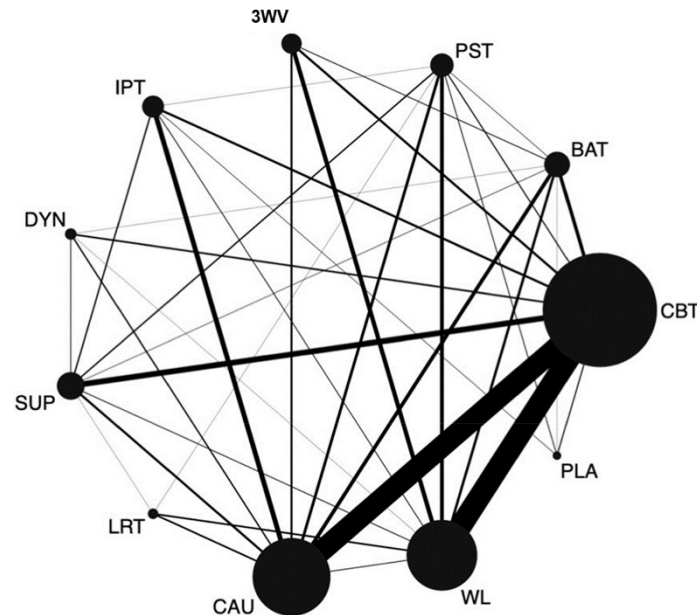


# Network Meta-analysis: Principles and applications



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# Disclaimer

- No conflicts of interest

# Objectives

- By the end of this session, we hope you will understand the:
  - What?
  - Why?
  - How? (in Stata)

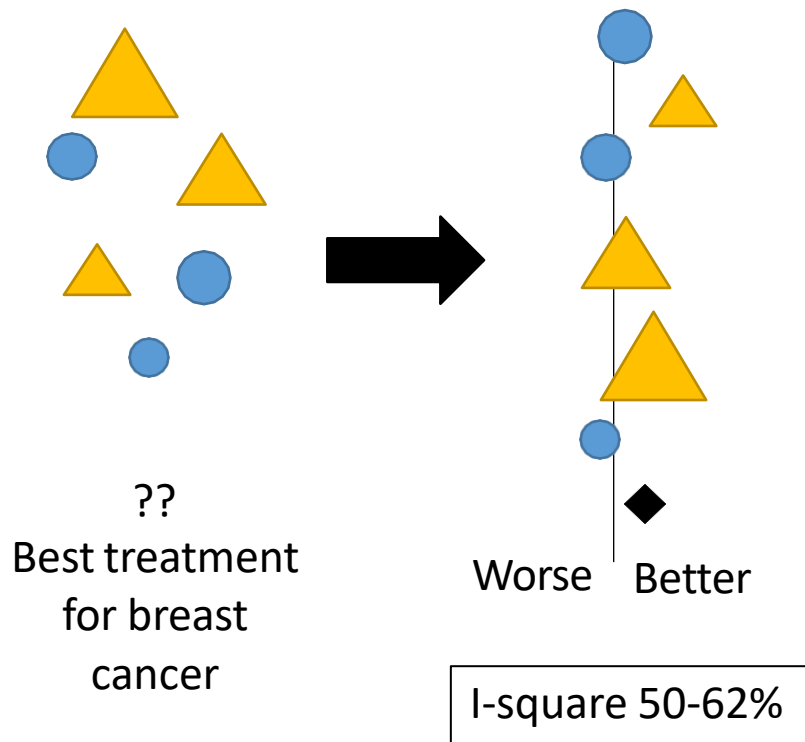
Of Network Meta-analysis

# Two big problems with modern medicine

- Contradictory studies on almost every topic
- Flood of new data

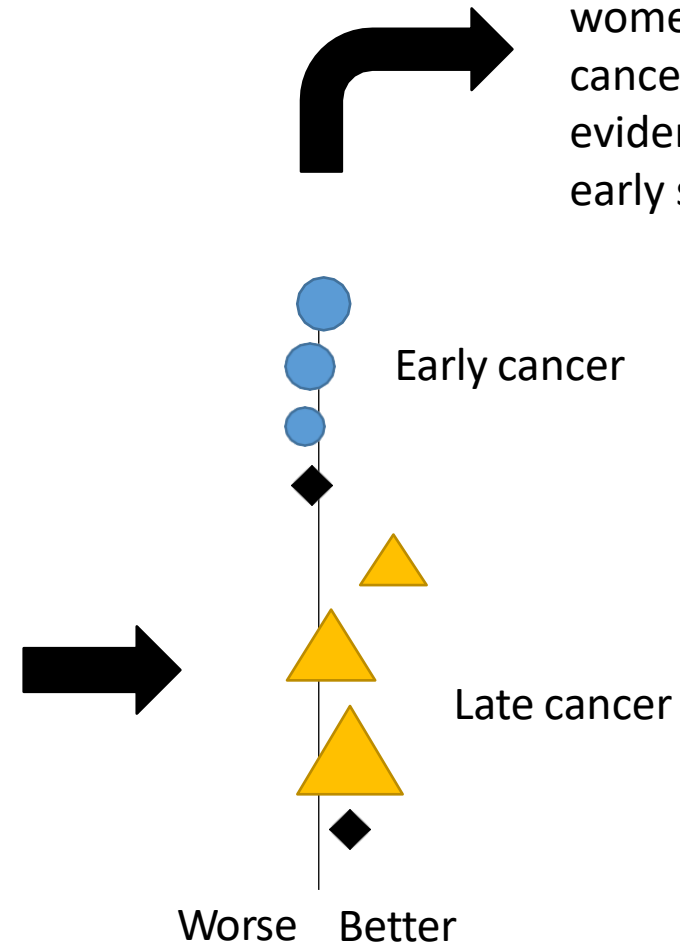


# The (quick) story of meta-analysis



Apple and oranges problem

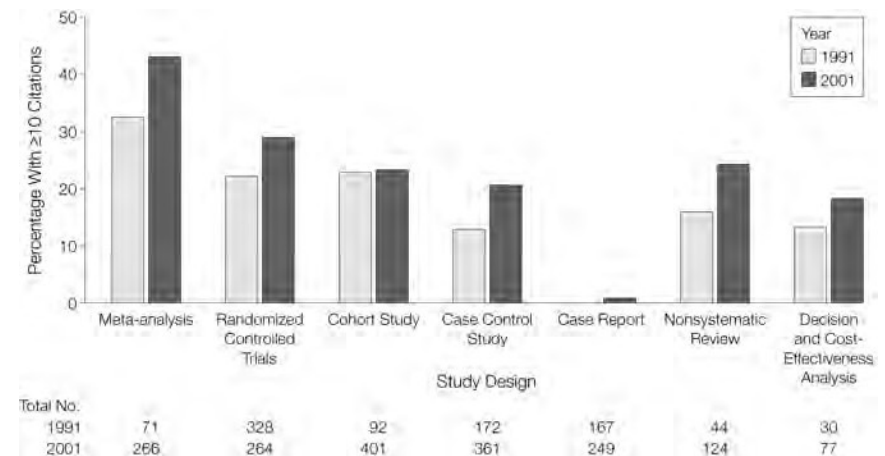
Using the force of heterogeneity to understand more!



