

### Statistical considerations in indirect comparisons and network meta-analysis

Said Business School, Oxford, UK March 18-19, 2013

Cochrane Comparing Multiple Interventions Methods Group Oxford Training event, March 2013

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#### Handout S10-L

# Presenting and evaluating the evidence from network meta-analysis

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### **Reporting a network meta-analysis**

- PRISMA items to report in meta-analysis Moher at al PLoS 2009
- Network meta-analysis needs more information
- Reporting in network meta-analysis is suboptimal See Donegan S et al. PLoS One 2010, Song F et al. BMJ 2009

Section/Topic	#	Checklist Item
TITLE		
Title	1	Identify the report as a systematic review, meta-analysis, or bo
ABSTRACT		
Structured summary	2	Provide a structured summary including, as applicable: backgro criteria, participants, and interventions; study appraisal and syr and implications of key findings; systematic review registratior
INTRODUCTION		
Rationale	3	Describe the rationale for the review in the context of what is
Objectives	4	Provide an explicit statement of questions being addressed wi comparisons, outcomes, and study design (PICOS).
METHODS		
Protocol and registration	5	Indicate if a review protocol exists, if and where it can be acces registration information including registration number.
Eligibility criteria	6	Specify study characteristics (e.g., PICOS, length of follow-up) $a$ language, publication status) used as criteria for eligibility, givi
Information sources	7	Describe all information sources (e.g., databases with dates of additional studies) in the search and date last searched.

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# **Reporting a network meta-analysis (1)**

- Describe a **clear rationale regarding the choice of interventions**, with respect to the condition and outcomes of interest.
  - Treatments need to be competing or alternative treatments for the condition of interest;
  - In principle, treatments should be 'jointly randomizable'; meaning that a trial including all treatments would be clinically reasonable.
- Describe the inclusion criteria for patients and condition
- Report whether treatments will be included that are not of direct interest but may provide useful indirect evidence (as long as the risk of inconsistency and bias does not outweigh their value!)

# **Reporting a network meta-analysis (2)**

- Report whether transitivity is likely to hold in the set of trials you expect to find
- Describe the intended **strategy to evaluate consistency** (including statistical and strategic approaches).
- Think about **possible sources of heterogeneity and inconsistency** by describing *a priori* any effect modifiers that may vary across studies and comparisons.
- Predefine whether network meta-analysis will be used to combine all identified studies (or whether planned for only on a subset of studies).
- Register the protocol?

# **Reporting a network meta-analysis (3)**

- Identify and consider all studies that evaluate at least two of the treatments of interest. (report flow chart)
- Report on heterogeneity (e.g. calculate the heterogeneity variance, calculate I<sup>2</sup>, derive predictive intervals, perform network meta-regression).
- Report on inconsistency (e.g. perform loop-based tests, implement inconsistency models, compare the fit of the models, perform network meta-regression).
- Report on estimation of the model fit of and the fit of alternative models, and check sensitivity to analysis assumptions.

# **Reporting a network meta-analysis (4)**

- Present **graphically the network** of trials (with information about number of studies informing each comparison).
- Provide (in an appendix if needed) individual **study data**.
- Present estimates of direct and mixed effects if possible and appropriate.
- Present the effect sizes for each treatment versus a pre-defined reference treatment as obtained from the network meta-analysis model.
- Explicitly **discuss issues of bias** and the validity of findings in the light of the presence/absence of consistency.

### Graphical presentation of the network STATA function networkplot

- Present the network with information on evidence characteristics
  - Nodes and edges: size can be proportional to number of studies, number of patients, mean age of participant in studies, price of the drug etc.



