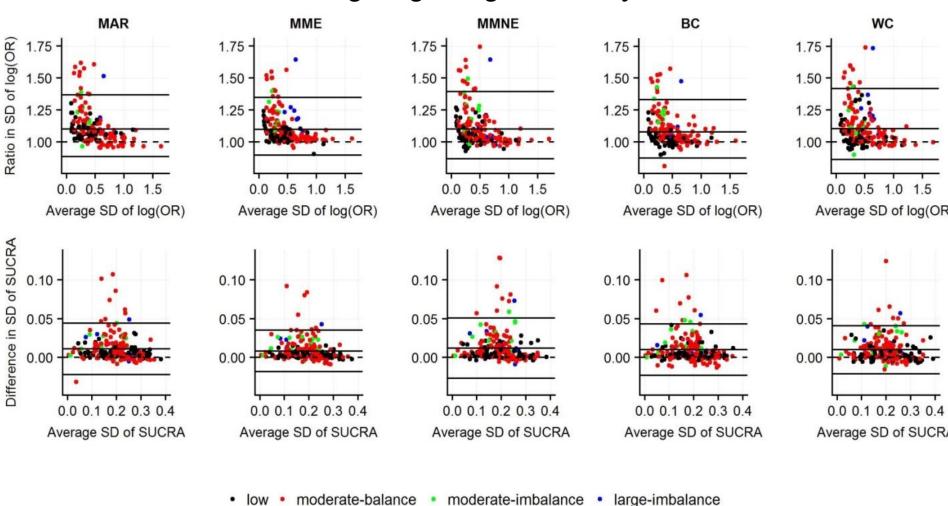
Posterior mean/median

Accounting vs ignoring uncertainty due to MOD



Posterior standard deviation

Accounting vs ignoring uncertainty due to MOD

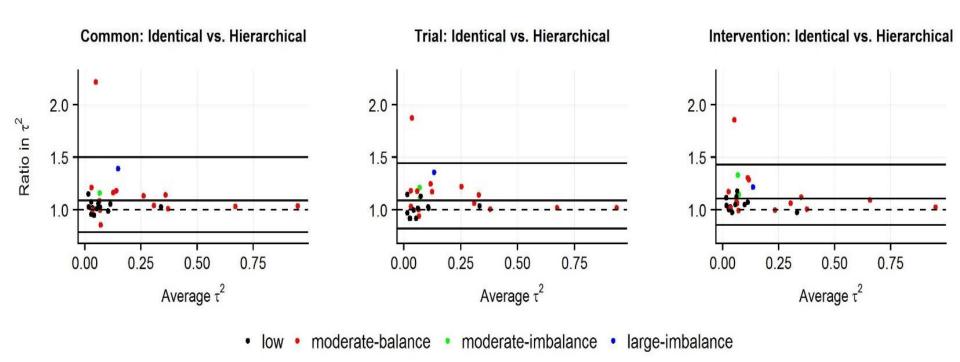


Research questions (3)

Agreement between identical and hierarchical structure of log IMOR

- MAR on average scenario
- ➤ In both structures use of normal prior distribution on log IMORs assumed:
 - ☐ Common-within network
 - ☐ Intervention-specific
 - ☐ Trial-specific

Posterior median



Take-home message

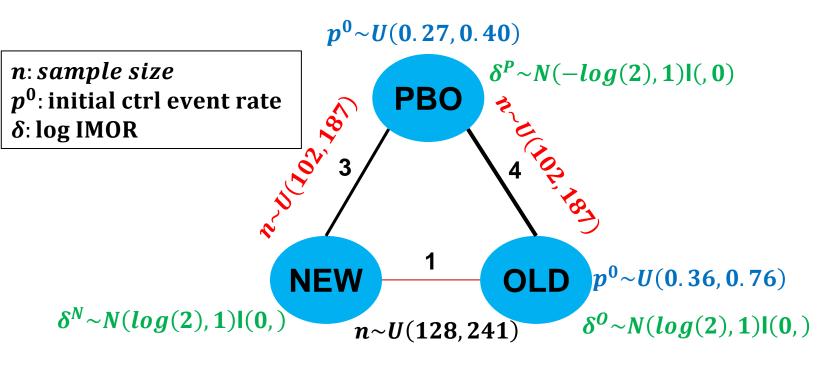
- ✓ Use 'on average MAR' for primary analysis and assumptions with clinical plausibility as sensitivity analyses.
- ✓ Model the missingness mechanism via the IMOR parameter in order to accommodate the uncertainty about the missingness scenarios considered.
- Avoid imputing or excluding MOD <u>either</u> before analysis <u>or</u> by fixing the missingness parameter!
 - Simulation study follows ...