Treatment for breast cancer appears to differ by cancer stage. Overall benefit for women with late cancer but absent evidence of benefit for early stages

# Huge advantages of meta-analysis

- Understand how different treatments work in different settings
- Find when treatments are harmful
- Compost huge volumes of data into something useable
- Understand the quality of the evidence
- Policy-making bodies love them
- Highly cited and influential

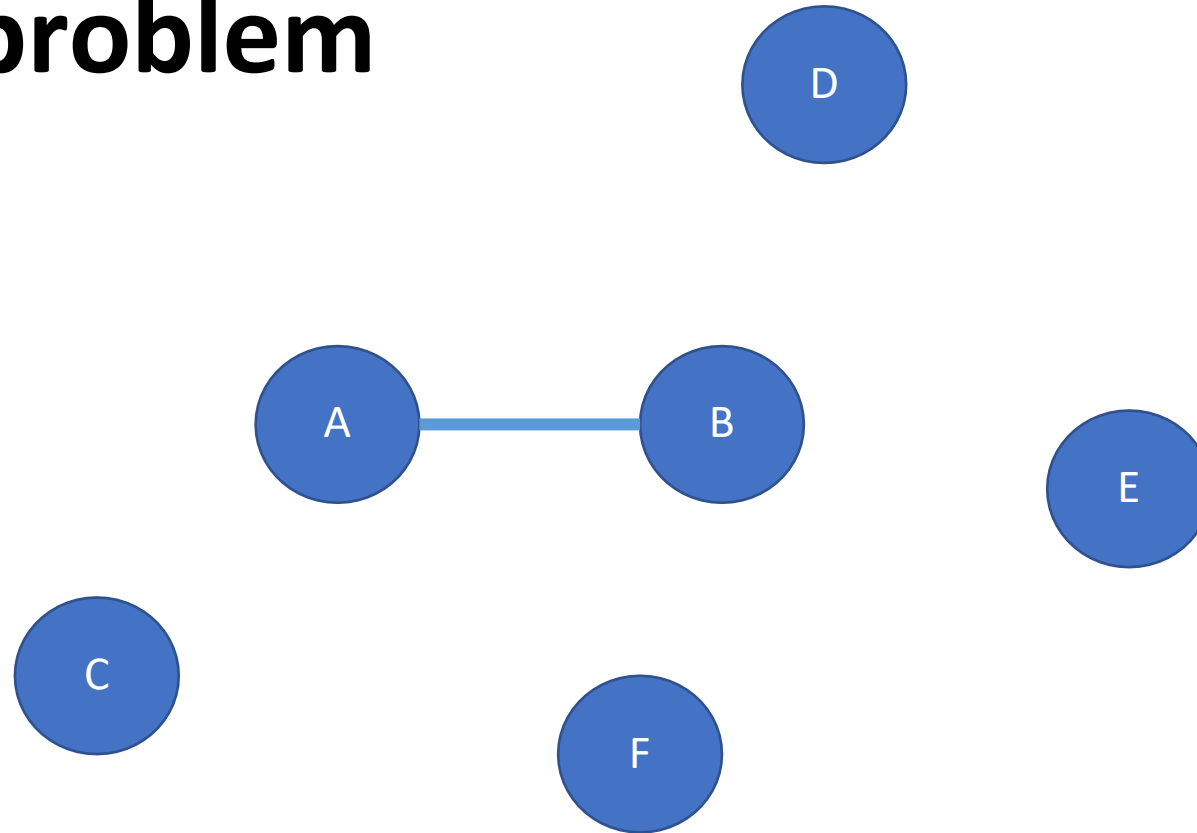


# (Old) New problem



Traditional meta-analysis  
Is treatment A better than treatment B?