### Graphical presentation of the network STATA function networkplot

- Present the network with information on evidence characteristics
  - Nodes and edges: size can be proportional to number of studies, number of patients, mean age of participant in studies, price of the drug etc.
  - Nodes and edges color can be used to present risk of bias characteristics
    Haloperidol



# Numerical presentation of results from network meta-analysis

- Typically effect sizes (and their uncertainty) for all pairwise comparisons are reported
- With many treatments judgments based on pairwise effect sizes are difficult to make
- Example: Antidepressants



Efficacy (i	response rate	) (95% CI)	Comp	arison 🔛	Acceptability	(dropout rat	e) (95% CI)				
BUP	1.00 (078-1.28)	0.75 (0.55-1.01)	1.06 (0.86-1.32)	0.89 (074-1.08)	0.73 (0.53-1.00)	0.87 (0.58-1.24)	0.87 (0.66-1.14)	0.81 (0.65-1.00)	(0. <u>45-0.86</u> )	1.01 (0.82-1.27)	0.84 (0.68-1.02)
0.98 (0.78-1.23)	сп	0.75 (0.55-1.02)	1.07 (0.86-1.31)	0.90 (073-1.09)	<u>0.73</u> (0.54-0.99)	0.87 (0.60-1.24)	0.87 (0.66-1.15)	0.81 (0.65-1.01)	(0. <u>45-0.84</u> )	1.02 (0.81-1.28)	0.84 (0.67-1.06)
1.09 (0.83-1.43)	1·12 (0·87-1·44)	DUL	<u>1.43</u> (1.09-1.85)	1·19 (0·91-1·57)	0.98 (0.67-1.41)	1.16 (0.77-1.73)	1.16 (0.83-1.61)	1.08 (0.84-1.40)	0.83 (0.57-1.22)	(1·01-1·83)	1.12 (0.84-1.50)
0.82 (0.67-1.01)	0.84 (0.70-1.01)	<u>0.75</u> (0.60-0.93)	ESC	0.84 (0.70-1.01)	0.69 (0.50-0.94)	0.81 (0.55-1.15)	0.81 (0.62-1.07)	(0. <u>62-0.93</u> )	0.58 (0.43-0.81)	0.95 (0.77-1.19)	0.78 (0.64-0.97)
1.08 (0.90-1.29)	1·10 (0·93-1·31)	0.99 (0.79-1.24)	(1·12-1·55)	FLU	0.82 (0.62-1.07)	0.97 (0.69-1.32)	0.97 (0.77-1.21)	0.91 (079-1.05)	(0. <u>53-0.9</u> 2)	1·14 (0·96–1·36)	0.94 (0.81-1.09)
1.10 (083-1.47)	1·13 (0·86–1·47)	1.01 (0.74-1.38)	<u>1.35</u> (1.02-1.76)	1.02 (0.81-1.30)	FVX	1.18 (0.76-1.75)	1.18 (0.87-1.61)	1·10 (0·84-1·47)	0.85 (0.57-1.26)	(1.03-1.89)	1·14 (0·86–1·54)
1.07 (0.77-1.48)	1.09 (0.78-1.50)	0.97 (0.69-1.38)	1.30 (0.95-1.78)	0.99 (0.74-1.31)	0.97 (0.68 <b>-1</b> .37)	MIL	0.99 (0.69-1.53)	0.94 (0.68 <b>-1</b> .31)	0.72 (0.48-1.10)	1.17 (0.84-1.72)	0.97 (0.69-1.40)
0.79 (0.72-1.00)	0.80 (0.63-1.01)	<u>0.72</u> (0.54-0.94)	0.96 (076-1.19)	0-73 (0-60-0-88)	(0.55-0.92)	0.74 (0.53-1.01)	MIR	0.93 (0.75-1.17)	0.72 (0.51-1.03)	1.17 (0.91-1.51)	0.97 (0.76-1.23)
1.06 (0.87-1.30)	1.08 (0.90-1.30)	0.97 (0.78–1.20)	<u>1·30</u> (1·10-1·53)	0.98 (0.86-1.12)	0.96 (0.76-1.23)	1.00 (0.74-1.33)	<u>1·35</u> (1·11-1·64)	PAR	0.77 (0.56-1.05)	<u>1·25</u> (1·04-1·52)	1.03 (0.86–1.24)
(1.20-2.16)	(1.25-2.14)	<u>1.46</u> (1.05-2.02)	<u>1.95</u> (1.47-2.59)	(1·16-1·90)	(1·03-2·02)	(1·03-2·18)	<u>2.03</u> (1.52-2.78)	(1·16-1·98)	REB	(1·19-2·24)	1.34 (0.99-1.83)
0.87 (0.72-1.05)	0.88 (072-1-07)	0.79 (0.62-1.01)	1.06 (0.88–1.27)	<u>0-80</u> (0-69-0-93)	0.79 (0.61-1.01)	0.81 (0.60-1.11)	1·10 (0·90-1·36)	<u>0-82</u> (0-69-0-96)	<u>0.54</u> (0.41-0.71)	SER	0.82 (0.67-1.00)
0.85 (0.70-1.01)	0.86 (0.71-1.05)	<u>0.77</u> (0.60-0.99)	1.03 (0.86-1.24)	(0. <u>68-0.9</u> 0)	<u>0.77</u> (0.59-0.99)	0.79 (0.58-1.08)	1.08 (0.87-1.33)	(0. <u>67</u> -0.94)	0- <u>53</u> (0-40-0-69)	0.98 (0.82-1.16)	VEN