Which factors may affect the performance of

different models for binary MOD in terms of

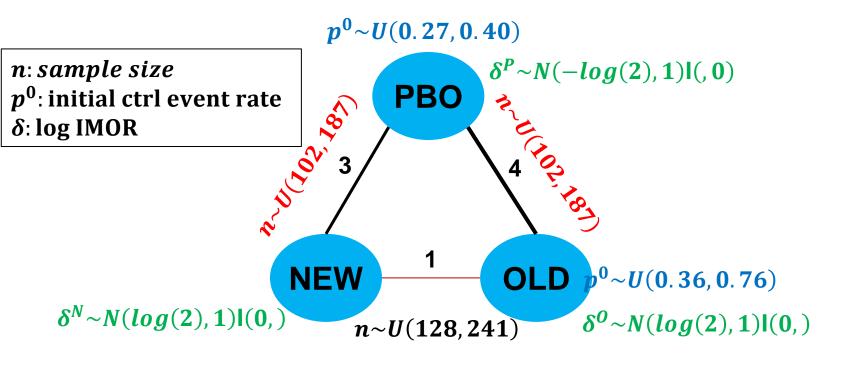
core components of NMA?

Simulation set-up (1)



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

Simulation set-up (2)



		small/low	substantial	$LOR_{NP} = log(2)$
T ²	2	all-cause mortality	generic health	$LOR_{OP} = log(1.5)$
IF	F	t(0, 0.44, df = 3)	t(1, 0.44, df = 3)	$LOR_{NO} = LOR_{NP} - LOR_{OP} + IF$

Hartung & Knapp data generating model

1. Obtain initial event risks for the experimental arms, p^E :

$$p^{E,0} = \frac{p^{C,0} \cdot exp(LOR)}{1 - p^{C,0} \cdot p^{C,0} \cdot exp(LOR)}$$

2. Obtain initial log odds for experimental and control arms, respectively:

$$logit^{E,0} = log\left(rac{p^{E,0}}{1-p^{E,0}}
ight)$$
 and $logit^{C,0} = log\left(rac{p^{C,0}}{1-p^{C,0}}
ight)$

3. Generate **true** log odds for experimental and control arms, respectively:

$$logit^E \sim N\left(logit^{E,0}, \frac{2\tau^2}{3}\right)$$
 and $logit^C \sim N\left(logit^{C,0}, \frac{\tau^2}{3}\right)$

