

# 1 Reporting standards and Bayesian network meta-analysis

As with many Bayesian algorithms, recent developments in underlying sampling algorithms (such as **stan**'s implementation in **R**), have enabled advances in Bayesian network meta-analysis (NMA). Software has become available, such as **multinma::**, released in 2021, which computes Bayesian estimates for NMA using **stan** from an **R** interface. The key attraction of NMA is how it allows a comparison of multiple treatments, even where direct pairwise comparisons are not empirically available. However, a challenge exists for researchers who wish to both use these techniques and adhere to appropriate evidence synthesis reporting standards for a given discipline. One way of meeting this challenge is to provide a computational waystation for the development of toolchains for reporting NMA according to different standards, such as Cochrane or the Campbell Collaboration. This manuscript provides some toolchain and commentary about the process of computationally implementing a network meta-analysis according to Cochrane's reporting standards and how we might expand our research compendia to include in-development computational science workflows.

## 20 1.1 Network meta-analysis

21 Pairwise analyses between treatment and control, exposed and unexposed,  
22 intervention and no intervention, are conventionally undertaken with meta-  
23 analysis in fields such as ecology, medicine, and the social sciences [2]. NMA  
24 provides a means of comparing three or more treatments or interventions,  
25 including control or placebo [5]. The question answered by an NMA is not *if*  
26 a treatment works, but *which* treatments perform better, comparatively [3].  
27 A particularly useful aspect of NMA is combining the results of more than  
28 one pairwise analysis and constructing indirect comparisons, where pairwise  
29 evidence is unavailable, from a network of direct comparisons. An example  
30 of direct comparisons provided by existing evidence is shown in Figure 1.  
31 NMA converts the network to a complete graph, where all treatments are  
32 compared with all other treatments, that is, every node connects to every  
33 other node via **direct** or **indirect** comparison.

## 34 1.2 Reporting standards

35 Reporting NMA according to Cochrane’s expectations is a non-trivial task;  
36 there are multiple visualisations and tables required, in accordance with stan-  
37 dards that have been adapted and derived from existing evidence synthesis  
38 standards. Some familiarity with this ecosystem of reporting systems is re-  
39 quired to meet the standards required for reporting a NMA according to  
40 Cochrane’s expectations.

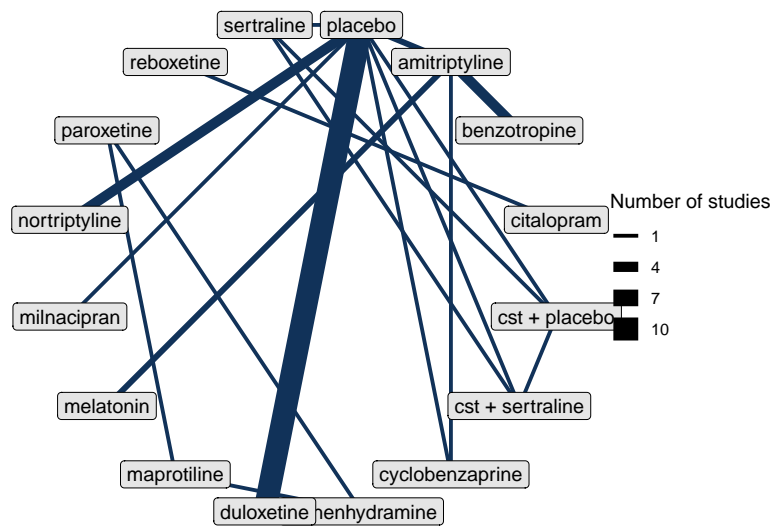


Figure 1: Network of treatment comparisons, the direct evidence, included in the network meta-analysis for how antidepressants affect sleep. This analysis will be discussed in detail in Section 2, where we describe the data and the toolchains for analysis and reporting.