OR>1 means the treatment in top-left is better

# Numerical and graphical presentation of results from network meta-analysis

- With many treatments judgments based on pairwise effect sizes are difficult to make
- Example: Antidepressants
- Example: Antiplatelet regimens for serious vascular events



### Serious vascular events with antiplatelet regimens: Network pairwise estimates STATA function intervalplot



#### Odds Ratio for serious vascular event

# Odds-ratios for serious vascular events with antiplatelet treatments compared to placebo



# **Probabilities**

- Estimate for each treatment the probability of being the best
- This is straightforward within a Bayesian framework and fairly easy in frequentist setting (use re-sampling techniques)
  - Use the (posterior) distributions for all relative treatment effects
  - 'Draw' many random samples
  - Find which intervention outperforms in each sample
  - The number of times that an intervention ranks first out of the total number of random samples gives the P(best).

% probability	Α	В	С	D
Best	0.25	0.50	0.25	0.00
Second	0.25	0.25	0.50	0.00
Third	0.25	0.25	0.25	0.25
Last	0.25	0	0	0.75



Ranking for efficacy (solid line) and acceptability (dotted line). Ranking: probability to be the best treatment, to be the second best, the third best and so on, among the 12 comparisons).

% probability	Α	В	С	D
Best	0.25	0.50	0.25	0.00
Second	0.50	0.75	0.75	0.00
Third	0.75	1.00	1.00	0.25
Last	1.00	1.00	1.00	1.00



The areas under the cumulative curves for the four treatments of the example above are A=0.5 B=0.75 C=0.67 D=0.08

# Surface Under the Cumulative RAnking curve (SUCRA)

Use <u>posterior probabilities</u> for each treatment to be among the *n* - best options (cumulative probabilities)



# STATA function sucra



#### Serious vascular events with antiplatelet regimens



Salanti et al JCE 2011

# Ranking based on probabilities – caution is needed

- Using P(best) to rank treatments can be misleading!
- Ranking based on SUCRAs accounts better for the uncertainty in the estimated treatment effects
- SUCRAs are conditional on the set of treatments being compared
  - This means SUCRAs and possibly the ranking will change if a subset of the treatments are compared
- Ranking measures are not a substitute for relative treatment effects!
  - They cannot be interpreted clinically







# SUCRAs can be used to examine the impact of different models on ranking: ranking of treatments for RA



### **Present ranking of treatments for two outcomes**



# **Grading the evidence from a network**

- A network has two outputs: Pairwise network estimates and ranking
- We need to grade the evidence of each output separately
  - Grading each pairwise network estimate
  - Grade the ranking
- GRADE:
  - Study limitations
  - Indirectness
  - Inconsistency
  - Imprecision
  - Publication bias
- All this is still work in progress!!