# (Old) New problem

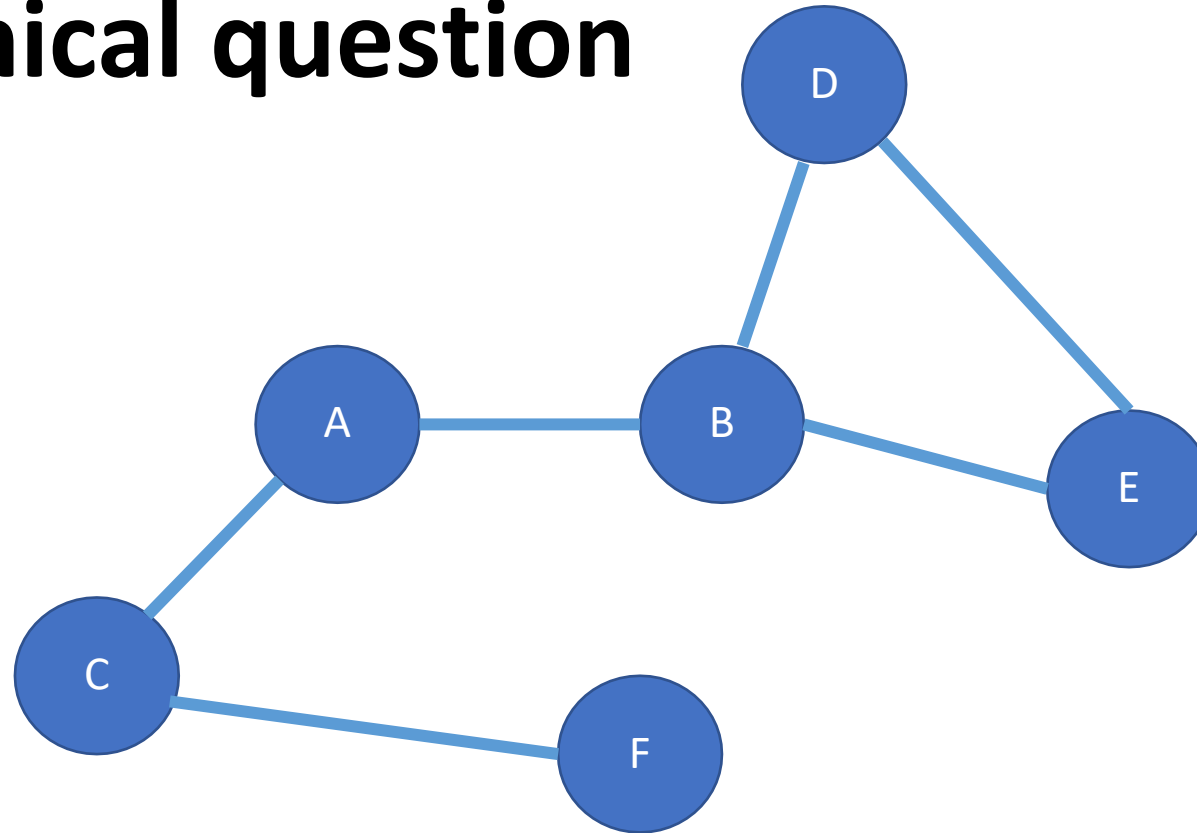


Traditional meta-analysis

Is treatment A better than treatment B?



# The real clinical question



Which of the six available treatments is the most effective and safest?  
Is treatment B better than treatment F?

# Networks: using indirect comparisons

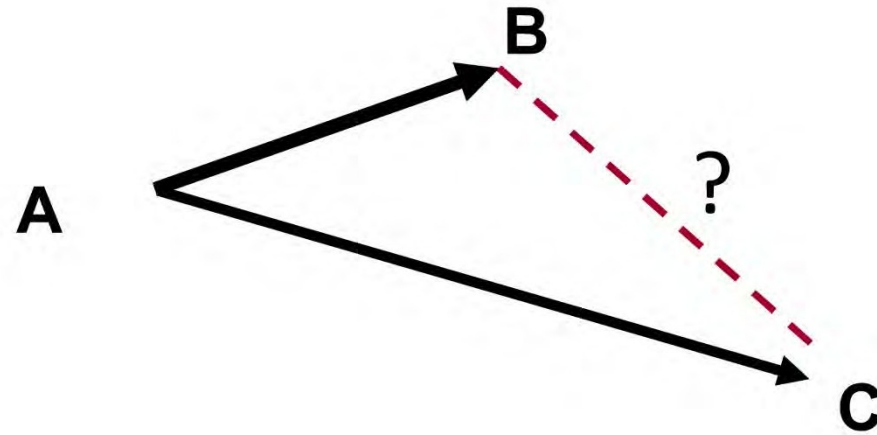
- If we know how much taller building B is to A and how much taller is C to A, we know how much taller is B compared to C



- For any pair B and C, **typical difference of C over B = difference of C over A minus difference of B over A**

# Indirect comparison

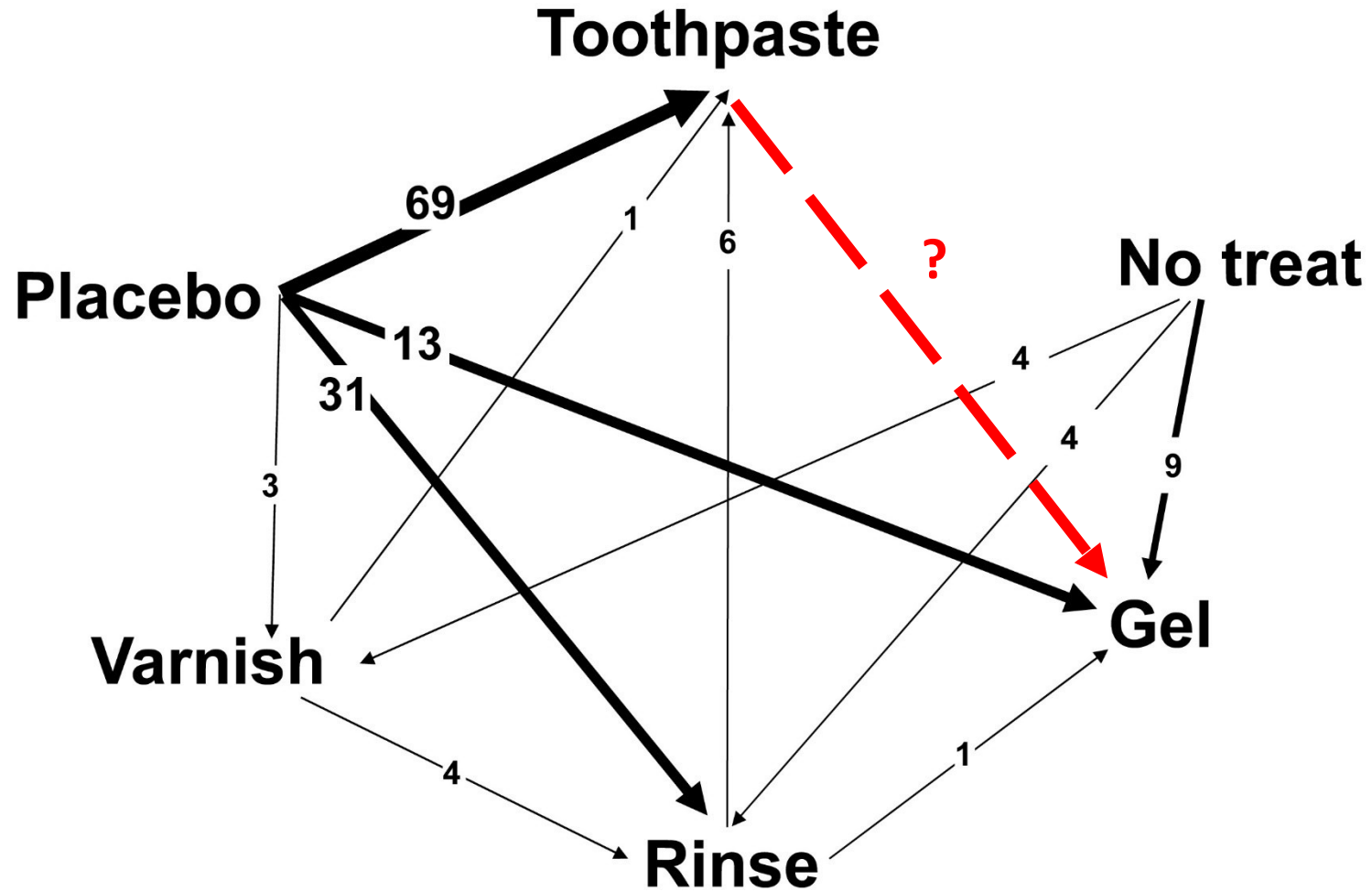
- We can obtain an **indirect estimate** of treatment effect for B vs C from trials comparing A v B and A v C



