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ASSESSING CONFIDENCE IN THE RESULTS OF NETWORK META-ANALYSIS (CINEMA)

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Abstract

Evaluation of the credibility of results from a meta-analysis has become an intrinsic part of the evidence synthesis process. We present a methodological framework to evaluate Confidence In the results from Network Meta-Analysis (CINeMA) when multiple interventions are compared. CINeMA considers six domains and we outline the methods used to form judgements about within-study bias, across-studies bias, indirectness, imprecision, heterogeneity and incoherence. Key to judgements about within-study bias and indirectness is the percentage contribution matrix, which shows how much information each study contributes to the results from network meta-analysis. The use of contribution matrix allows the semi-automation of the process, implemented in a freely available web application (cinema.ispm.ch). In evaluating imprecision, heterogeneity and inconsistency we consider the impact of these components of variability in forming clinical decisions. Via three examples, we show that CINeMA improves transparency and avoids the selective use of evidence when forming judgements, thus limiting subjectivity in the process. CINeMA is easy to apply even in large and complicated networks, like a network involving 18 different antidepressant drugs.

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INTRODUCTION

Network meta-analysis has become an increasingly popular tool for developing treatment guidelines and making recommendations on reimbursement. However, less than one per cent of published network meta-analyses assess the credibility of their conclusions (1). The Grading of Recommendations Assessment, Development and Evaluation (GRADE) approach requires such an assessment of the confidence in the results from systematic reviews and meta-analyses, and many organizations, including the World Health Organization (WHO), have adopted the GRADE approach (2,3). Based on GRADE, two systems have been proposed to evaluate the credibility of results from network meta-analyses (4,5). However, the complexity of the methods and lack of suitable software have limited their uptake.

In this article we introduce the methodology underpinning the CINeMA approach (Confidence In Network Meta-Analysis), and present the advances that have recently been implemented in a freely available web application (*cinema.ispm.ch*) (6). CINeMA is based on the GRADE framework, with several conceptual and semantic differences (5). It covers six confidence domains: within-study bias (referring to the impact of risk of bias in the included studies), across-studies bias (referring to publication and other reporting bias), indirectness, imprecision, heterogeneity and incoherence. CINeMA assigns judgements at three levels (no concerns, some concerns or major concerns) to each of the six domains. Judgements across the six domains are then summarized to obtain four levels of confidence for each relative treatment effect, corresponding to the usual GRADE approach: very low, low, moderate or high.

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Most network meta-analyses include only randomized controlled trials (RCTs), so we will focus on this study design, and on relative treatment effects. A network meta-analysis involves the integration of direct and indirect evidence in a network of relevant trials. We assume that evaluation of the credibility of results takes place once all primary analyses and sensitivity analyses have been undertaken. We assume that reviewers have implemented their pre-specified study inclusion criteria, which may include risk of bias considerations, and have obtained the best possible estimates of relative treatment effects using appropriate statistical methods (e.g. those described in (7–10)). The question is then how to make judgements about the credibility of relative treatment effects, given that trials with variable risk of bias, precision, relevance and heterogeneity contribute information to the estimate.

This paper addresses how judgements should be formed about the six CINeMA domains. We illustrate the methods using three examples: a network of trials that compare outcomes of various diagnostic strategies in patients with suspected acute coronary syndrome (11), a network of trials comparing the effectiveness of 18 antidepressants for major depression (12), and a network comparing adverse events of statins (13). The three examples are introduced in **Error! Reference source not found.**. All analyses were done in R software using the *netmeta* package and the CINeMA web application (Box 2) (6,14).

WITHIN-STUDY BIAS

BACKGROUND AND DEFINITIONS

Within-study bias refers to shortcomings in the design or conduct of a study that can lead to an estimated relative treatment effect that systematically differs from the truth. In

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our framework we assume that studies have been assessed for risk of bias. The majority of published systematic reviews of RCTs currently use a tool developed by Cochrane to evaluate risk of bias (15). This tool classifies studies as having low, unclear or high risk of bias for various bias components (such as allocation concealment, attrition, blinding etc.), and these judgements are then summarized across domains. A revision of the tool takes a similar approach but labels the levels as low risk of bias, some concerns and high risk of bias (16).

THE CINEMA APPROACH

While it is straightforward to gauge the impact of within-study biases on the summary relative treatment effect in a pairwise meta-analysis (17), in network meta-analysis studies contribute data to the estimation of each summary effect in a complex manner. In the first example discussed below we show the complexity underpinning the flow of information in the network of diagnostic modalities used to detect coronary artery disease. A treatment comparison directly evaluated in studies with low risk of bias might also be estimated indirectly (via a common comparator) using studies at high risk of bias, and vice versa. While studies at low risk of bias are expected to provide more credible results, it is often impractical to restrict the analysis to such studies. The treatment comparison of interest might not have been tested directly in any trial, or tested in only a few small trials with high risk of bias. Thus, even when direct evidence is present, judgements about the relative treatment effect cannot ignore the risk of bias in the studies providing indirect evidence.

If direct evidence is supplemented by indirect evidence via exactly one intermediate comparator, the risk of bias in such a one-step loop is considered along with the direct evidence. In complex networks, indirect evidence is often obtained via several routes,

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including one-step loops and loops involving several steps (see example). In general, it is not desirable to derive judgements by considering only the risk of bias in studies in a single one-step loop (4,18). This is because most studies in a network contribute *some* indirect information to every estimate of a relative treatment effect. Studies contribute more when their results are precise (e.g. large studies), when they provide direct evidence or when the indirect evidence does not involve many "steps". For example, studies in a one-step indirect comparison contribute more than studies of the same precision in a two-step indirect comparison. We can quantify the contribution made by each study to each relative treatment effect on a 0 to 100 percent scale. These quantities can be written as a 'percentage contribution matrix', as shown elsewhere (19).

CINeMA combines the studies' contributions with the risk of bias judgements to evaluate study limitation for each estimate of a relative treatment effect from a network meta-analysis. It uses the percentage contribution matrix to approximate the contribution of each study and then stratifies the percentage contribution from studies judged to be at low, moderate and high risk of bias. Using different colors, study limitations in direct comparisons can be shown graphically in the network plot, while study limitations in the estimates from a network meta-analysis are presented for each comparison in bar charts.

EXAMPLE: COMPARING DIAGNOSTIC MODALITIES TO DETECT CORONARY ARTERY DISEASE

Consider the comparison of Exercise ECG versus Standard care (Box 1). The direct evidence from a single study is at low risk of bias (3-arm study 12); so there are no study limitations when interpreting the direct odds ratio of 0.42 (Table 1). However, the odds ratio 0.52 from the network meta-analysis is estimated also by using indirect information via seven studies that compare standard care and CCTA and one study comparing exercise ECG

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and CCTA. Additionally, we have indirect evidence via stress echo. The risk of bias in these eleven studies providing indirect evidence varies. Every study in the two one-step loops contributes information proportional to its precision (the inverse of the squared standard error, largely driven by sample size). Consequently, some judgement about study limitations for the indirect evidence can be made by considering that a there is a large amount of information from studies at high risk of bias (2162 participants randomized) and low risk of bias (2788 participants) and relatively little information from studies at moderate risk of bias (362 participants). Direct evidence from the small study number 12 (130 participants) at low risk of bias is considered separately, as it has greater influence than the indirect evidence.

Calculations become more complicated because studies in the indirect comparisons contribute information not only proportional to their study precision but also to their location in the network. Indirect evidence about exercise ECG versus SPECT-MPI comes from two one-step loops (via CCTA or via Standard Care) and three two-step loops (via CCTA-Standard Care, Stress Echo-Standard Care, Standard Care-CCTA) (Figure 1A). In each loop of evidence, a different subgroup of studies contributes indirect information and their sizes and risks of bias vary. For the odds ratio from the network meta-analysis comparing exercise ECG and SPECT-MPI, study 2 with sample size 400 will be more influential than study 8 (with sample size 1392) because study 2 contributes one-step indirect evidence (via standard care).

Table 2 shows the percentage contribution matrix for the network and the columns represent the studies, grouped by comparison. The rows represent all relative treatment effects from network meta-analysis. The matrix entries show how much each study contributes to the estimation of each relative treatment effect. This information combined with the risk of bias judgements can be presented as a bar chart, as shown in Figure 2. Now,

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154 it is much easier to judge study limitations for each odds ratio; the larger the contribution
155 from studies at high or moderate risk of bias, the more concerned we are about study
156 limitations. Using this graph, we can infer that the total evidence from the network meta-
157 analysis for the comparison of exercise ECG with SPECT-MPI involves low, moderate and
158 high risk of bias studies with percentages 44%, 32% and 24%, respectively.

159 The CINeMA software offers the option to automate production of judgments, based
160 on the data presented in these bar graphs combined with specific rules. One possible rule is
161 to compute a weighted average level of risk of bias, assigning scores of −1, 0 and 1 to low,
162 moderate and high risk of bias. For the comparison exercise ECG vs SPECT-MPI, this would
163 produce a weighted score of $0.44 \times (-1) + 0.32 \times 0 + 0.24 \times 1 = -0.20$, which corresponds to
164 some concerns in the scoring scheme.

165 **EXAMPLE: COMPARING ANTIDEPRESSANTS**

166 We will focus on evaluating the results for three comparisons; amitriptyline vs
167 milnacipran (one direct study at low and one at moderate risk of bias), mirtazapine versus
168 paroxetine (three direct studies at low risk of bias and two at moderate) and amitriptyline vs
169 clomipramine (no direct studies). The odds ratios for treatment response are presented in
170 Table 3. We use this example to illustrate the use of sensitivity analysis and how it can
171 inform the amount of contribution of studies at moderate and high risk of bias that we can
172 tolerate.