4. Finally, obtain **true** event risks for experimental and control arms, respectively:

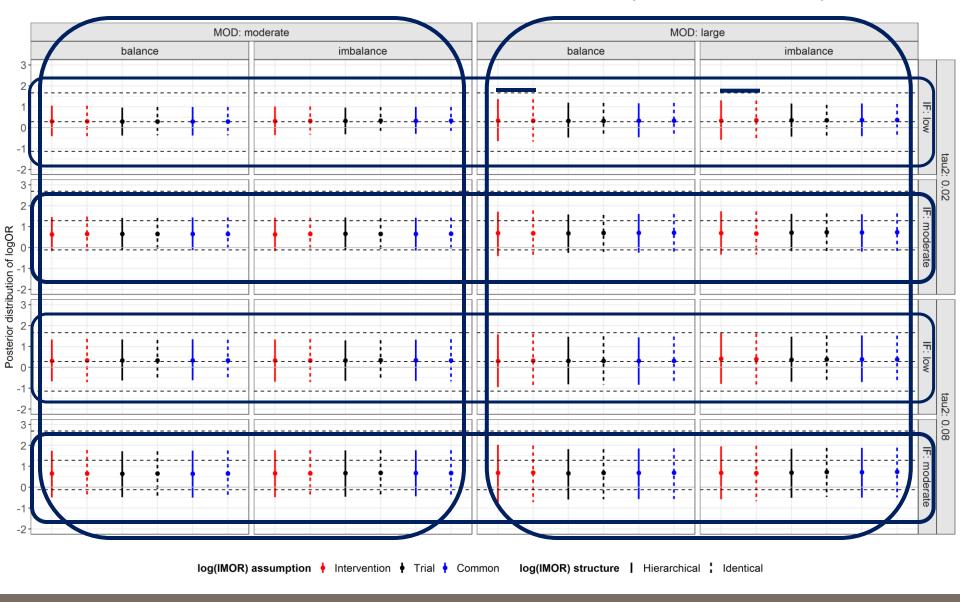
$$p^E = rac{1}{1 + exp(-logit^E)}$$
 and $p^C = rac{1}{1 + exp(-logit^C)}$

Model specification & Presentation of results

- Bayesian random-effects NMA model by Dias et al (2013);
- Application of the node-splitting model as developed by Dias et al (2010);
- Incorporation of log IMOR as described in Turner et al (2015);
- Pattern-mixture model for identical and hierarchical logIMOR (common-within-network, trial-specific, and intervention-specific).
- > For **location parameters**, we used **N(0, 10 000)**;
- > For between-trial variance, we used N(0, 1)I(0,);
- For logIMOR, we used N(0,1) (on average MAR) in all arms;
- > 1 000 triangles, 3 chains for 20 000 updates and 2 000 burn-in;
- ➤ Interval plots for LOR (new vs old) and between-trial variance to correctly present results from Bayesian methods (prior distribution is presented with dotted parallel lines).

JAGS/R2jags

Posterior distribution of LOR (new vs old)



Posterior distribution of between-trial variance



Limitations of the study

- ☐ We considered a simple network of **3 interventions and 2-arm trials**;
- ➤ Simulating a complex network with multi-arm trials* will shed more light on implications of different prior structures of log IMOR on NMA estimates;
- ☐ We investigated only **frequent events**;
- Carpenter et al (2007): 'if an event is rare, missing data on very few patients can markedly alter estimated event rates' → affect NMA estimates.
- □ The degree of unbalanced MOD was much smaller than the total extent of MOD in each trial → We observed that much in the empirical study!
- ➤ Larger imbalance of MOD may have resulted in more imprecise log OR under common-within-network and trial-specific prior structures.

^{*}Seide SE, Jensen K, Kieser M. Simulation and data-generation for random-effects network meta-analysis of binary outcome. Stat Med. 2019 May 9. doi: 10.1002/sim.8193. [Epub ahead of print]



Take-home message

- □ Always decide <u>at the protocol</u> on the proper prior structure of log IMOR that best aligns with the condition and interventions investigated;
- Misspecification of the prior structure may lead to spurious estimation of the uncertainty around log OR with implications for the conclusions;
- □ Both identical and hierarchical structure may be considered in the context of a sensitivity analysis;
- ☐ Though, we regard **hierarchical structure to be more plausible** in practice.
- ☐ Results may be also generalized to conventional meta-analyses with binary outcome.
- > Inferences are greatly restricted by the scenarios considered!!!

Future work in NEMO

- To assess and refine methods to address continuous MOD in onestage (empirical and simulation study);
- Develop a graphical tool for comprehensive sensitivity analyses on plausible scenarios for MOD;
- Develop an R-package to ,accommodate all strategies that have been investigated within the project;
- Submit Habilitation on ,NMA with MOD'.