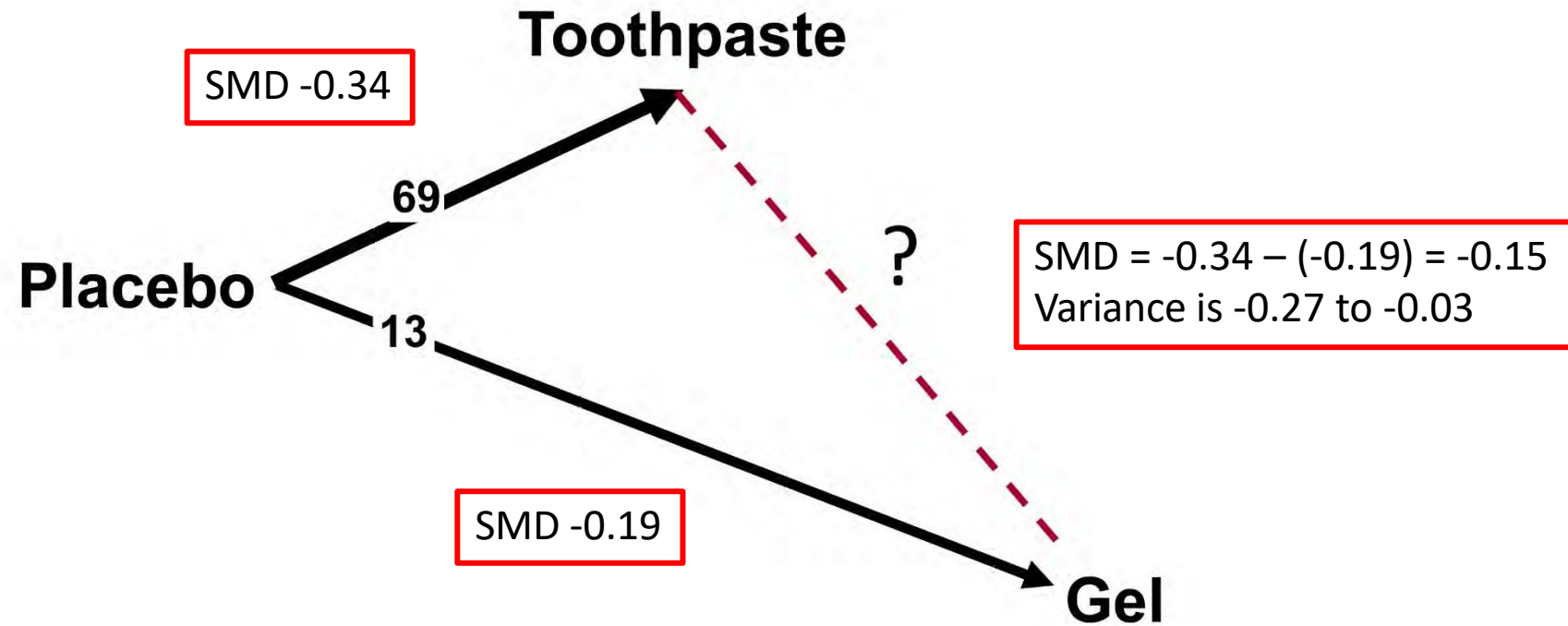
Treatment effect BC = Treatment effect AC – Treatment effect AB