173 For the first two treatment comparisons in Table 3, the contribution from studies at
174 low risk of bias is more than 50%. Moreover, the sensitivity analysis excluding studies at
175 moderate risk of bias provides results comparable to those obtained from all studies. Thus,
176 one can derive the judgment of no concerns for amitriptyline versus milnacipran and

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177 mirtazapine versus paroxetine. However, the estimation of the relative treatment effect of
178 amitriptyline versus clomipramine comes by more than 60% from studies at moderate risk
179 of bias. Given also that the odds ratio from the sensitivity analysis is quite different from to
180 the one obtained from all studies, we judge as some concerns the amitriptyline versus
181 clomipramine comparison.

182 # ACROSS-STUDIES BIAS

183 ## BACKGROUND AND DEFINITIONS

184 Across-studies bias occurs when the studies included in the systematic review are not
185 a representative sample of the studies undertaken. This phenomenon can be the result of
186 the suppression of statistically significant (or "negative") findings (publication bias), their
187 delayed publication (time-lag bias) or omission of unfavorable study results (outcome
188 reporting bias). The presence and the impact of such biases has been well documented (20–
189 26). Across-studies bias is a missing data problem, and hence it is impossible to conclude
190 with certainty for or against its presence in a given dataset. Consequently, and in agreement
191 with the GRADE system, CINeMA assumes two possible descriptions for across-studies bias:
192 suspected and undetected.

193 ## THE CINEMA APPROACH

194 Assessment of the risk of across-studies bias follows considerations on pairwise meta-
195 analysis (27). Conditions associated with 'suspected' across-studies bias include:

196 - Failure to include unpublished data and data from grey literature.

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- The meta-analysis is based on a small number of positive early findings, for example for a drug newly introduced on the market (as early evidence is likely to overestimate its efficacy and safety) (27).
- The treatment comparison is studied exclusively or primarily in industry-funded trials (28,29).
- There is previous evidence documenting the presence of reporting bias. For example the study by Turner et al. documented publication bias in placebo-controlled antidepressant trials (30).

Across-studies bias is considered 'undetected' when

- Data from unpublished studies have been identified and their findings agree with those in published studies
- There is a tradition of prospective trial registration in the field and protocols or clinical trial registries do not indicate important discrepancies with published reports
- Empirical examination of patterns of results between small and large studies, using the comparison-adjusted funnel plot (31,32), regression models (33) or selection models (34) do not indicate that results from small studies differ from those in published studies.

EXAMPLE: COMPARING ANTIDEPRESSANTS

The literature search retrieved supplementary and unpublished information from clinical trial registries, regulatory agencies' repositories and drug companies' websites (particularly for the newest and most recently marketed antidepressants). Results from published and unpublished studies did not differ materially, no asymmetry was observed in the funnel plot (12) and meta-regression did not indicate an association between study

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220 precision and study odds ratio. However, the authors decided that they cannot completely
221 rule out the possibility that some studies are missing because the field of antidepressant
222 trials has been shown to be prone to publication bias. Consequently, the review team
223 decided to assume that across-studies bias was 'suspected' for all drug comparisons.

224 # INDIRECTNESS

225 ## BACKGROUND AND DEFINITIONS

226 Systematic reviews are based on a focused research question, with a clearly defined
227 population, intervention and setting of interest. In the GRADE framework for pairwise meta-
228 analysis, indirectness refers to the relevance of the included studies to the research
229 question (35). Study populations, interventions, outcomes and study settings should match
230 the inclusion criteria of the systematic review but might not be representative of the
231 settings, populations or outcomes about which reviewers want to make inferences. For
232 example, a systematic review aiming to provide evidence about treating middle-aged adults
233 might identify studies in elderly patients; these studies will have an indirect relevance.

234 ## THE CINEMA APPROACH

235 We suggest that each study included in the network is evaluated according to its
236 relevance to the research question and classified into low, moderate or high indirectness.
237 Note that only participant, intervention and outcome characteristics that are likely
238 associated with the relative effect of an intervention against another (that is, effect
239 modifying variables) should be considered. Then, the study-level judgments can be
240 combined with the percentage contribution matrix to produce a bar plot similar to the one
241 presented in Figure 2. Evaluation of indirectness for each relative treatment effect can then

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proceed by judging whether the contribution from studies of high or moderate indirectness is important.

This approach also addresses the assumption of transitivity in network meta-analysis. Transitivity assumes that we can learn about the relative treatment effect of, say treatment A versus treatment B from an indirect comparison via C. This holds when the distributions of all effect modifiers are comparable in A versus C and B versus C studies. Differences in the distribution of effect modifiers across studies and comparisons will indicate intransitivity. Evaluation of the distribution of effect modifiers is only possible when enough studies are available per comparison. Consequently, the proposed approach will not address intransitivity in sparse networks (when there are few studies compared to the total number of treatments). Assessment of transitivity will be challenging or impossible for interventions that are poorly connected to the network. A further potential obstacle is that details of important effect modifiers might not always be reported in trial reports. For these reasons, we recommend that the network structure and the amount of available data are considered, and that judgments are on the side of caution, as highlighted in the following example.

EXAMPLE: COMPARING ANTIDEPRESSANTS

Cipriani et al concluded that there is no indirectness in any of the studies included and that the distribution of modifiers was similar across studies and comparisons (12). However, they decided to downgrade evidence about drugs that are poorly connected to the network. For example, vortioxetine was examined in a single study and consequently it was difficult to assess the comparability of effect modifiers in the comparisons with vortioxetine.

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Consequently, Cipriani et al. voiced concerns about indirectness for all comparisons with vortioxetine.

IMPRECISION

BACKGROUND AND DEFINITIONS

One of the key advantages of network meta-analysis compared to pairwise meta-analysis is the ability to gain precision (36): adding indirect evidence on a particular treatment comparison on top of direct evidence leads to narrower confidence intervals than using the direct evidence alone. However, in network meta-analysis treatment effects are also estimated with uncertainty, typically expressed as 95% confidence intervals that give an indication of where the true effect is likely to lie. To evaluate imprecision it is customary to define relative treatment effects that exclude any clinically important differences in outcomes between interventions (26). At its simplest, this treatment effect might correspond to no effect (0 on an additive scale, 1 on a ratio scale). This would mean that even a small difference is considered important, leading to one treatment being preferred over another. Alternatively, ranges may be defined that divide relative treatment effects into three categories: 'in favour of A', 'no important difference between A and B', and 'in favour of B'. The middle range is the 'range of equivalence', which includes treatment effects that correspond to clinically unimportant differences between interventions. The range of equivalence can be symmetrical (when a clinically important difference is defined, and its reciprocal constitutes the clinically important difference in the opposite direction) or asymmetrical (when clinically important differences vary by direction of effect). For simplicity, we will assume symmetrical ranges of equivalence.

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THE CINEMA APPROACH

The approach to imprecision consists of comparing the range of treatment effects included in the 95% confidence interval with the range of equivalence. If the 95% confidence interval extends to differences in treatment effects that would lead to different conclusions, for example covering two or all three of the categories defined above, then the results would be considered imprecise, reducing confidence in the treatment effect estimate. Figure 3 shows a hypothetical forest plot that illustrates the CINeMA rules to assess imprecision of network treatment effect estimates for an odds ratio of 0.8. 'Major concerns' are assigned to NMA treatment effects with 95% confidence intervals that cross both limits of the range of equivalence, 'some concerns' if only the lower or the upper limit of the range of equivalence is crossed and 'no concerns' apply to estimates that do not cross either value.

EXAMPLE: ADVERSE EVENTS OF STATINS

Consider the network comparing adverse events of different statins, introduced in Box 1 and shown in Figure 1C (37). Let us assume a range of equivalence such that an odds ratio greater than 1.05 or below 0.95 ($\frac{1}{1.05}$) would lead to favouring one the two treatments. Odds ratios between 0.95 and 1.05 would be interpreted as no important differences in the safety profile of the two statins. The 95% confidence interval of pravastatin versus rosuvastatin is quite wide, including odds ratios from 1.09 to 1.82 (Figure 4), but any treatment effect in this range would lead to the conclusion that pravastatin is safer than rosuvastatin. Thus, in this case the imprecision does not reduce the confidence that can be placed in the comparison of pravastatin with rosuvastatin ('no concerns'). The 95% confidence interval of pravastatin versus simvastatin is slightly wider (0.84 to 1.42) and, more importantly, the

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interval covers all three areas, i.e. favouring pravastatin, favouring simvastatin and no important difference. This result is very imprecise, and a rating of 'major concerns' applies. The comparison of rosuvastatin versus simvastatin is more certain, but it is again unclear which drug has fewer adverse effects: most estimates within the 95% confidence interval favour simvastatin, but the interval crosses into the range of equivalence. A rating of 'some concerns' will be appropriate here.

EXAMPLE: EFFICACY OF ANTIDEPRESSANTS

In the network of antidepressants, the authors defined clinically important effects as an odds ratio smaller than 0.8 and larger than its reciprocal 1.25 (12). We use this range of equivalence (0.8 to 1.25) in this example. We will concentrate on three comparisons, clomipramine versus fluvoxamine, citalopram versus venlafaxine and amitriptyline versus paroxetine (Table 4). The 95% confidence interval of the odds ratio comparing clomipramine with fluvoxamine (0.75 to 1.32) includes clinically important effects in both directions, implying large uncertainty in which drug should be favored ('major concerns') (Table 5). The odds ratio for citalopram versus venlafaxine is 1.12 (95% confidence interval 0.90 to 1.39), favoring venlafaxine, but the interval includes values within the range of equivalence. The verdict therefore is 'some concerns'. Finally, the odds ratio of amitriptyline versus paroxetine is 0.96 (95% confidence interval 0.82 to 1.13) in favor of amitriptyline. Despite the fact that the estimate includes 1, it is not imprecise because the 95% confidence interval is within the range of equivalence ('no concerns').

