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REVIEW

What are the best methods for rapid reviews of the research evidence? A systematic review of reviews and primary studies

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Abstract

Rapid review methodology aims to facilitate faster conduct of systematic reviews to meet the needs of the decision-maker, while also maintaining quality and credibility. This systematic review aimed to determine the impact of different methodological shortcuts for undertaking rapid reviews on the risk of bias (RoB) of the results of the review. Review stages for which reviews and primary studies were sought included the preparation of a protocol, question formulation, inclusion criteria, searching, selection, data extraction, RoB assessment, synthesis, and reporting. We searched 11 electronic databases in April 2022, and conducted some supplementary searching. Reviewers worked in pairs to screen, select, extract data, and assess the RoB of included reviews and studies. We included 15 systematic reviews, 7 scoping reviews, and 65 primary studies. We found that several commonly used shortcuts in rapid reviews are likely to increase the RoB in the results. These include restrictions based on publication date, use of a single electronic database as a source of studies, and use of a single reviewer for screening titles and abstracts, selecting studies based on the full-text, and for extracting data. Authors of rapid reviews should be transparent in reporting their use of these shortcuts and acknowledge the possibility of them causing bias in the results. This review also highlights shortcuts that can save time without increasing the risk of bias. Further research is needed for both systematic and rapid reviews on faster methods for accurate data extraction and RoB assessment, and on development of more precise search strategies.

KEYWORDS

evidence synthesis, rapid review methods, recommendations, risk of bias, systematic review

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Highlights

What is already known?

- A large variety of methodological shortcuts are used to save time when conducting rapid reviews.
- The rationale for the choice of specific shortcuts is often only loosely based on high quality evidence and can increase the risk of bias of the results; this is rarely acknowledged by rapid review authors.

What is new?

- This systematic review updates and extends previous reviews that were limited to specific systematic review stages or did not include a risk of bias assessment.
- We found that several commonly used shortcuts in rapid reviews are likely to increase the risk of bias in the results.

Potential impact for *Research Synthesis Methods* readers

- We highlight shortcuts that can save time without increasing the risk of bias of rapid reviews.
- Areas where further research is required for both rapid and systematic reviews include methods for accurate (and faster) data extraction and risk of bias assessment, and on development of more precise search strategies.

1 | BACKGROUND

The evidence-informed approach to decision-making aims to achieve better decisions for better health, avoid harm, make more effective use of scarce resources, and to improve transparency and accountability in decision-making.^{1,2} Evidence-informed decision-making is a systematic and transparent approach and includes decisions about clinical practice, public health, and health policy and systems. Decision-makers can include healthcare policy-makers, government agencies, clinicians and their professional associations, patients, caregivers, patient groups, and the public.³ The most frequently reported barriers to evidence uptake for evidence-informed decision-making are poor access to good quality relevant research and lack of timely and relevant research output.⁴ In relation to good quality relevant research, high-quality systematic reviews are considered the gold standard,^{2,5} and these are used as the basis for evidence products, such as policy and practice guidelines, health technology assessments, and evidence briefs for policy.² However, there is a well-recognized need to conduct systematic reviews faster and with the needs of the decision-maker in mind, while also maintaining quality (low risk of bias) and credibility. This need was particularly emphasized during the COVID-19 pandemic with various rapid reviews produced.^{6–9}

Rapid reviews are “a type of systematic review in which components of the systematic review process are

simplified, omitted or made more efficient in order to produce information in a shorter period of time, preferably with minimal impact on quality. Further, they involve a close relationship with the end-user and are conducted with the needs of the decision-maker in mind.”^{10,11} In 2015, we conducted a rapid review of systematic reviews and primary studies to answer the question: What are the best methodologies to enable a rapid review of research evidence for evidence-informed decision making in health policy and practice?¹⁰ As well as offering the definition for rapid reviews cited above, and informing suitable methods for rapid reviews, the review was used to inform the design of a rapid response program to support evidence-informed decision-making.¹² The review has been widely cited, showing its usefulness, particularly during the COVID-19 pandemic. However, there have been some significant research outputs related to rapid reviews since then, including in response to the COVID-19 pandemic, that need to be considered for inclusion in an updated version.^{6,8,13–15} There have also been updates to methods for conducting systematic reviews,^{5,16} and for assessing their quality or risk of bias (RoB).^{17,18}