Variance BC = variance AC + variance AB

# Example: Toothpaste versus gel

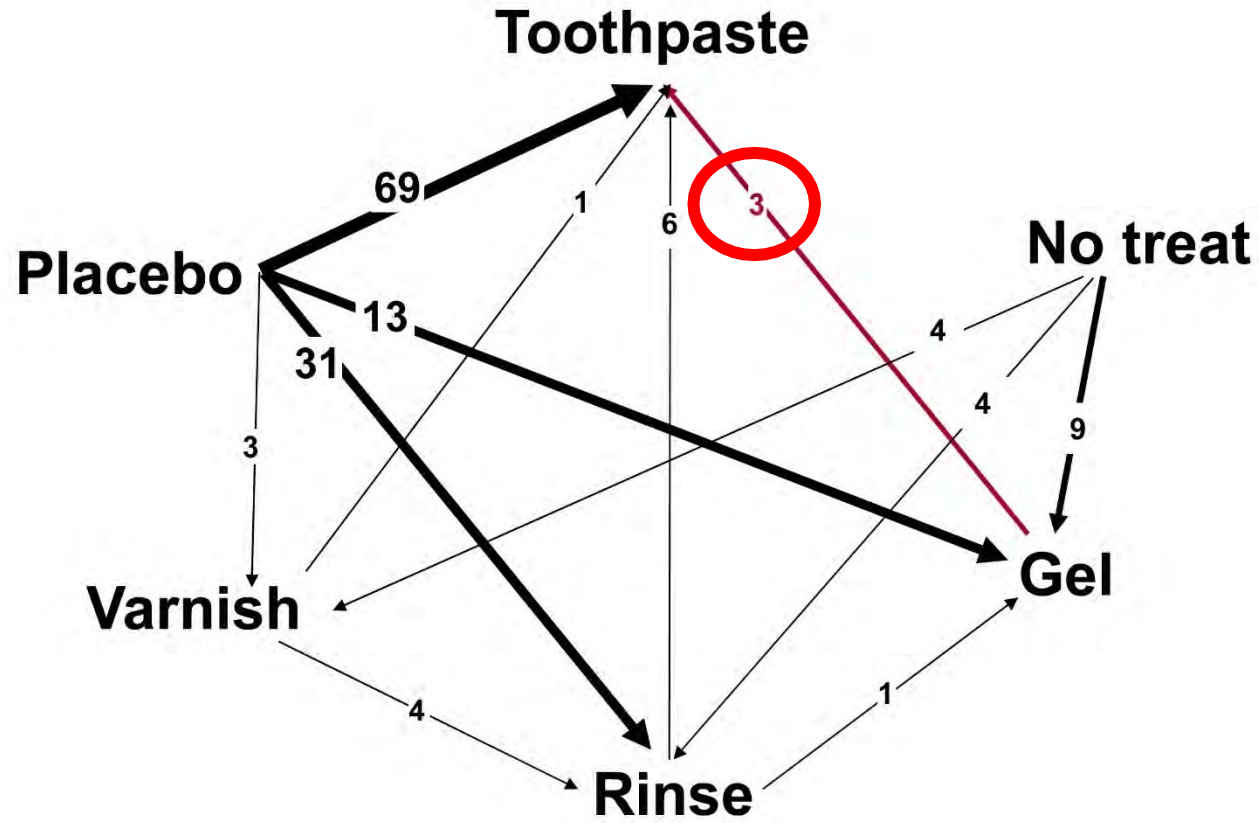


# Example: Toothpaste versus gel

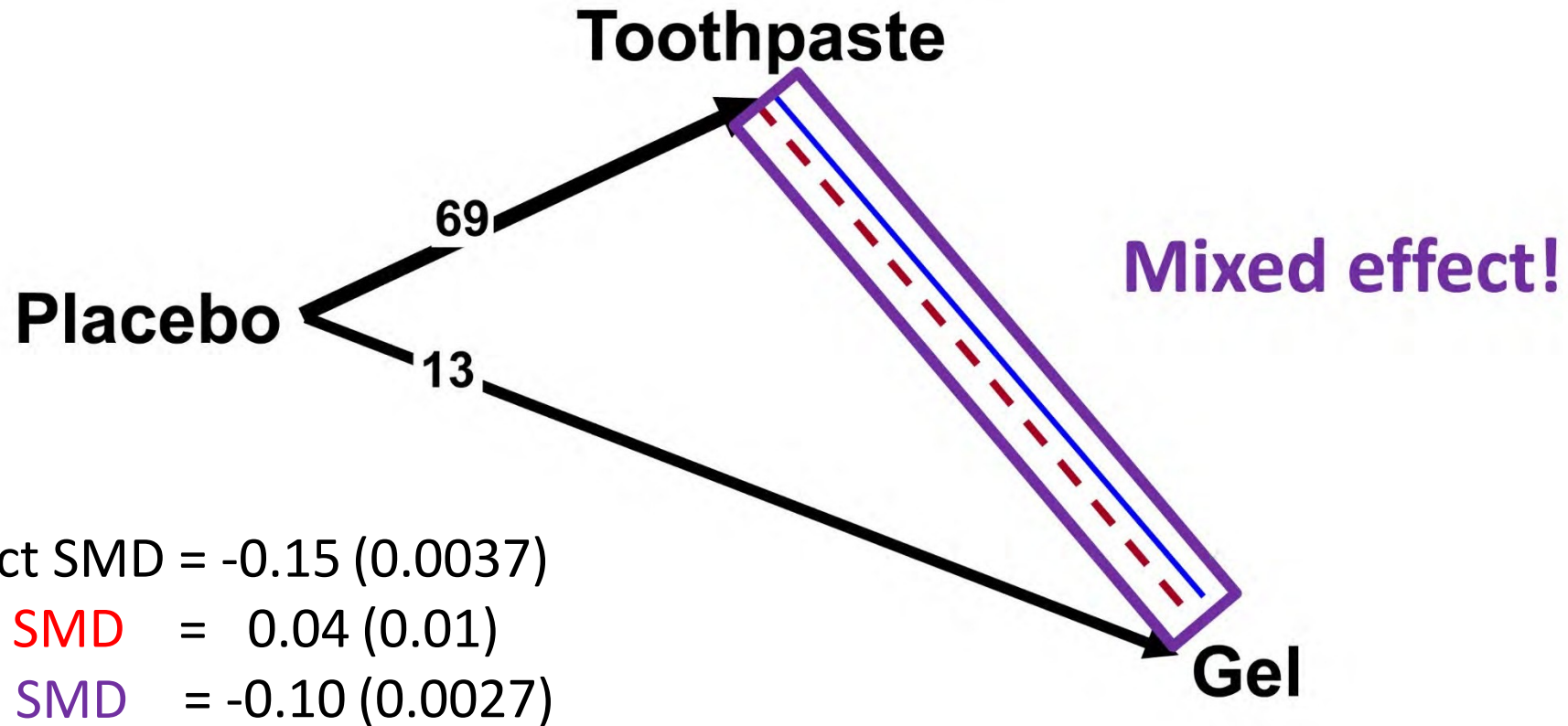


Even when there are no studies – we can estimate that toothpaste is better than gel

# Using direct and indirect effects



# Mixed effects: more precise?



# Problem of combining information



+



?



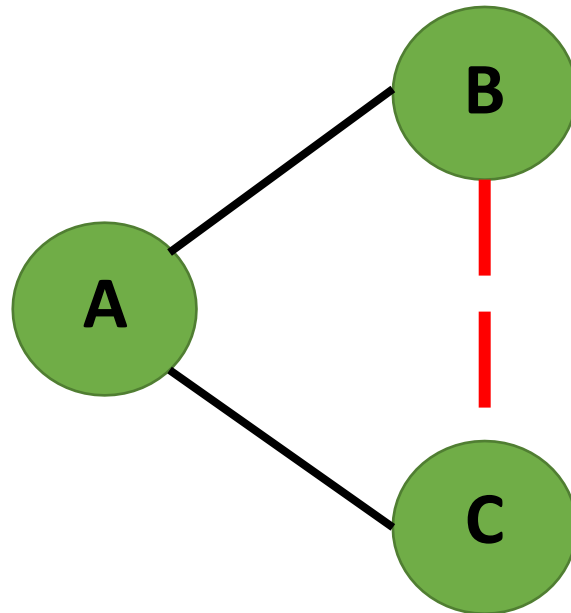


# Criticism of indirect comparisons

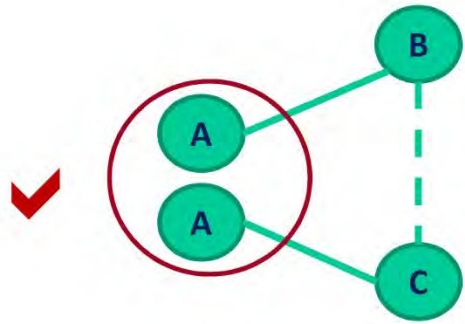
- Indirect comparison respects randomization but it is not randomized evidence
- Indirect and mixed effects (toothpaste versus gel) can answer policy questions taking a broad approach (e.g., which is the safest of all treatments)
- BUT
- They use non-randomized evidence and extra considerations are needed