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HETEROGENEITY

BACKGROUND AND DEFINITIONS

Variability in the results of studies contributing to a particular comparison influences the confidence we have in the result for that comparison. If this variability reflects genuine differences between studies, rather than random variation, it is usually referred to as heterogeneity. The GRADE system for pairwise meta-analysis uses the term inconsistency to describe such variability (38). In network meta-analysis, there may be variation in the relative treatment effects between studies within a comparison, i.e. heterogeneity, or variation between direct and indirect sources of evidence across comparisons, i.e. incoherence (39–42) which we discuss in the next paragraph. The two notions are closely related; incoherence can be seen as a special form of heterogeneity.

There are several ways of measuring heterogeneity in a set of trials. The variance of the distribution of the underlying treatment effects (τ^2), is a useful measure of the magnitude of heterogeneity. One can estimate heterogeneity variances from each pairwise meta-analysis and, under the usual assumption of a single variance across comparisons, a common heterogeneity variance for the whole network. The magnitude of τ^2 is usefully expressed in a prediction interval, which shows where the true effect of a new study similar to the existing studies is expected to lie (28).

THE CINEMA APPROACH

Similarly to imprecision, the CINeMA approach to heterogeneity considers its influence on clinical conclusions. Large variability in the included studies does not necessarily affect conclusions, while even small amounts of heterogeneity may be important in some cases. The concordance between assessments based on confidence intervals (which

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do not capture heterogeneity) and prediction intervals (which do capture heterogeneity) can be used to assess the importance of heterogeneity. For example, a prediction interval may include values that would lead to different conclusions than suggested by the CI; in such a case, heterogeneity would be considered having important implications. The hypothetical forest plot of Figure 3 serves as an illustration of the CINeMA rules to assess heterogeneity of treatment effects for a clinically important odds ratio of below 0.8 or above 1.25.

With only a handful of trials, one cannot adequately estimate the amount of heterogeneity: prediction intervals derived from meta-analyses with very few studies can be unreliable. In this situation it may be more reasonable to interpret an estimate of heterogeneity (and its uncertainty) using empirical distributions. Turner et al. and Rhodes et al. analyzed many meta-analyses of binary and continuous outcomes, categorized them according to the outcome and type of intervention and comparison, and derived empirical distributions of heterogeneity values (16, 17). These empirical distributions can help to interpret the magnitude of heterogeneity, complementary to considerations based on prediction intervals.

EXAMPLE: ADVERSE EVENTS OF STATINS

In the statins example (Figure 1C), we assumed that the range of equivalence was 0.95 to 1.05. The prediction interval of pravastatin versus simvastatin is wide (Figure 4). However, the confidence interval for this comparison already extended into clinically important effects in both directions; thus, the implications of heterogeneity is not important and does not change the conclusion. The confidence interval for pravastatin versus rosuvastatin lies entirely above the equivalence range and is consequently considered

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sufficiently precise. However, the corresponding prediction interval crosses both boundaries (0.95 and 1.05), and we therefore would have 'major concerns' about the impact of heterogeneity. Similar considerations result in 'some concerns' regarding heterogeneity for the comparison rosuvastatin versus simvastatin.

EXAMPLE: EFFICACY OF ANTIDEPRESSANTS

In the antidepressants network, the estimated amount of heterogeneity is small ($\tau^2 = 0.03$). The prediction interval for clomipramine versus fluvoxamine does not add further uncertainty to clinical conclusions beyond that already represented by the confidence interval (Table 4), so we have 'no concerns' about heterogeneity for that comparison (Table 5). The prediction interval of citalopram versus venlafaxine extend into clinically important effects in both directions (0.74 to 1.70) while the confidence interval does not extend into values in favour of citalopram, thus suggesting potential implications of heterogeneity ('some concerns'). We have 'major concerns' about the impact of heterogeneity for the comparison amitriptyline versus paroxetine, since the confidence interval lies entirely within the range of equivalence, whereas the prediction interval includes clinically important effects in favour of both treatments (0.65, 1.42).

INCOHERENCE

BACKGROUND AND DEFINITIONS

The assumption of transitivity stipulates that we can compare two treatments indirectly via an intermediate treatment node. Incoherence is the statistical manifestation of intransitivity; if transitivity holds, the direct and indirect evidence will be in agreement (45,46). Conversely, if estimates from direct and indirect evidence disagree we conclude

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399 that transitivity does not hold. There are two approaches to quantifying incoherence. The
400 first comprises methods that examine the agreement between direct and indirect evidence
401 for specific comparisons in the network, while the second includes methods that examine
402 incoherence in the entire network. SIDE (Separate Indirect from Direct Evidence) or "node
403 splitting" (39)) is an example of the first set of methods, which are often referred to as local
404 methods. It compares direct and indirect evidence for each comparison and computes an
405 inconsistency factor with a confidence interval. The inconsistency factor is calculated as the
406 difference of the two estimates for an additive measure (e.g. log odds ratio, log risk ratio,
407 standardized mean difference) or as the ratio of the two estimates for measures on the
408 ratio scale. This method can be applied to comparisons that are informed by both direct and
409 indirect evidence. Consider for example the hypothetical example in Figure 3 (Incoherence,
410 Scenario A). The studies directly comparing the two treatments result in a direct odds ratio
411 of 1.75 (1.5 to 2) while the rest studies of the network that provide indirect evidence to the
412 particular comparison gives an indirect odds ratio of 1.37 (1.2 to 1.55). The disagreement
413 between direct and indirect odds ratios is expressed as the 'inconsistency factor' (1.27)
414 which can be used to construct a confidence interval (1.05 to 1.55) and a test statistic, here
415 resulting to a p-value of 0.07. A simpler version of SIDE splitting considers a single loop in
416 the network (loop-specific approach (47)). The second set of methods are global methods
417 that model all treatment effects and all possible inconsistency factors simultaneously,
418 resulting in an omnibus test of incoherence in the whole network. The design-by-treatment
419 interaction test is such a global method for incoherence (41,42). An overview of other
420 methods for testing incoherence can be found elsewhere (40,48).
421

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THE CINEMA APPROACH

Both global and local incoherence tests have low power (49,50) and it is therefore important to consider the inconsistency factors as well as their uncertainty. As a large inconsistency factor may be indicative of a biased direct or indirect estimate, judging its magnitude is always important. As for imprecision and heterogeneity, the CINeMA approach to incoherence considers the impact on clinical conclusions, based on visual inspection of the 95% confidence interval of direct and indirect odds ratios and the range of equivalence. Consider the hypothetical examples in Figure 3 (Incoherence). The inconsistency factor using the SIDE splitting approach is the same for the three examples (1.27 with confidence interval 1.05 to 1.55), but their position relative to the range of equivalence differs and affects the interpretation of incoherence. In the first example, the 95% confidence intervals of both direct and indirect odds ratios lie above the range of equivalence: treatment A is clearly favourable, and there are 'no concerns' regarding inconsistency. In the second example, the 95% confidence interval of the indirect odds ratio straddles the range of equivalence while for the direct odds ratio the 95% confidence interval lies entirely above 1.05. In this situation, a judgement of 'some concerns' is appropriate. In the third example, the odds ratios from direct and indirect comparisons are in opposite directions and the disagreement will therefore lead to an expression of 'major concerns'.

Note that in the three hypothetical examples above, both direct and indirect estimates exist. It could be, however, that there is only direct (e.g. venlafaxine versus vortioxetine in the network of antidepressants) or only indirect (e.g. agomelative versus vortioxetine) evidence. In this situation, we can neither estimate an inconsistency factor nor judge potential implications with respect to the range of equivalence. Considerations of indirectness and intransitivity are nevertheless important. Statistically, incoherence can only

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be judged using the global design-by-treatment interaction test. When a comparison is informed only by direct evidence, no disagreement between sources of evidence occurs and thus 'no concerns' for incoherence apply. If only indirect evidence is present then there will always be 'some concerns'. There will be 'major concerns' if the p-value of the design-by-treatment interaction test is <0.01. As in comparisons informed only by indirect evidence coherence cannot be tested, having 'no concerns' for the particular treatment effects would be difficult to defend.

EXAMPLE: COMPARING ANTIDEPRESSANTS

In the network of antidepressants, the direct odds ratio comparing clomipramine with fluvoxamine is almost double the indirect odds ratio: the ratio of the two odds ratios (i.e., the inconsistency factor) is 1.94 (95% confidence interval 0.65 to 5.73, Table 4). However, both direct and indirect estimates contain values that extend to clinically important values in both directions. Thus, incoherence will not affect the interpretation of the NMA treatment effect: there are 'no concerns' (Table 5). In contrast, there are 'major concerns' regarding the confidence in the citalopram versus venlafaxine comparison: the direct odds ratio contains values within and above the range of equivalence while the indirect odds ratio includes values within and below the range of equivalence. The resulting estimated ratio of odds ratios is 2.08 (95% confidence interval 1.03 to 4.18) and the respective p-value of the SIDE test is 0.04 (Table 4). For the comparisons of amitriptyline versus paroxetine, the ratio of direct to indirect odds ratios is 1.05 (with 95% confidence interval (0.76, 1.46) and p-value 0.75) implying that the two sources of evidence are in agreement (Table 4). Direct and indirect estimates are very close in terms of odds ratios, 95% confidence intervals and the

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468 range of equivalence and we therefore have 'no concerns' regarding incoherence for this
469 particular comparison.

470 SUMMARIZING JUDGMENTS ACROSS THE SIX DOMAINS

471 The output of the CINeMA framework is a table with the level of concern for each of
472 the six domains. Some of the domains are interconnected: factors that may reduce the
473 confidence in a treatment effect may affect more than one domain. Indirectness includes
474 considerations on intransitivity, which manifest itself in the data as statistical incoherence.
475 Heterogeneity may be related to most of the other domains. Pronounced heterogeneity will
476 increase imprecision in treatment effects and may be related to variability in within-study
477 biases or the presence of publication bias. Finally, in the presence of heterogeneity the
478 ability to detect important incoherence will decrease (49).