There is a need to provide stronger guidance about choice of methods for rapid reviews that considers potential impact on RoB and covers the key steps of review conduct. While we have previously offered areas where “shortcuts” could be considered to reduce time to completion of rapid reviews based on our review findings and

other sources of evidence,¹² our 2015 rapid review has been used selectively¹⁰ to justify a range of shortcuts regardless of their impact on the RoB. Work done since the last version of our rapid review on best methodologies for rapid reviews could help to inform the update.^{14,15,19} These include a scoping review of rapid review methodology conducted by Hamel and colleagues¹⁵ that was used by Cochrane to offer interim recommendations for conducting rapid reviews⁶; and a systematic review of methods for conducting various steps of systematic reviews that was conducted by Robson and colleagues.^{14,19} However, the review by Hamel et al. 2020¹⁵ is limited by a lack of RoB assessment of the primary studies included, which is usual practice for scoping reviews, and the date of last search for studies is now over 4 years ago (February 2019). The review by Robson et al.^{14,19} used high-quality systematic review methods, but the date of last search for studies was September 1, 2016. While the review was aimed at systematic review methods, many of these would also be applicable for rapid reviews.

We aimed to answer the questions:

1. What is the accuracy, reliability, impact and/or efficiency of different methodological shortcuts for undertaking rapid reviews, including for preparation of a protocol, question formulation, inclusion criteria, searching, selection, data extraction, RoB assessment, synthesis, and reporting? and
2. What is the potential impact of the methodological shortcut/s on the RoB of the results of the rapid review?

2 | METHODS

High-quality systematic review methods were used.⁵ The protocol was registered on the International prospective register of systematic reviews (PROSPERO).²⁰ Changes made to the protocol after registration can be found in Supporting Information File S1. This review is an update of a review published in 2016^{10,21} but with some adjustments to the methods to account for recent developments and learning in the field, as well as limitations in the original review. The Preferred Reporting Items for Systematic Reviews and Meta-Analysis (PRISMA) statement was used for reporting.²²

2.1 | Criteria for considering studies for inclusion

Publications in any language and from any country were included. Both gray and peer-reviewed literature was included. There was no date of publication restrictions.

2.1.1 | Types of studies

Systematic reviews and scoping reviews not included in our original review were included. Primary studies were included if they compared or evaluated the accuracy, reliability, or efficiency of a potential methodological shortcut or described factors that affect the method's accuracy, reliability, or efficiency. Primary research studies providing quantitative data and one of the following designs were eligible: randomized controlled trials, non-randomized controlled trials, controlled before-after studies, interrupted time series studies, repeated measures studies, cohort, case-control, analytical cross-sectional studies. Modeling/simulation studies were included. Qualitative studies were excluded.

2.1.2 | Types of participants

These include reviewers, reviews (systematic, rapid, scoping, and living reviews), studies, articles, and database records.

2.1.3 | Types of interventions

Methodological shortcuts for undertaking rapid reviews, including for preparation of a protocol, question formulation, inclusion criteria, searching, selection, data extraction, RoB assessment, synthesis, and reporting were included.

2.1.4 | Types of comparisons

Suitable comparisons include an alternative method or no comparator.

2.1.5 | Types of outcome measures

Relevant outcomes included accuracy, for example, sensitivity, specificity; reliability; efficiency, for example, time to complete, resources required to complete (e.g., monetary and personnel); concordance; measures of synthesis quality; and agreement in review conclusions.

2.2 | Search strategy

We actively searched for systematic reviews and scoping reviews published since January 2015. For the primary studies, we used the two previous reviews as the source of studies prior to 2019.^{14,15} Then we actively searched for primary studies published after January 2019.