41 Figure 2 shows the immediate development relationships between report-  
42 ing standards for NMA in Cochrane intervention reviews. A researcher un-  
43 dertaking NMA must ensure they meet (MECIR) and (PRISMA) Exten-  
44 sion guidelines, as MECIR was developed for pairwise meta-analysis and the  
45 PRISMA Extension provides further guidance on NMA.

46 This figure also provides further information, to demonstrate different  
47 ways a researcher may encounter reporting standards. (QUOROM) guide-  
48 lines have been superseded by PRISMA, from which MECIR was developed.  
49 If a researcher is undertaking a Campbell systematic review, they will consult  
50 (MECCIR), which was derived from MECIR. MECIR was derived from the  
51 Cochrane Handbook, which was informed by PRISMA guidelines.

## 52 **2 Toolchains for NMA reporting**

53 To illustrate the computational challenges of reporting results using newer  
54 statistical methods, we step through providing reporting NMA results for a  
55 single analysis from an in-development Cochrane intervention review. Before  
56 discussing the codeflow in detail, we describe the data analysed.

### 57 **2.1 Antidepressants for chronic pain in adults**

58 These data are a preliminary subset of results from a currently underway  
59 Cochrane study examining the use of antidepressants to treat chronic pain  
60 in adults.

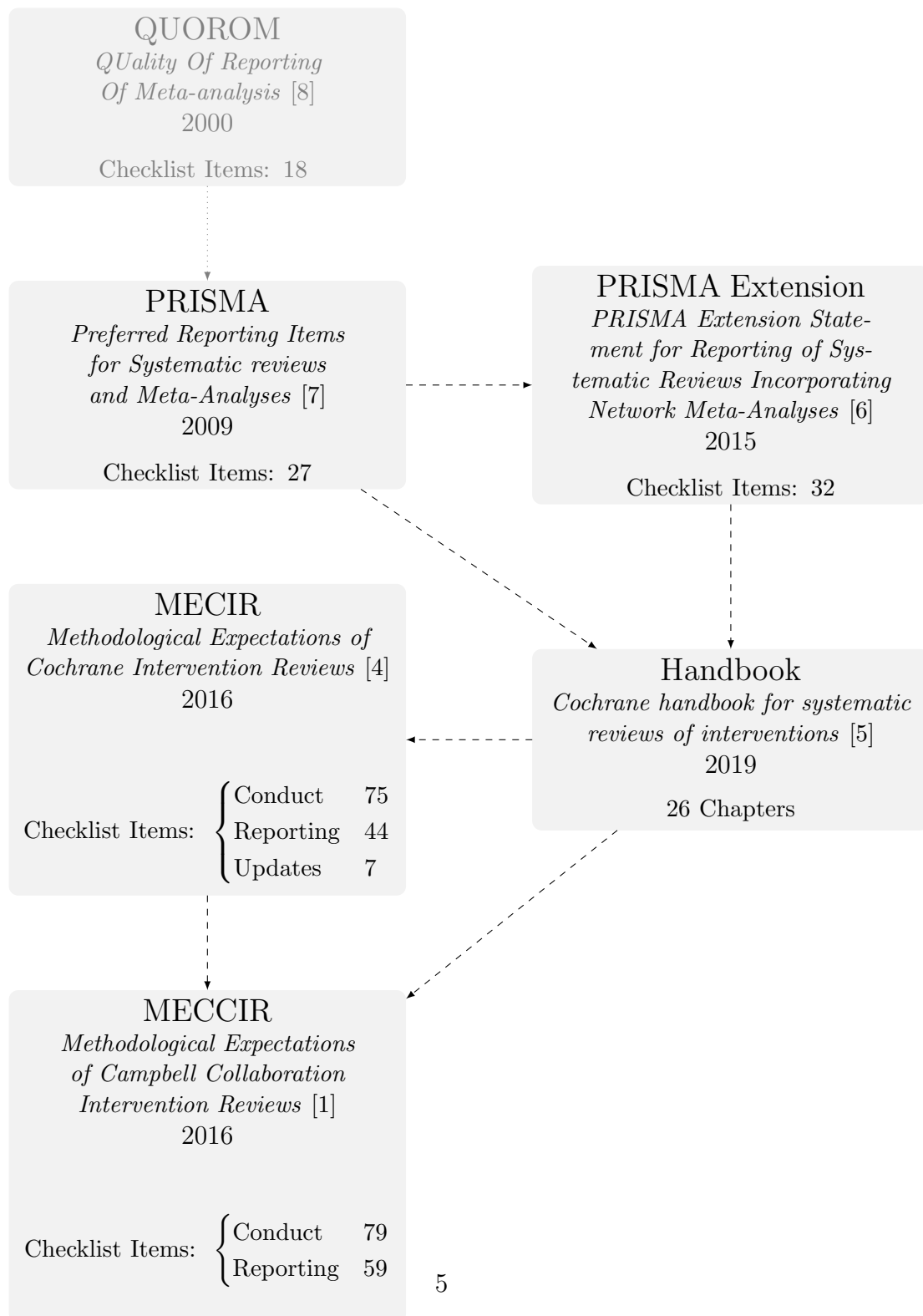


Figure 2: Development network in which reporting protocols for NMA in Cochrane intervention reviews have been developed.

Table 1: Number of observations we have, and of some types we care about in this analysis.

*draft notes* Cite protocol

These data are intended to be computationally illustrative only, as results and models are not fully developed and only reflect half of the extraction to be completed.

*draft notes* May need to update this if I change datasets

In addition to pain, the study examines other ways individuals are affected by chronic pain, such as depression and quality of life. These data present measures of insomnia experienced by chronic pain, with pairwise comparisons between antidepressant interventions and placebo groups.

*draft notes* Discuss number of studies and other characteristics, treatments, dosages, etc.