479 Although the final output of CINeMA is a table with the level of concern for each of
480 the six domains, reviewers may choose to summarize judgements across domains. If such an
481 overall assessment is required, one may use the four levels of confidence using the usual
482 GRADE approach: 'very low', 'low', 'moderate' or 'high' (24). For this purpose, an initial
483 strategy would be to start at 'high' confidence and to drop a rating for each domain with
484 some concerns and to drop two levels for each domain with major concerns. However, the
485 six CINeMA domains should be considered jointly rather than in isolation, avoiding
486 downgrading the overall level of confidence more than once for related concerns. For
487 example, for the 'citalopram versus venlafaxine' comparison, we have 'some concerns' for
488 imprecision and heterogeneity and 'major concerns' for incoherence (Table 3). However,
489 downgrading by two levels will be sufficient in this situation, because imprecision,
490 heterogeneity and incoherence are interconnected.

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DISCUSSION

We have outlined and illustrated the CINeMA approach for evaluating confidence in treatment effect estimates from NMA, covering the six domains of within-study bias, across-study bias, indirectness, imprecision, heterogeneity and incoherence. Our approach avoids selective use of indirect evidence, while considering the characteristics of all studies included in the network. Thus, we are not using assessments of confidence to decide whether to present direct or indirect (or combined) evidence, as has been suggested by others (4,5). We differentiate between the three sources of variability in a network, namely, imprecision, heterogeneity and incoherence and we consider the impact that each source might have on decisions for treatment. The approach can be operationalized and is easy-to-implement even for very large networks.

Any approach to evaluating confidence in evidence synthesis results will inevitably involve some subjectivity. Our approach is no exception. While the use of bar charts to infer about the impact of within study biases and indirectness provides a consistent assessment across all comparisons in the network, their summary is difficult. Setting up a margin of equivalence might be equivocal. Further limitations of the framework are associated with the fact that published articles are used to make judgements and these reports do not necessarily reflect the way studies were undertaken. For instance, judging indirectness requires study data to be collected on pre-specified effect modifiers; reporting limitations will inevitably impact on the reliability of the judgements.

A consequence of the inherent subjectivity of the system is that interrater agreement may be modest. Studies of the reproducibility of assessments made by researchers using CINeMA will be required in this context. We believe however that

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514 transparency is key. Although judgements may differ across reviewers, they are made using
515 explicit criteria. These should be specified in the review protocol so that data-driven
516 decisions are avoided. The web application at *cinema.ispm.ch* will greatly facilitate the
517 implementation of all steps involved in the application of CINeMA (6).

518 This paper proposes a refinement of a previously suggested framework (51). An
519 alternative approach has also been refined (52) since its initial introduction (53). The two
520 methods have similarities but also notable differences. For example, Puhan et al (53)
521 suggest a process of deciding whether indirect estimates are of sufficient certainty to
522 combine them with the direct estimates. In contrast CINeMA evaluates relative treatment
523 effects without considering separately the direct and indirect sources. Evaluation of the
524 impact of within-study bias and indirectness differs materially between the two approaches.
525 The need to choose the most influential one-step loop in the GRADE approach as described
526 by Puhan et al. (53) and Brignardello-Petersen (18) discards a large amount of information
527 that contributes to the results and makes the approach difficult to apply to large networks.
528 The percentage contribution matrix appears to be the only viable option to acknowledge
529 the impact of each and every study included in a network. Moreover, our framework
530 naturally includes the results from sensitivity analyses in the interpretation of the bar
531 charts. Finally, in contrast to GRADE approach, we do not rely on metrics for judging
532 heterogeneity and incoherence: we consider instead the impact that these can have when a
533 stakeholder needs to make informed decisions. An alternative approach to assessing
534 confidence findings from network meta-analysis is to explore how robust treatment
535 recommendations are to potential degrees of bias in the evidence (54). The method is easy
536 to apply but focuses on the impact of bias and does not explicitly address heterogeneity,
537 indirectness and inconsistency.

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538 Evidence synthesis is increasingly used by national and international agencies (55,56)
539 to inform decisions about the reimbursement of medical interventions, by clinical guideline
540 panels to recommend one drug over another and by clinicians to prescribe a treatment or
541 recommend a diagnostic procedure for individual patients. However, it is the exception
542 rather than the rule for published network meta-analyses to formally evaluate confidence in
543 relative treatment effects (57). With the use of open-source free software (see Box 2), our
544 approach can be routinely applied to any network meta-analysis (6) and offers a step
545 forward in transparency and reproducibility. The suggested framework operationalizes,
546 simplifies and accelerates the process of evaluation of results from large and complex
547 networks without compromising in statistical and methodological rigor. The CINeMA
548 framework is a transparent, rigorous and comprehensive system for evaluating the
549 confidence of treatment effect estimates from network meta-analysis.
550

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Box 1. Description of three network meta-analyses used to illustrate the CINeMA approach to assess confidence in network meta-analysis.

Diagnostic strategies for patients with low risk of acute coronary syndrome

Siontis et al performed a network meta-analysis to of randomized trials to evaluate the differences between the non-invasive diagnostic modalities used to detect coronary artery disease in patients presenting with symptoms suggestive of acute coronary syndrome (11). Differences between the diagnostic modalities were evaluated with respect to the number of downstream referrals for invasive coronary angiography and other clinical outcomes. For outcome referrals, 18 studies were included. The network is presented in Figure 1A and the data in Table S1. The results from the network meta-analysis are presented in Table 1.

Antidepressants for moderate and major depression

Cipriani et al compared 18 commonly prescribed antidepressants, which were studied in 179 head-to-head randomized trials involving patients diagnosed with major/moderate depression (12). The primary efficacy outcome was response measured as 50% reduction in the symptoms scale between baseline and 8 weeks of follow-up. According to the inclusion criteria specified in the protocol only studies at low or moderate risk of bias were included (58). The methodological and statistical details presented in the published article and its appendix. Here, we will focus on how judgements about credibility of the results were derived. The network is presented in Figure 1B and the data is available in Mendeley Data (DOI:10.17632/83rthbp8ys.2).

Comparative tolerability and harms of statins

The aim of the systematic review by Naci et al. (37) was to determine the comparative tolerability and harms of eight statins. The outcome considered here is the number of patients who discontinued therapy due to adverse effects, measured as an odds ratio. This

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575 outcome was evaluated in 101 studies. The network is presented in Figure 1C and the
576 outcome data are given in Table S4. The results of the network meta-analysis are presented
577 in Table S5 and the results from SIDE splitting in Table S6.

578

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Box 2. Description of the CINeMA web-application.

THE CINeMA WEB APPLICATION

CINeMA framework has been implemented in a freely available, user-friendly web-application aiming to facilitate the evaluation of confidence on the results from network meta-analysis (http://cinema.ispm.ch/ (6)). The web application is programmed in javascript, uses docker and is linked with R; in particular, packages *meta* and *netmeta* are used (59). Knowledge of the aforementioned languages and technologies is however not required from the users.

Loading the data

In 'My Projects' tab, CINeMA users are able to upload a .csv file with the by-treatment outcome study data and study-level risk of bias (RoB) and indirectness judgments. CINeMA web-application can handle all the formats used in network meta-analysis (long or wide format, binary or continuous, arm level or study level data) and provides flexibility in labelling variables as desired by the user. A demo dataset is available in 'My Projects' tab.

Evaluating the confidence in the results from network meta-analysis

A preview of the evidence (network plot and outcome data) and options concerning the analysis (fixed or random effects, effect measure etc.) are given in the 'Configuration' tab. The next six tabs guide users to make informed conclusions on the quality of evidence based on within-study bias, across-studies bias, indirectness, imprecision, heterogeneity and incoherence. Features implemented include the percentage contribution matrix, relative

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599 treatment effects for each comparison, estimation of the heterogeneity variance, prediction
600 intervals and tests for the evaluation of the assumption of coherence.

601 **Summarising judgments**

602 The last tab 'Report' includes a summary of the evaluations made in the six domains and
603 gives users the possibility to either not downgrade, or downgrade by one or two levels each
604 relative treatment effect. Users can download a report with the summary of their
605 evaluations along with their final judgements. CINeMA is accompanied by a documentation
606 describing each step in detail (tab 'Documentation').

607

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Table 1. Results from pairwise (upper triangle) and network meta-analysis (lower triangle) from the network of non-invasive diagnostic strategies for the detection of coronary artery disease in Figure 1A. Odds ratios and their 95% confidence intervals are presented for referrals for invasive coronary angiography. Odds ratios in the lower triangle less than one favor the strategy in the column; odds ratios in the upper triangle less than one favor the strategy in the row. Cells with a dot indicate that no direct studies examine the particular comparison.

CCTA	.	2.25 [1.04 - 4.90]	1.04 [0.70 - 1.55]	1.23 [1.00 - 1.50]	.
3.07 [1.46 - 6.45]	**CMR**	.	.	0.38 [0.18 - 0.78]	.
2.24 [1.22 - 4.11]	0.73 [0.28 - 1.88]	**Exercise ECG**	.	0.42 [0.14 - 1.30]	1.93 [1.39 - 2.67]
1.27 [1.01 - 1.60]	0.42 [0.20 - 0.87]	0.57 [0.30 - 1.07]	**SPECT-MPI**	0.87 [0.71 - 1.06]	.
1.17 [0.97 - 1.40]	0.38 [0.18 - 0.78]	0.52 [0.28 - 0.96]	0.92 [0.76 - 1.10]	**Standard Care**	2.95 [0.97 - 8.98]
4.31 [2.23 - 8.32]	1.40 [0.53 - 3.74]	1.93 [1.39 - 2.66]	3.38 [1.71 - 6.68]	3.69 [1.90 - 7.17]	**Stress Echo**

ECG: electrocardiogram; echo: echocardiography; SPECT-MPI: single photon emission computed tomography-myocardial perfusion imaging; CCTA: coronary computed tomographic angiography; CMR: cardiovascular magnetic resonance.