# Interpret the evidence from a network: principals

- Pieces of direct evidence contribute to the network relative treatment effects and the ranking
- The contribution of each piece of direct evidence is different
  - Precise direct comparisons (e.g. comparisons with many, large studies) contribute more
  - "Central" comparisons contribute more
- The **exact contribution** of each piece of evidence can be derived mathematically (using a fixed effects network meta-analysis model)
  - Koenig J et al., 33rd Annual Conference of the International Society for Clinical Biostatistics



#### **Direct Comparisons in the Network**

		AB	AD	BC	BD	CD
Meta-Analysis Estimates	Mixed Estimates AB AD BC BD CD	 12.0	39.0	10.0	29.0	10.0
Network	Indirect Estimates AC	8.9	32.8	25.5	16.7	16.1

Percenta each	age contribution of direct estimate	<b>Direct Comparisons in the Network</b>					
<mark>STATA fur</mark>	nction netweight	AB	AD	BC	BD	CD	
ates	Mixed Estimates						
stim	AB	12.0	39.0	10.0	29.0	10.0	
С Ш S	AD	12.3	72.3	3.1	9.1	3.2	
lysi	BC	1.9	1.9	66.6	13.9	15.8	
Ana	BD	6.8	6.8	17.6	51.2	17.6	
Meta-	CD	4.1	4.1	35.0	30.9	25.9	
work	Indirect Estimates						
Net	AC	8.9	32.8	25.5	16.7	16.1	
Enti	re Network	8.0	26.6	25.3	25.0	15.1	
Inclu	uded Studies	2	1	7	5	4	



### **Evaluating Risk of Bias in the ranking**



# **Evaluating inconsistency (1)**

- Inconsistency: heterogeneity and network inconsistency
- <u>Heterogeneity for each network estimate</u>: as usual, plus consider the common r<sup>2</sup>
- <u>Heterogeneity for ranking</u>: consider the common τ<sup>2</sup>
- How large is a large  $\tau^2$ ? Compare it to its empirical distribution

Empirical distributions for the heterogeneity variance (T<sup>2</sup>) across different categories of (dichotomous) outcome and intervention comparison

	Pharmacological vs.	Pharmacological vs.	Non-Pharmacological	
Outcome type	Placebo	Pharmacological	(Any)	
	50% quantile	50% quantile	25% quantile	
	75% quantile	75% quantile	50% quantile	
All-cause	0.007	0.005	0.007	
mortality	1.017	1.014	1.02	
Semi-Objective	0.014	0.011	0.016	
Semi-Objective	1.05	1.04	1.058	
Subjective	0.34	1.10	0.045	
	1.12	3.28	1.14	

# **Evaluating inconsistency (2)**

- Inconsistency for each network estimate: Evaluate inconsistency placing emphasis on how much each piece of direct evidence fits together with the indirect evidence
  - apply node splitting or the loop-specific approach
- Inconsistency for ranking: Evaluate the assumption of consistency as a whole
  - Lu& Ades model
  - Design by treatment model
- Do NOT rely on statistical tests: Evaluate the transitivity assumption!

### **Our STATA functions**

• You can get the functions by typing in STATA

net from http://www.mtm.uoi.gr

intervalplot	Predictive intervals plot
ifplot	Inconsistency plot
netfunnel	Comparison-adjusted funnel plot
netweight	Contribution plot
networkplot	Network plot
sucra	Ranking plots for a single outcome using probabilities

Please cite our site <u>www.mtm.uoi.gr</u> and acknowledge **Anna Chaimani**, Dimitris Mavridis, Julian Higgins, Georgia Salanti

(we hope our paper will be submitted soon....)

# **References (Methods)**

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# **References (Applications)**

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