Transparent documentation of MOD from trials

Discuss the implications of MOD on all core NMA/MA paramaters

Decide in the protocol the models and assumtions to handle MOD

Model MOD using on average MAR in primary analysis

Sensitivity
analyses using
more stringer
scenarios

Publications within NEMO

- 1. Spineli LM, Kalyvas C, Pateras K. Participants' outcomes gone missing within a network of interventions: Bayesian modeling strategies. *Stat Med* 2019. DOI: 10.1002/sim.8207 [in press]
- 2. Spineli LM. An empirical comparison of Bayesian modelling strategies for missing binary outcome data in network meta-analysis. *BMC Med Res Methodol* 2019;19(1):86.
- 3. Spineli LM. Modeling missing binary outcome data while preserving transitivity assumption yielded more credible network meta-analysis results. *J Clin Epidemiol* 2019;105:19-26.
- Spineli LM, Yepes-Nuñez JJ, Schünemann H. A systematic survey shows that reporting and handling of missing outcome data in networks of interventions is poor. *BMC Med* Res Methodol 2018;18(1):115.
- 5. Spineli LM. Missing binary data extraction challenges from Cochrane reviews in mental health and Campbell reviews with implications for empirical research. *Res Synth Methods* 2017;8(4):514-525.

References (1)

- Akl EA, Carrasco-Labra A, Brignardello-Petersen R, et al. Reporting, handling and assessing the risk of bias associated with missing participant data in systematic reviews: a methodological survey. BMJ Open 2015;5:e009368.
- Bafeta A, Trinquart L, Seror R, Ravaud P. Analysis of the systematic reviews process in reports of network meta-analyses: methodological systematic review. BMJ 2013;347:f3675.
- Carpenter JR, Kenward MG. Missing data in randomised controlled trials—a practical guide. Health Technology Assessment Methodology Programme. Birmingham; 2007.
- Dias S, Sutton AJ, Ades AE, Welton NJ. Evidence synthesis for decision making 2: a generalized linear modeling framework for pairwise and network meta-analysis of randomized controlled trials. Med Decis Making 2013;33:607–617.
- Dias S, Welton NJ, Caldwell DM, Ades AE. Checking consistency in mixed treatment comparison meta-analysis. Stat Med 2010;29:932-44.
- Nikolakopoulou A, Chaimani A, Veroniki AA, et al. Characteristics of networks of interventions: A description of a database of 186 published networks. PLoS One 2014;9:1–10.
- Petropoulou M, Nikolakopoulou A, Veroniki AA, et al. Bibliographic study showed improving statistical methodology of network meta-analyses published between 1999 and 2015. J Clin Epidemiol 2017;82:20–28.

References (2)

- Sackett DL, Richardson WS, Rosenberg W, Haynes RB. Evidence-based medicine: how to practice and teach EBM. New York: Churchill Livingstone, 1997
- Spineli LM, Pandis N, Salanti G. Reporting and handling missing outcome data in mental health: A systematic review of cochrane systematic reviews and meta-analyses. Res Synth Methods 2015;6:175–87.
- Tan SH, Bujkiewicz S, Sutton A, et al. Presentational approaches used in the UK for reporting evidence synthesis using indirect and mixed treatment comparisons. J Health Serv Res Policy 2013;18:224–32.
- Turner NL, Dias S, Ades AE, Welton NJ. A Bayesian framework to account for uncertainty due to missing binary outcome data in pairwise meta-analysis. Stat Med 2015;34:2062-80.
- van Valkenhoef G, Dias S, Ades AE, Welton NJ. Automated generation of nodesplitting models for assessment of inconsistency in network meta-analysis. Res Synth Methods 2016;7:80-93.
- White IR, Higgins JP, Wood AM. Allowing for uncertainty due to missing data in metaanalysis--part 1: two-stage methods. Stat Med 2008;27:711-727
- Zarin W, Veroniki AA, Nincic V, et al. Characteristics and knowledge synthesis approach for 456 network meta-analyses: a scoping review. BMC Med 2017;15:3.

Thank you for your attention!