Table 2. The percentage contribution matrix for the network presented in Figure 1A. The columns refer to the studies (grouped by comparison) and the rows refer to the relative treatment effects (grouped into mixed and indirect evidence) from network meta-analysis. The entries show how much each study contributes (as percentage) to the estimation of relative treatment effects.

Direct comparisons (number of studies)	**CCTA vs Exercise ECG (1)**	**CCTA vs SPECT-MPI (2)**		**CCTA vs Standard care (7)**							**CMR vs Standard care (2)**		**Exercise ECG vs Standard care (1)**	**Exercise ECG vs Stress Echo (4)**				**SPECT-MPI vs Standard care (2)**		**Standard care vs Stress Echo (1)**
NMA Estimates/study IDs	**3**	**2**	**9**	**1**	**10**	**13**	**14**	**4**	**7**	**8**	**11**	**6**	**12**	**12**	**15**	**16**	**17**	**18**	**5**	**12**
Mixed estimates																				
CCTA:Exercise ECG	52	1	1	3	0	3	1	3	4	4	0	0	14	0	3	0	2	1	1	6
CCTA:SPECT-MPI	1	18	16	5	1	5	1	6	7	7	0	0	0	0	0	0	0	22	10	0
CCTA:Standard care	1	4	4	13	2	13	3	15	18	17	0	0	1	0	0	0	0	6	3	0
CMR:Standard care	0	0	0	0	0	0	0	0	0	0	60	40	0	0	0	0	0	0	0	0
Exercise ECG:Standard care	23	1	1	3	0	3	1	4	5	4	0	0	30	1	6	1	3	2	1	11
Exercise ECG:Stress Echo	1	0	0	0	0	0	0	0	0	0	0	0	1	5	52	8	29	0	0	2
SPECT-MPI:Standard care	0	5	4	1	0	1	0	2	2	2	0	0	0	0	0	0	0	57	26	0
Standard care:Stress Echo	14	1	1	2	0	2	1	2	3	3	0	0	14	2	16	2	9	1	1	27
Indirect estimates	--	--	--	--	--	--	--	--	--	--	--	--	0	0	--	--	--	--	--	0
CCTA:CMR	1	3	2	6	1	7	2	8	9	8	28	19	1	0	0	0	0	4	2	0
CCTA:Stress Echo	24	1	1	3	0	3	1	3	4	4	0	0	8	2	18	3	10	1	1	13
CMR:Exercise ECG	16	1	1	2	0	2	1	3	3	3	22	15	15	0	4	1	2	1	1	7
CMR:SPECT-MPI	0	3	3	1	0	1	0	1	1	1	28	19	0	0	0	0	0	28	13	0
CMR:Stress Echo	11	1	1	1	0	2	0	2	2	2	20	14	9	1	11	2	6	1	0	13
Exercise ECG:SPECT-MPI	21	7	6	1	0	1	0	1	2	2	0	0	15	0	4	1	2	20	9	7
SPECT-MPI:Stress Echo	14	5	4	1	0	1	0	1	1	1	0	0	9	1	13	2	7	19	9	13

ECG: electrocardiogram; echo: echocardiography; SPECT-MPI: single photon emission computed tomography-myocardial perfusion imaging; CCTA: coronary computed tomographic angiography; CMR: cardiovascular magnetic resonance.

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Table 3. Summary odds ratios from network meta-analysis comparing six antidepressants and sensitivity analyses excluding studies at moderate risk of bias.

Comparison	**Response odds ratio [95% confidence interval]**	
	All studies *(179 studies)*	***Studies at low risk of bias*** *(83 studies)*
Amitriptyline versus Milnacipran	1.11 [0.85; 1.43]	1.10 [0.77; 1.59]
Mirtazapine versus Paroxetine	1.07 [0.88; 1.30]	1.08 [0.83; 1.39]
Amitriptyline versus Clomipramine	1.24 [0.97; 1.59]	0.96 [0.59; 1.57]

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Table 4. Results from direct, indirect and mixed evidence along with confidence and prediction intervals and incoherence ratio of odds ratios for the network of antidepressants. Odds ratios lower than 1 favour the first treatment.

Comparison	**Direct OR (95% CI)**	**Indirect OR (95% CI)**	**Ratio of ORs (95% CI)**	**NMA OR (95% CI)**	**95% PrI of NMA OR**
Clomipramine versus Fluvoxamine	1.85 (0.65 to 5.27)	0.96 (0.71 to 1.29)	1.94 (0.65 to 5.73)	0.99 (0.75 to 1.32)	(0.63 to 1.57)
Citalopram versus Venlafaxine	1.72 (0.89 to 3.32)	0.83 (0.66 to 1.04)	2.08 (1.03 to 4.18)	1.12 (0.90 to 1.39)	(0.74 to 1.70)
Amitriptyline versus Paroxetine	1.07 (0.85 to 1.36)	1.02 (0.82 to 1.27)	1.05 (0.76 to 1.46)	0.96 (0.82 to 1.13)	(0.65 to 1.42)

NMA: network meta-analysis, OR: odds ratio, PrI: prediction interval, CI: confidence interval.

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Table 5. Level of concern for three network meta-analysis odds ratios from the network of antidepressants for the domains imprecision, heterogeneity and incoherence. See Table 4 for odds ratios.

Comparison	**Imprecision**	**Heterogeneity**	**Incoherence**
Clomipramine versus Fluvoxamine	Major concerns	No concerns	No concerns
Citalopram versus Venlafaxine	Some concerns	Some concerns	Major concerns
Amitriptyline versus Paroxetine	No concerns	Major concerns	No concerns

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Figure 1. Network plots of the three network meta-analyses used as examples. The width of the edges are proportional to the number of patients randomised in each comparison. A: Network of randomised controlled trials comparing non-invasive diagnostic strategies for the detection of coronary artery disease in patients with low risk acute coronary syndrome. The colours of edges and nodes refer to the risk of bias; low (green), moderate (yellow) and red (high). In square brackets are the study IDs as presented in Table S1. B: Network of randomised controlled trials comparing active antidepressants in patients with moderate/major depression. The colours of edges refer to the risk of bias; low (green), moderate (yellow) and red (high). The size of nodes is proportional to the number of studies examining each treatment. C: Network of randomised controlled trials comparing statins with respect to adverse effects.

ECG: electrocardiogram; echo: echocardiography; SPECT-MPI: single photon emission computed tomography-myocardial perfusion imaging; CCTA: coronary computed tomographic angiography; CMR: cardiovascular magnetic resonance.

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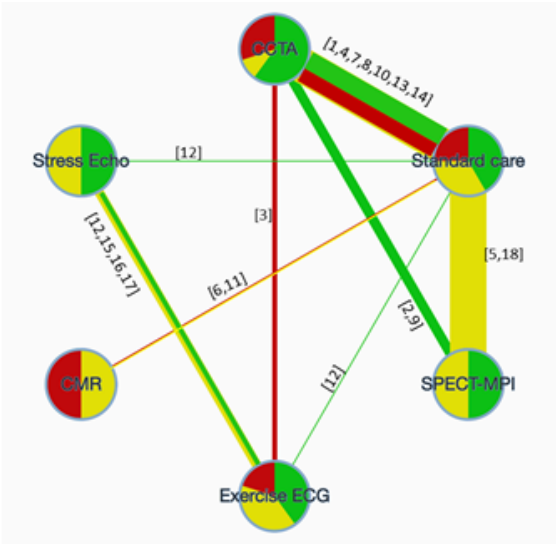
A