# Is network meta-analysis valid?

- That one can learn about B versus C via A
- (That one can learn about toothpaste versus gel via placebo)

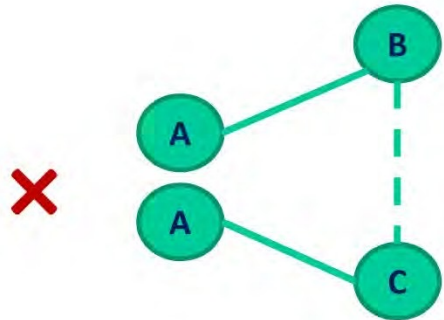


# Is the common treatment similar?

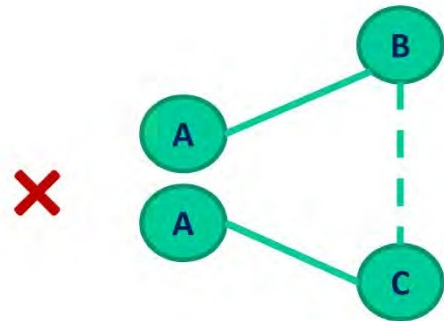
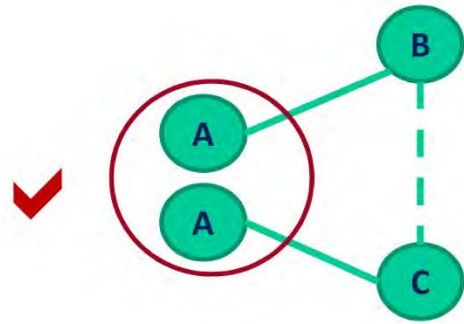


Treatment A is similar when it appears in AB and AC trials

Plausible when A is placebo given in different forms (e.g. injection versus pill )?



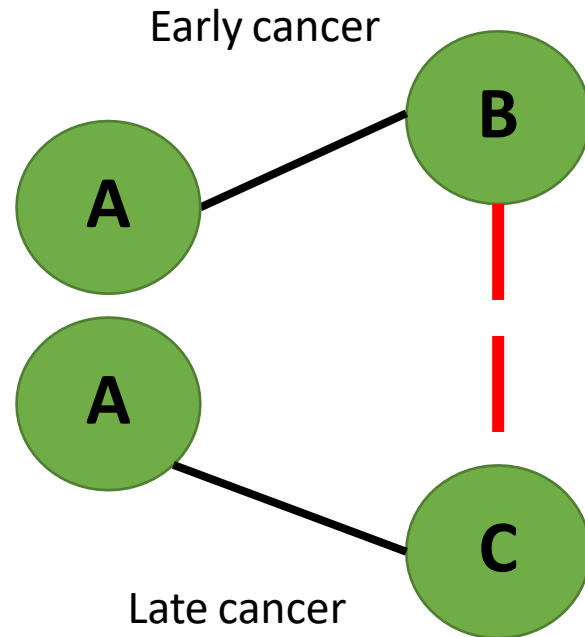
# Is the common treatment similar?



For example, placebo rinse and placebo toothpaste might not be comparable as the mechanical action of brushing might have a different effect on caries

Issue must be addressed when building the network (at the start of the project)

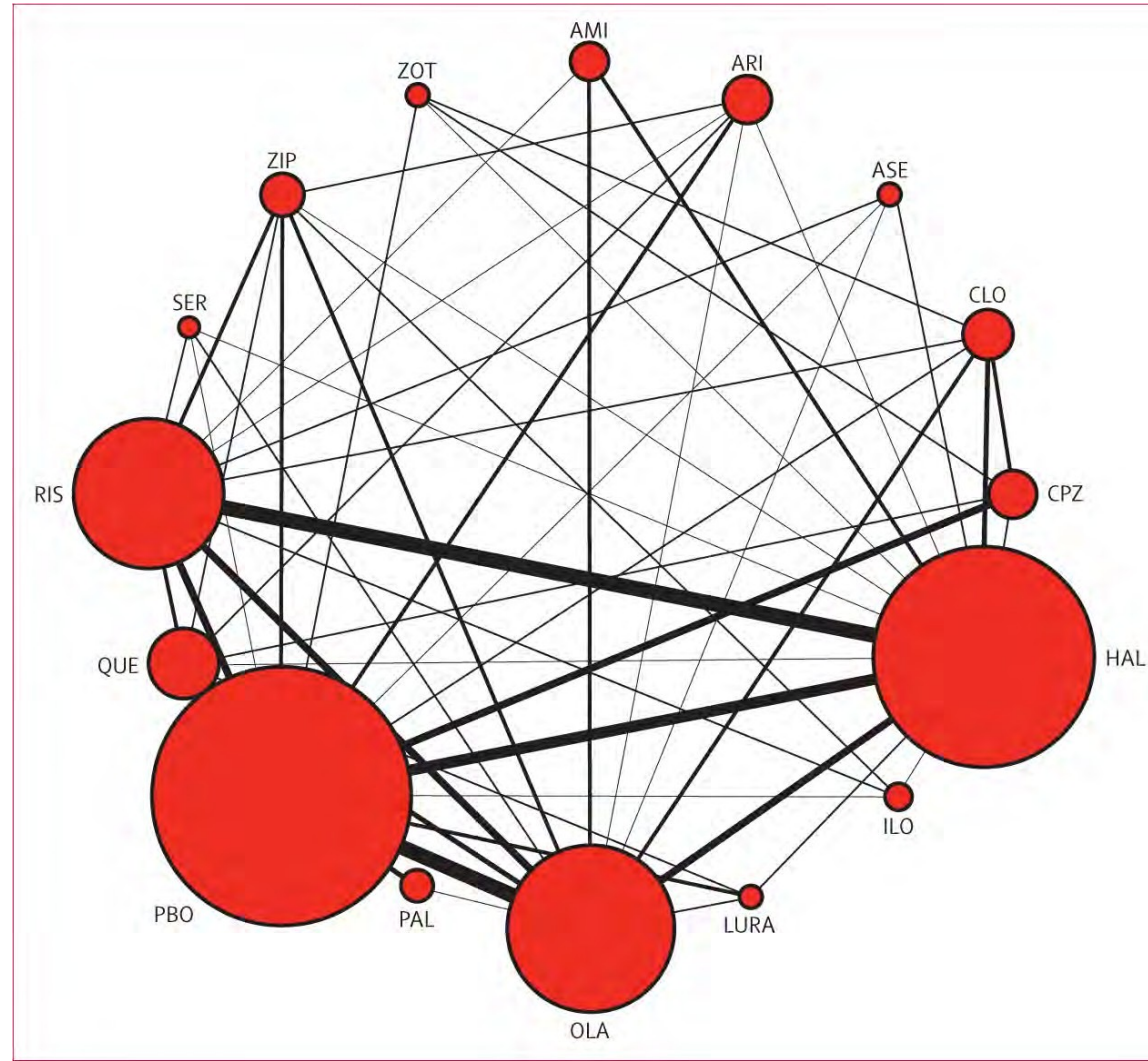
# Are the populations in the network similar?