## 2.2 Toolchain for reporting the results of one outcome

In this section we focus on computationally achieving mandatory MECIR reporting for effects and interventions (R76-R99) [4] on the sleep dataset described above. The examples provided are selected to demonstrate the challenges and nuances of reporting NMA according to MECIR in R.

*draft notes* Discuss with Gav how this relates to MECIR; which reporting protocols to include, R76-99?

### 77 2.2.1 Number of studies and participants: R78

78 To meet this expectation, we need to *state how many studies and how many*  
79 *participants contributed data to results for each outcome, along with the pro-*  
80 *portion of included studies.* This can be achieved with a summary table  
81 with a **contribution matrix**, which shows the proportions of studies' direct  
82 evidence contributions to the treatment comparisons.

83 *draft notes* Is this how to interpret this reporting standard?

### 84 2.2.2 Estimated outcomes and associated uncertainty (R82-84, 85 R88, R93, R95)

86 Outcome effect estimates need to be provided (R82), along with a measure  
87 (R82) of uncertainty (we will use confidence interval) with a specified level of  
88 confidence, instead of p-values (R84) [4]. The direction (R88) of the results  
89 also needs to be reported. Much of this can be achieved with a forest plot  
90 (R93) with appropriate labels (R95) and a table.

### 91 2.2.3 Sensitivity analysis (R94)

92 Multiple sensitivity analyses should be presented in summary form, not mul-  
93 tiple forest plots (R94) [4]. A threshold analysis is not only a convenient way  
94 to summarise multiple sensitivity analyses, but more robust [9].

### 95 3 Role of computational waystations

96 The toolchain in Section 2 demonstrates the computational gap between  
97 statistical methodology and reporting implementation according to a par-  
98 ticular protocol. Even where there are well-featured software tools, such  
99 as `multinma::`, different reporting protocols for particular disciplines across  
100 which a statistical methodology may be applied will have different require-  
101 ments. Thus it is unsurprising that for recently-developed statistical method-  
102 ology there are toolchain gaps that an applied scientist will need to bridge.  
103 Computational waystations provide an open science forum in which to de-  
104 velop toolchains, share knowledge, and extend on existing tools.

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