B

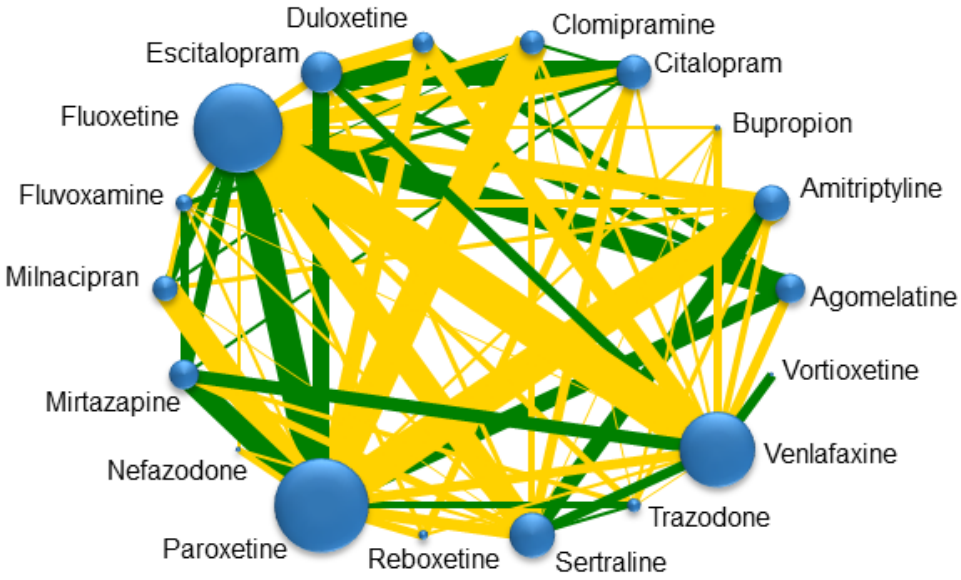


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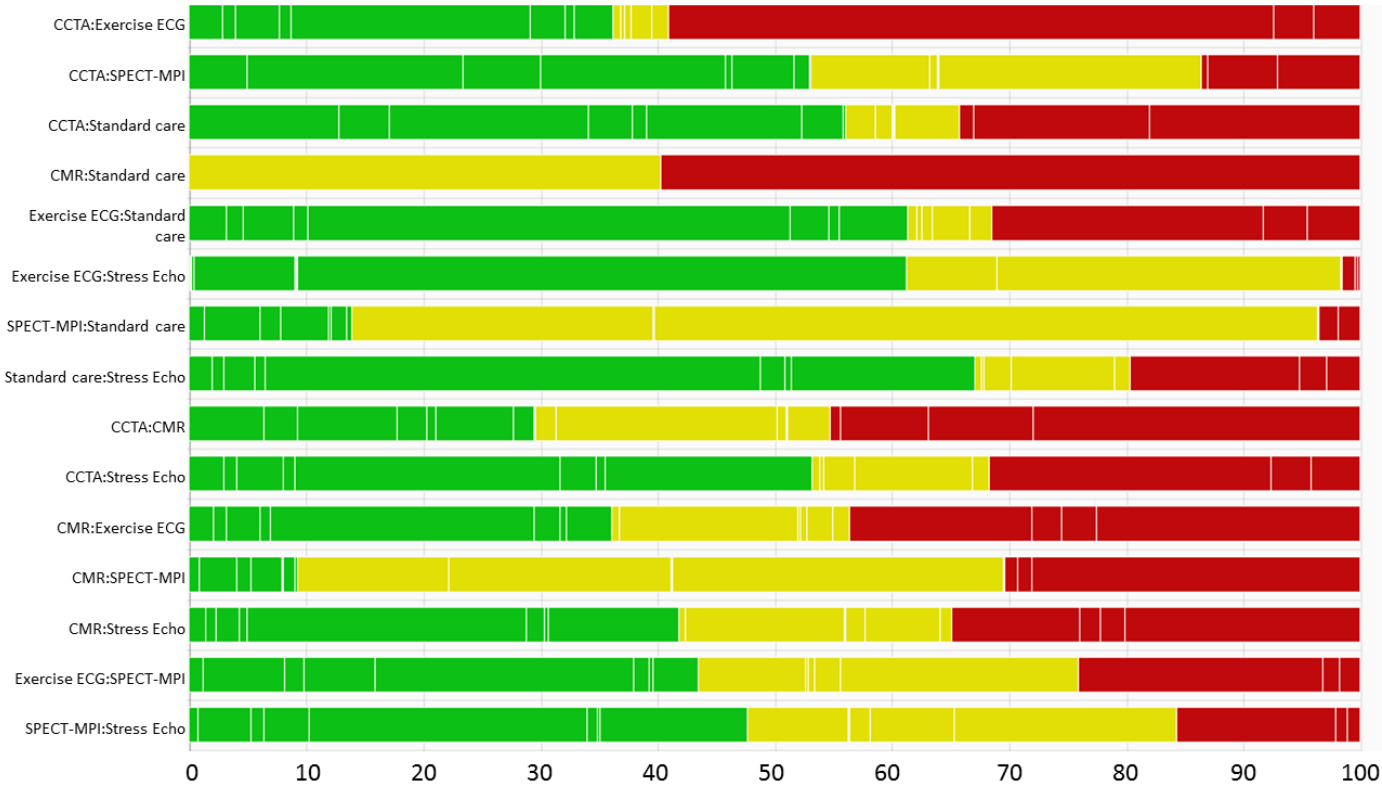
Figure 2. Risk of bias bar chart for the comparison of non-invasive diagnostic strategies for the detection of coronary artery disease. Each bar represents a relative treatment effect estimated using the data in the network in *Error! Reference source not found.****A. White vertical lines indicate the percentage contribution of separate studies. Each bar shows the percentage contribution from studies judged to be at low (green), moderate (yellow) and high (red) risk of bias.***



ECG: electrocardiogram; echo: echocardiography; SPECT-MPI: single photon emission computed tomography-myocardial perfusion imaging; CCTA: coronary computed tomographic angiography; CMR: cardiovascular magnetic resonance.

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Figure 3. CINeMA rules to assess imprecision, heterogeneity and incoherence of network treatment effects. The range of equivalence is from 0.8 to 1.25. Black lines indicate confidence intervals and red lines indicate prediction intervals. For the three scenarios presented for incoherence, inconsistency factor is 1.27 (1.05 to 1.55).

OR: odds ratio.

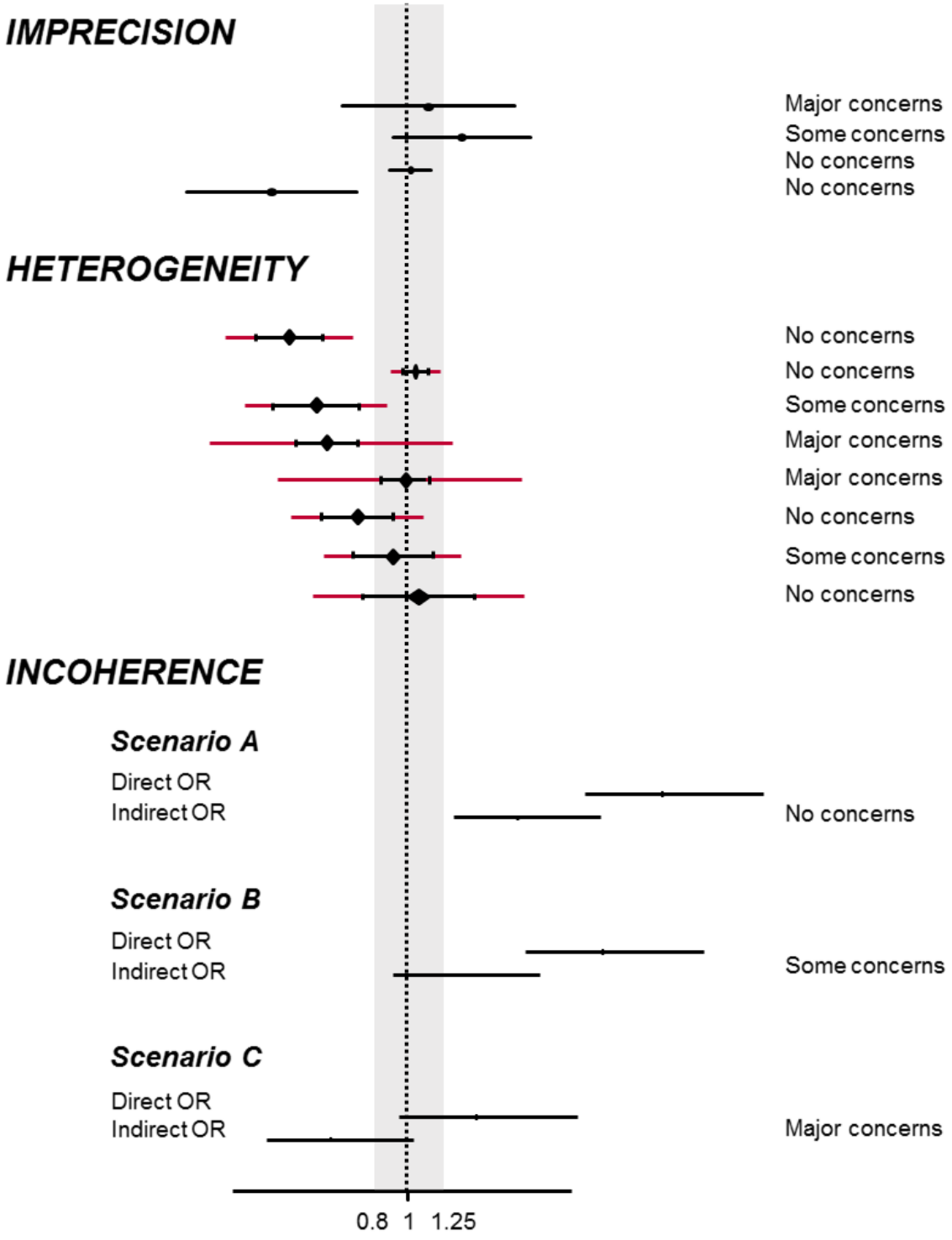

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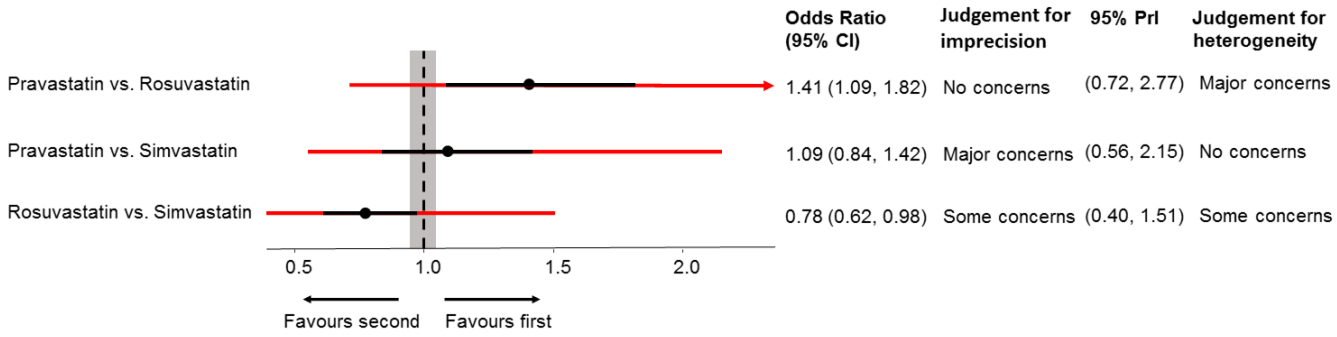
Figure 4. Network meta-analysis odds ratios from the network of statins their 95% confidence intervals (black lines) and their 95% prediction intervals (red lines). The range of equivalence is from 0.95 to 1.05.