Just like in conventional meta-analysis, there are ways to explore and understand heterogeneity and inconsistency

# **PRACTICAL APPLICATIONS OF NMA**

# 1. Show the geometry of the evidence (antipsychotics in schizophrenia)



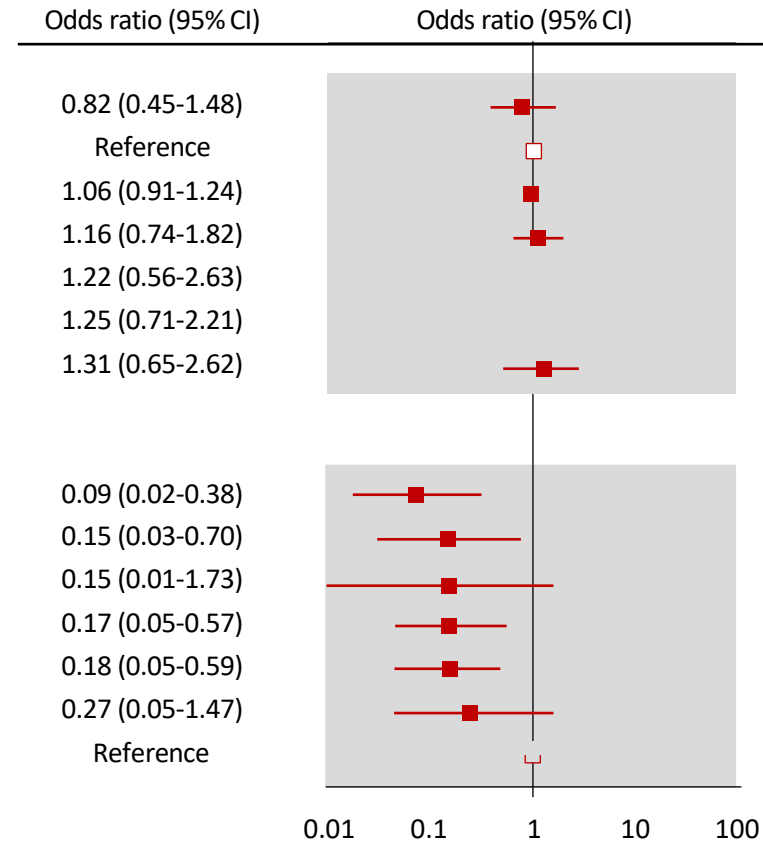
# 2. Show effects of drugs in which there are no trials

## All-cause mortality

Treatment	Odds ratio (95% CI)
Epoetin beta	0.82 (0.45-1.48)
Placebo	Reference
Darbepoetin alfa	1.06 (0.91-1.24)
Methoxy-polyethylene glycol epoetin beta	1.16 (0.74-1.82)
No treatment	1.22 (0.56-2.63)
Epoetin alfa	1.25 (0.71-2.21)
Biosimilar ESA	1.31 (0.65-2.62)