PrI: 95% prediction interval, CI: 95% confidence interval, vs: versus.



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SUPPLEMENTARY MATERIAL

Table S1 Data from Network of randomised controlled trials comparing non-invasive diagnostic strategies for the detection of coronary artery disease in patients with low risk acute coronary syndrome. The data was originally published by Siontis et al.

id	trial	group	n	r	rob	t
1	BEACONR1	Anatomical testing	250	41	1	CCTA
1	BEACONR1	Standard care	250	31	1	Standard care
2	Levsky JM., et al.R2	Anatomical testing	200	30	1	CCTA
2	Levsky JM., et al.R2	Functional testing	200	32	1	SPECT-MPI
3	CT-COMPARER3	Anatomical testing	322	26	3	CCTA
3	CT-COMPARER3	Functional testing	240	9	3	Exercise ECG
4	CATCHR4,R5	Anatomical testing	299	49	3	CCTA
4	CATCHR4,R5	Standard care	301	36	3	Standard care
5	Lim SH., et al.R6	Functional testing	1126	73	2	SPECT-MPI
5	Lim SH., et al.R6	Standard care	564	56	2	Standard care
6	Miller CD., et al.R7	CMR	52	5	2	CMR
6	Miller CD., et al.R7	Standard care	53	11	2	Standard care
7	ROMICAT-IIR8	Anatomical testing	501	59	3	CCTA
7	ROMICAT-IIR8	Standard care	499	40	3	Standard care
8	ACRIN-PAR9,R10	Anatomical testing	929	69	1	CCTA
8	ACRIN-PAR9,R10	Standard care	463	32	1	Standard care
9	CT-STATR11	Anatomical testing	375	26	1	CCTA
9	CT-STATR11	Functional testing	374	22	1	SPECT-MPI
10	Miller AH., et al.R12	Anatomical testing	30	4	2	CCTA
10	Miller AH., et al.R12	Standard care	30	4	2	Standard care
11	Miller CD., et al.R13,R14	CMR	52	8	3	CMR
11	Miller CD., et al.R13,R14	Standard care	57	19	3	Standard care

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12	Nucifora G., et al.R15	Functional testing	77	5	1	Stress Echo
12	Nucifora G., et al.R15	Functional testing	75	9	1	Exercise ECG
12	Nucifora G., et al.R15	Standard care	55	8	1	Standard care
13	Chang SA., et al.R16	Anatomical testing	133	47	1	CCTA
13	Chang SA., et al.R16	Standard care	133	57	1	Standard care
14	Goldstein JA., et al.R17	Anatomical testing	99	12	1	CCTA
14	Goldstein JA., et al.R17	Standard care	98	7	1	Standard care
15	Jeetley P., et al.R18	Functional testing	215	41	1	Stress Echo
15	Jeetley P., et al.R18	Functional testing	218	72	1	Exercise ECG
16	Nucifora G., et al.R19	Functional testing	110	6	2	Stress Echo
16	Nucifora G., et al.R19	Functional testing	89	6	2	Exercise ECG
17	Jeetley P., et al.R20	Functional testing	148	21	2	Stress Echo
17	Jeetley P., et al.R20	Functional testing	154	36	2	Exercise ECG
18	Udelson JE., et al.R21	Functional testing	1215	156	2	SPECT-MPI
18	Udelson JE., et al.R21	Standard care	1260	162	2	Standard care

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Table S2 Number of "one-step loops" providing indirect evidence for NMA relative treatment effects between treatment comparisons for the network of antidepressants.

Number of "one-step loops" providing indirect evidence	Nr of treatment comparisons	Cumulative frequency	% Cumulative frequency
0	3	3	2%
1	16	19	14%
2	18	37	28%
3	32	69	51%
4	30	99	74%
5	19	118	88%
6	13	131	98%
7	13	144	107%
8	3	147	110%
10	1	148	110%
11	1	149	111%
12	3	152	113%
13	1	153	114%

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Table S3 Average contribution to NMA relative treatment effects from direct evidence and indirect evidence via intermediate comparators (steps). The "one-step loop" provides one-step indirect comparison via a single common treatment.

Source of evidence		% Contribution
Direct evidence		11.1%
Indirect evidence	1 step	56.1%
	2 steps	84.3%
	3 steps	95.0%
	4 steps	98.0%
	5 steps	99.1%
	6 steps	99.5%
	7 steps	99.7%
	8 steps	99.9%
	9 steps	100.0%
	10 steps	100.0%
	11 steps	100.0%
	12 steps	100.0%
	13 steps	100.0%

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Table S4. Data from the network of randomised controlled trials comparing adverse effects of statins. The data was originally published by Naci et al. id: id of the study, t: treatment name, r: number of adverse effects, n: sample size.

year	study	id	t	r	n
1993	PMSG	1	pravastatin	25	530
1993	PMSG	1	placebo	33	532
1993	SPSG	2	simvastatin	5	275
1993	SPSG	2	pravastatin	5	275
1993	LPSG	3	lovastatin	10	339
1993	LPSG	3	pravastatin	8	333
1993	MARS	4	lovastatin	3	123
1993	MARS	4	placebo	6	124
1994	4S	5	placebo	129	2223
1994	4S	5	simvastatin	126	2221
1994	PMSG-Diabetes	6	pravastatin	2	167
1994	PMSG-Diabetes	6	placebo	9	158
1994	EXCEL	7	placebo	100	1663
1994	EXCEL	7	lovastatin	329	6582
1994	OCS	8	simvastatin	18	414
1994	OCS	8	placebo	6	207
1995	Jacobson	9	pravastatin	9	182
1995	Jacobson	9	placebo	1	63
1995	REGRESS	10	pravastatin	16	450
1995	REGRESS	10	placebo	10	434
1995	KAPS	11	placebo	12	223
1995	KAPS	11	pravastatin	8	224
1995	WOSCOPS	12	placebo	106	3293
1995	WOSCOPS	12	pravastatin	116	3302
1995	Guillen	13	placebo	1	74
1995	Guillen	13	pravastatin	0	76
1996	SHIGA Pravastatin study	14	pravastatin	2	102
1996	SHIGA Pravastatin study	14	placebo	0	105
1996	CARE	15	placebo	74	2078
1996	CARE	15	pravastatin	45	2081
1996	QLMG	16	lovastatin	3	211
1996	QLMG	16	pravastatin	4	215
1996	CHESS	17	simvastatin	27	453
1996	CHESS	17	atorvastatin	65	464
1997	Bertolini	18	atorvastatin	7	227
1997	Bertolini	18	pravastatin	2	78
1997	ASG-I	19	atorvastatin	16	529
1997	ASG-I	19	lovastatin	5	120
1998	AFCAPS-TexCAPS	20	placebo	455	3301

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1998	AFCAPS-TexCAPS	20	lovastatin	449	3304
1998	Brown	21	atorvastatin	3	78
1998	Brown	21	fluvastatin	4	76
1998	Brown	21	lovastatin	2	78
1998	Brown	21	simvastatin	2	76
1999	TARGET TANGIBLE	22	atorvastatin	89	1897
1999	TARGET TANGIBLE	22	simvastatin	45	959
1999	Riegger	23	fluvastatin	11	187
1999	Riegger	23	placebo	8	178
1999	IQLMG	24	simvastatin	7	194
1999	IQLMG	24	pravastatin	7	193
1999	FLARE	25	fluvastatin	4	409
1999	FLARE	25	placebo	11	425
2000	Barter	26	atorvastatin	48	691
2000	Barter	26	simvastatin	24	337
2000	Farnier	27	atorvastatin	1	109
2000	Farnier	27	simvastatin	1	163
2000	Stein	28	placebo	0	130
2000	Stein	28	simvastatin	1	260
2000	Recto	29	simvastatin	1	251
2000	Recto	29	atorvastatin	5	251
2000	Gentile	30	atorvastatin	1.5	85
2000	Gentile	30	simvastatin	0.5	79
2000	Gentile	30	pravastatin	1.5	82
2000	Gentile	30	lovastatin	1.5	81
2000	Gentile	30	placebo	0.5	87
2001	ASSET	31	atorvastatin	7	730
2001	ASSET	31	simvastatin	7	694
2001	MIRACL	32	placebo	33	1548
2001	MIRACL	32	atorvastatin	40	1538
2001	Paoletti	33	rosuvastatin	8	230
2001	Paoletti	33	pravastatin	3	136
2001	Paoletti	33	simvastatin	1	129
2001	Andrews	34	atorvastatin	129	1902
2001	Andrews	34	fluvastatin	64	477
2001	Andrews	34	lovastatin	42	476
2001	Andrews	34	pravastatin	20	462
2001	Andrews	34	simvastatin	39	468
2002	GREACE	35	atorvastatin	6	800
2002	GREACE	35	placebo	3	800
2002	Davidson	36	placebo	3	70
2002	Davidson	36	simvastatin	14	263
2002	FLORIDA	37	fluvastatin	30	265
2002	FLORIDA	37	placebo	37	275