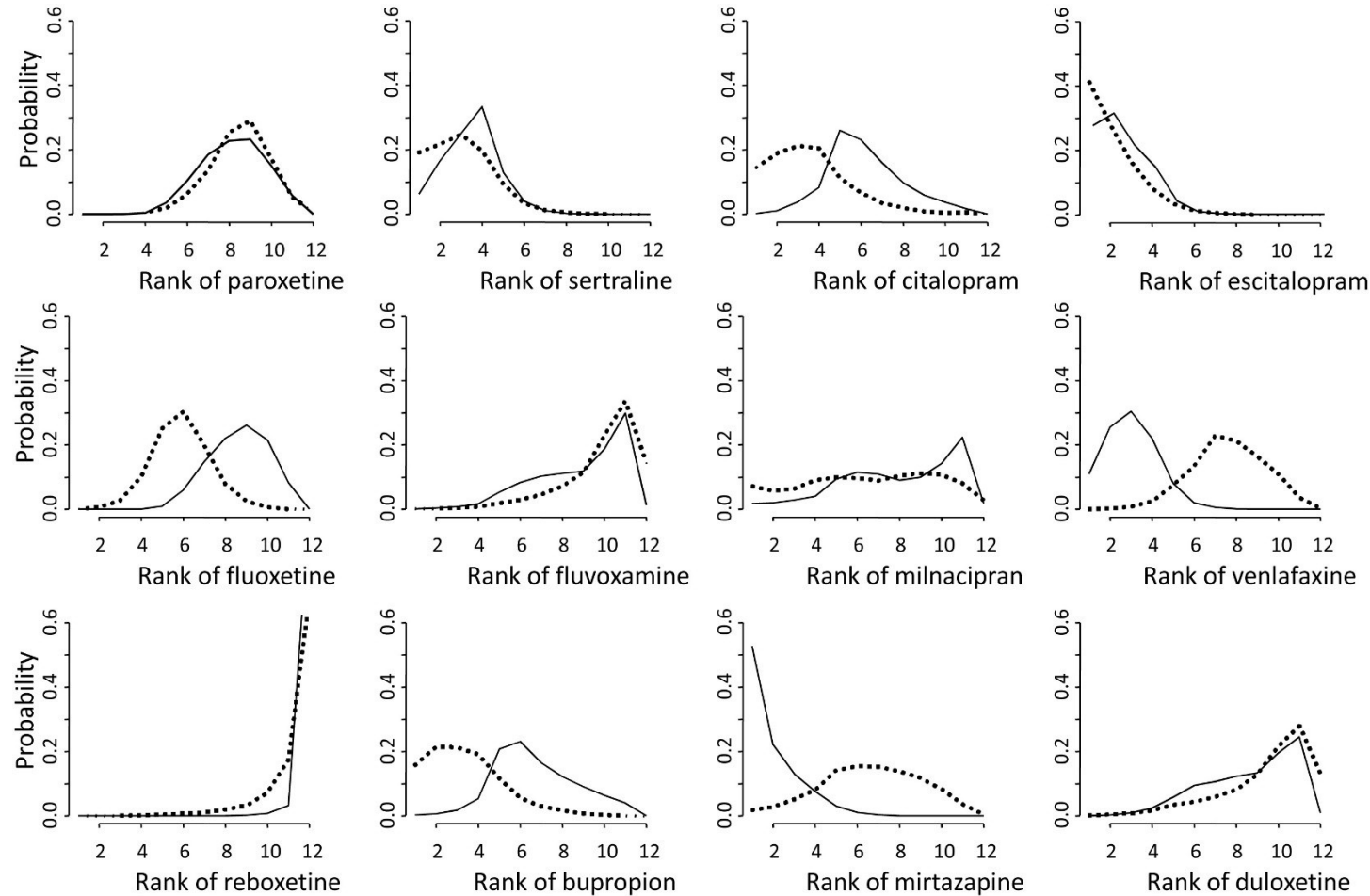
## Transfusion

Treatment	Odds ratio (95% CI)
Epoetin beta	0.09 (0.02-0.38)
Methoxy polyethylene glycol-epoetin beta	0.15 (0.03-0.70)
No treatment	0.15 (0.01-1.73)
Darbepoetin alfa	0.17 (0.05-0.57)
Epoetin alfa	0.18 (0.05-0.59)
Biosimilar ESA	0.27 (0.05-1.47)
Placebo	Reference



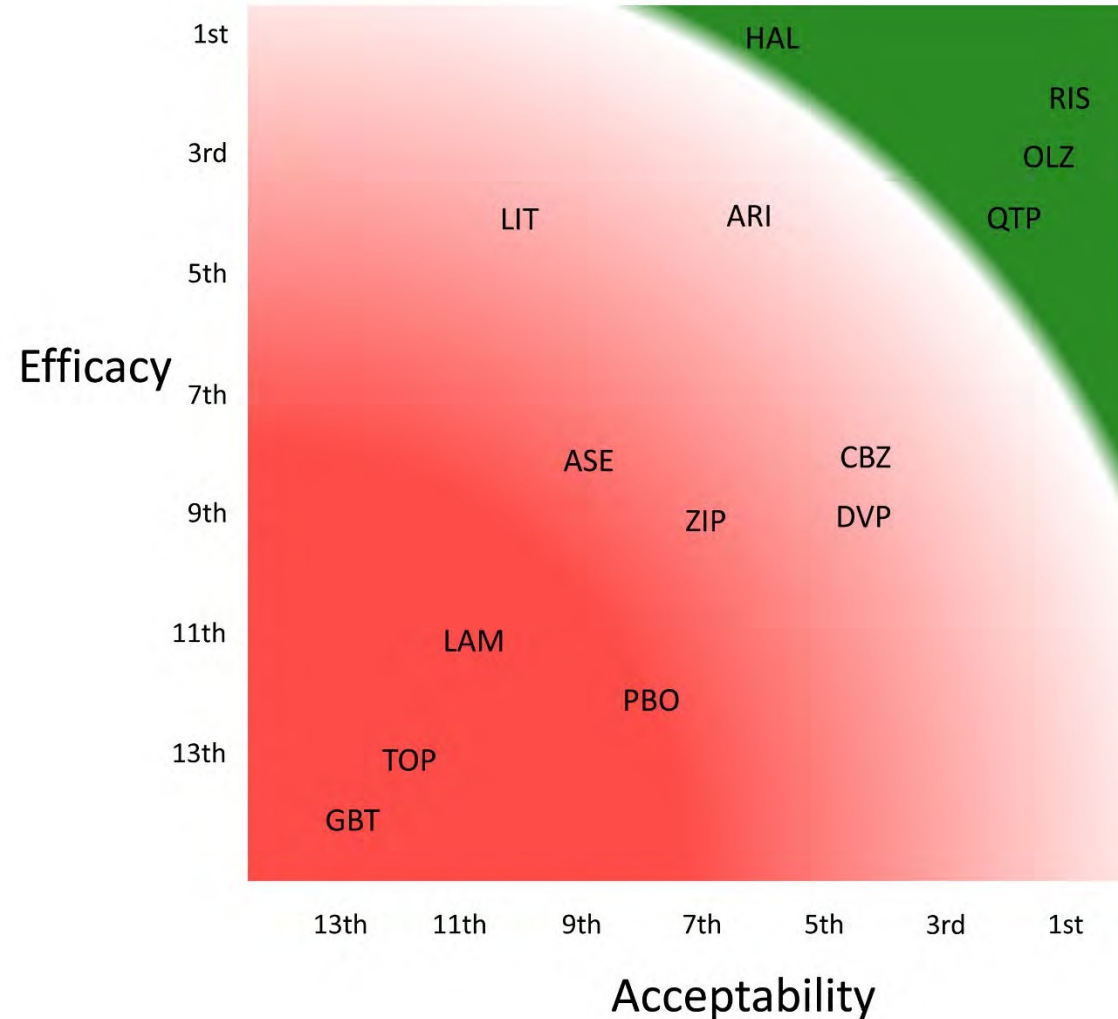


# 3. Rank treatments in order of best to worst (which antidepressant would you not want!)



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).

# 4. Display in single graphic entire relative evidence for a condition or drug



# Application in STATA

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PLOS ONE

## Graphical Tools for Network Meta-Analysis in STATA

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### Abstract

Network meta-analysis synthesizes direct and indirect evidence in a network of trials that compare multiple interventions and has the potential to rank the competing treatments according to the studied outcome. Despite its usefulness network meta-analysis is often criticized for its complexity and for being accessible only to researchers with strong statistical and computational skills. The evaluation of the underlying model assumptions, the statistical technicalities and presentation of the results in a concise and understandable way are all challenging aspects in the network meta-analysis methodology. In this paper we aim to make the methodology accessible to non-statisticians by presenting and explaining a series of graphical tools via worked examples. To this end, we provide a set of STATA routines that can be easily employed to present the evidence base, evaluate the assumptions, fit the network meta-analysis model and interpret its results.

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# Exercise versus drugs in absence of trials (or why you should exercise more!)