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2002	LIPS	38	fluvastatin	174	844
2002	LIPS	38	placebo	196	833
2002	PROSPER	39	placebo	116	1913
2002	PROSPER	39	pravastatin	107	2891
2002	Olsson	40	rosuvastatin	16	272
2002	Olsson	40	atorvastatin	12	140
2002	Davidson	41	placebo	7	132
2002	Davidson	41	rosuvastatin	10	259
2002	Davidson	41	atorvastatin	4	128
2002	CHALLENGE	42	atorvastatin	17	846
2002	CHALLENGE	42	simvastatin	10	848
2003	Ballantyne	43	atorvastatin	13	248
2003	Ballantyne	43	placebo	3	60
2003	ADVOCATE	44	atorvastatin	6	82
2003	ADVOCATE	44	simvastatin	2	76
2003	Bruckert	45	fluvastatin	13	607
2003	Bruckert	45	placebo	8	622
2003	Kerzner	46	lovastatin	10	220
2003	Kerzner	46	placebo	5	64
2003	Melani	47	pravastatin	3	205
2003	Melani	47	placebo	5	65
2003	TREAT TO TARGET	48	atorvastatin	20	552
2003	TREAT TO TARGET	48	simvastatin	14	535
2003	HeFH	49	rosuvastatin	16	436
2003	HeFH	49	atorvastatin	6	187
2003	Mohler	50	atorvastatin	16	240
2003	Mohler	50	placebo	10	114
2003	Davidson	51	lovastatin	21	501
2003	Davidson	51	fluvastatin	22	337
2003	STELLAR	52	rosuvastatin	9	480
2003	STELLAR	52	atorvastatin	25	641
2003	STELLAR	52	simvastatin	19	655
2003	STELLAR	52	pravastatin	11	492
2004	CARDS	53	placebo	145	1410
2004	CARDS	53	atorvastatin	122	1428
2004	Bays	54	simvastatin	31	622
2004	Bays	54	placebo	2	148
2004	PREVENT IT	55	placebo	22	431
2004	PREVENT IT	55	pravastatin	13	433
2004	Durazzo	56	atorvastatin	1	50
2004	Durazzo	56	placebo	0	50
2004	Goldberg	57	placebo	2	93
2004	Goldberg	57	simvastatin	7	349
2004	ALLIANCE	58	atorvastatin	75	1217

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2004	ALLIANCE	58	placebo	3	1225
2004	PCS	59	pravastatin	5	54
2004	PCS	59	placebo	0	66
2004	REVERSAL	60	pravastatin	22	327
2004	REVERSAL	60	atorvastatin	21	327
2004	DISCOVERY	61	rosuvastatin	24	686
2004	DISCOVERY	61	atorvastatin	9	338
2004	Schwatrz	62	rosuvastatin	12	255
2004	Schwatrz	62	atorvastatin	6	128
2004	Brown	63	rosuvastatin	22	239
2004	Brown	63	pravastatin	11	118
2004	Brown	63	simvastatin	9	120
2005	BELLES	64	atorvastatin	43	305
2005	BELLES	64	pravastatin	21	309
2005	DISCOVERY-Penta	65	rosuvastatin	17	358
2005	DISCOVERY-Penta	65	atorvastatin	7	383
2005	IDEAL	66	simvastatin	186	4449
2005	IDEAL	66	atorvastatin	426	4439
2005	CORALL	67	rosuvastatin	9	131
2005	CORALL	67	atorvastatin	11	132
2005	URANUS	68	rosuvastatin	3	232
2005	URANUS	68	atorvastatin	7	233
2005	COMETS	69	rosuvastatin	4	165
2005	COMETS	69	atorvastatin	4	157
2005	COMETS	69	placebo	3	79
2006	DISCOVERY-Alpha	70	rosuvastatin	23	555
2006	DISCOVERY-Alpha	70	atorvastatin	14	382
2006	SPARCL	71	atorvastatin	415	2365
2006	SPARCL	71	placebo	342	2366
2006	ASPEN	72	atorvastatin	33	1211
2006	ASPEN	72	placebo	38	1199
2006	PULSAR	73	rosuvastatin	14	504
2006	PULSAR	73	atorvastatin	11	492
2006	ARIES	74	rosuvastatin	13	391
2006	ARIES	74	atorvastatin	10	383
2006	STARSHIP	75	rosuvastatin	11	357
2006	STARSHIP	75	atorvastatin	5	339
2006	MERCURY II	76	rosuvastatin	15	392
2006	MERCURY II	76	atorvastatin	19	798
2006	MERCURY II	76	simvastatin	25	803
2007	METEOR	77	rosuvastatin	79	702
2007	METEOR	77	placebo	22	282
2007	SAGE	78	atorvastatin	48	446

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2007	SAGE	78	pravastatin	46	445
2007	Kyeong	79	rosuvastatin	2	60
2007	Kyeong	79	atorvastatin	3	57
2007	ASTRONOMER	80	rosuvastatin	25	134
2007	ASTRONOMER	80	placebo	26	135
2007	CORONA	81	placebo	302	2497
2007	CORONA	81	rosuvastatin	241	2514
2007	Lewis	82	pravastatin	11	163
2007	Lewis	82	placebo	16	163
2007	ANDROMEDA	83	rosuvastatin	15	248
2007	ANDROMEDA	83	atorvastatin	13	246
2007	POLARIS	84	rosuvastatin	22	432
2007	POLARIS	84	atorvastatin	27	439
2007	DISCOVERY-Asia	85	rosuvastatin	21	950
2007	DISCOVERY-Asia	85	atorvastatin	10	472
2007	SOLAR	86	rosuvastatin	15	542
2007	SOLAR	86	atorvastatin	20	544
2007	SOLAR	86	simvastatin	20	546
2007	IRIS	87	rosuvastatin	14	371
2007	IRIS	87	atorvastatin	7	369
2008	GISSI-HF	88	rosuvastatin	104	2285
2008	GISSI-HF	88	placebo	91	2289
2008	ECLIPSE	89	rosuvastatin	41	522
2008	ECLIPSE	89	atorvastatin	36	514
2008	SUBARU	90	atorvastatin	0	213
2008	SUBARU	90	rosuvastatin	8	214
2008	DISCOVERY-Beta	91	rosuvastatin	24	334
2008	DISCOVERY-Beta	91	simvastatin	7	170
2008	Sdringola	92	placebo	3	73
2008	Sdringola	92	atorvastatin	1	72
2009	SPACE ROCKET	93	rosuvastatin	20	633
2009	SPACE ROCKET	93	simvastatin	9	630
2009	Ose	94	simvastatin	4	219
2009	Ose	94	pitavastatin	21	638
2010	CENTAURUS	95	rosuvastatin	15	437
2010	CENTAURUS	95	atorvastatin	17	450
2010	Acala	96	placebo	0	70
2010	Acala	96	pravastatin	2	61
2011	SATURN	97	atorvastatin	48	519
2011	SATURN	97	rosuvastatin	45	520
2011	Eriksson	98	pitavastatin	9	236
2011	Eriksson	98	simvastatin	6	119
2011	Gumprecht	99	pitavastatin	8	275

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2011	Gumprecht	99	atorvastatin	6	137
2011	PATROL	100	atorvastatin	13	101
2011	PATROL	100	rosuvastatin	10	100
2011	PATROL	100	pitavastatin	12	101
2012	LUNAR	101	rosuvastatin	26	499
2012	LUNAR	101	atorvastatin	25	257

Table S5. NMA results from the network of randomised controlled trials comparing adverse effects of statins. Odds ratios and their 95% confidence intervals are presented. Odds ratios less than 1 favor the treatment specified in the row.

Atorvastatin	0.894 (0.637, 1.255)	1.196 (0.868, 1.647)	1.127 (0.637, 1.994)	1.073 (0.890, 1.294)	1.418 (1.126, 1.785)	1.007 (0.848, 1.196)	1.297 (1.065, 1.580)
1.119 (0.797, 1.571)	**Fluvastatin**	1.338 (0.915, 1.956)	1.261 (0.655, 2.426)	1.201 (0.879, 1.640)	1.586 (1.109, 2.268)	1.127 (0.787, 1.612)	1.451 (1.011, 2.083)
0.836 (0.607, 1.152)	0.747 (0.511, 1.093)	**Lovastatin**	0.942 (0.494, 1.797)	0.897 (0.668, 1.206)	1.185 (0.849, 1.655)	0.842 (0.599, 1.184)	1.085 (0.768, 1.532)
0.887 (0.501, 1.570)	0.793 (0.412, 1.526)	1.061 (0.557, 2.023)	**Pitavastatin**	0.952 (0.528, 1.718)	1.258 (0.687, 2.305)	0.893 (0.499, 1.598)	1.151 (0.648, 2.043)
0.932 (0.773, 1.123)	0.833 (0.610, 1.138)	1.114 (0.829, 1.498)	1.050 (0.582, 1.895)	**Placebo**	1.321 (1.070, 1.632)	0.938 (0.759, 1.160)	1.209 (0.960, 1.522)
0.705 (0.560, 0.888)	0.630 (0.441, 0.902)	0.844 (0.604, 1.178)	0.795 (0.434, 1.457)	0.757 (0.613, 0.935)	**Pravastatin**	0.710 (0.549, 0.918)	0.915 (0.702, 1.193)
0.993 (0.836, 1.179)	0.888 (0.620, 1.270)	1.188 (0.844, 1.671)	1.119 (0.626, 2.002)	1.066 (0.862, 1.317)	1.408 (1.089, 1.821)	**Rosuvastatin**	1.288 (1.024, 1.621)
0.771 (0.633, 0.939)	0.689 (0.480, 0.989)	0.922 (0.653, 1.302)	0.869 (0.489, 1.542)	0.827 (0.657, 1.041)	1.093 (0.839, 1.424)	0.776 (0.617, 0.977)	**Simvastatin**

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Table S6. Results from SIDE splitting for three network comparisons of the network of statins. OR: odds ratio. SIDE: separate indirect from direct approach.

Comparison	Direct OR	Indirect OR	Ratio of ORs	z-value	p-value
Pravastatin versus rosuvastatin	0.98	0.67	1.47	1.06	0.29
Pravastatin versus simvastatin	0.84	0.95	0.89	-0.42	0.67
Rosuvastatin versus simvastatin	1.23	1.32	0.93	-0.27	0.78