We searched MEDLINE (Ovid), EMBASE (Ovid), PsycINFO (Ovid), LILACS (BVSsalud), and the Cochrane Library (Ovid), including the Cochrane Database of Systematic Reviews and the Cochrane Central Register of Controlled Trials (CENTRAL). The databases were searched from January 2015 to present for systematic reviews and scoping reviews; and from January 2019 to present for primary studies—using search strategies specific to the study type and database. The date of last search was April 23–25, 2022. Details of the searches can be found in Supporting Information File S2.

We also searched the reference lists of the two published reviews for primary studies prior to January 2019.^{14,15} The author's own databases of knowledge translation literature and some of the included studies were also searched by hand for relevant studies. Searches were conducted by one review author (MMH) and references were imported into EndNote for management and removal of duplicates.

2.3 | Study selection and data collection

Titles and abstracts were screened according to the selection criteria by two review authors independently (MMH and DEGM) in Endnote. The full text of any potentially relevant papers selected by either reviewer was retrieved for closer examination. The inclusion criteria were applied against these papers by two reviewers independently (MMH and one of JOMB, SP, or LR) using Covidence. Discrepancies were resolved by discussion and consultation with a third reviewer (MT).

All relevant data was extracted from included papers by one reviewer and verified by a second reviewer (MH, CM, JOMB, MT, SP—working in pairs). Differences were resolved by discussion and consensus. Data were extracted into an Excel spreadsheet following a guidance document in Word (Supporting Information File S3). Data extracted included the objectives, target population, method/s tested, outcomes reported, date of last search or year of study (for reviews and primary studies, respectively), included study designs and number of studies (for reviews), study design and size (for primary studies), country of study, results, conclusions, and comments, for example, strengths, limitations, research gaps.

2.4 | Assessment of RoB of included studies

The RoB of included systematic reviews and scoping reviews was assessed by one reviewer and verified by

another (CM, MT, and SP—working in pairs) using the Risk of Bias Assessment Tool for Systematic Reviews (ROBIS).¹⁸ Disagreements regarding scores were resolved by discussion and consensus. ROBIS covers four domains: 1) study eligibility criteria, 2) identification and selection of studies, 3) data collection and study appraisal, and 4) synthesis and findings. Each domain consists of five to six questions with six possible options: Yes, Probably yes, Probably No, No, Not indicated, or Not applicable. By design, scoping reviews do not include an assessment of RoB of the included studies or a meta-analysis. Thus, some questions were not applicable to scoping reviews (i.e., questions 3.4, 3.5, 4.5, and 4.6 in domains 3 and 4). The overall assessment for each domain and for the review overall is low, unclear, or high RoB.¹⁸

The quality of included primary studies was assessed by one reviewer (JYHK) and verified by another (MH or SP) using a modified version of the Quality Assessment of Diagnostic Studies (QUADAS)-2 checklist,²³ which was used in the review by Robson et al. 2019.¹⁴ Disagreements regarding scores were resolved by discussion and consensus. We used the same modified version as Robson et al. 2019.¹⁴ This included the four domains: (1) Participant selection, (2) Record selection, (3) Evaluated methodological shortcut, and (4) Reference standard. The overall assessment for each domain is low, unclear, or high RoB.²³ For the primary studies that were included in the Robson review we used their assessment, which was also conducted by one reviewer and verified by another.

2.5 | Data synthesis

We conducted a narrative synthesis of the included reviews and studies. No meta-analyses were performed due to the heterogeneous study designs and outcome measures. Both the reviews and primary studies were classified according to the stage of the review (preparation of a protocol, question formulation, inclusion criteria, searching, selection, data extraction, RoB assessment, synthesis, and reporting) and then by the particular methodological shortcut. Reviews and primary studies that covered a range of shortcuts or rapid review methods in general are grouped together and findings are presented according to key themes.

The results of the reviews and primary studies were also mapped to the A MeaSurement Tool to Assess systematic Reviews (AMSTAR) 2, ROBIS, and the Methodological Expectations of Cochrane Intervention Reviews (MECIR) criteria.^{17,18,24} When interpreting the impact of the shortcut on the speed or RoB of the review, greater

weight was given to reviews and studies with the lowest RoB. Shortcuts that may impact on the RoB of the review were distinguished from those that could help to speed up the process but do not impact on the RoB.

In addition, when reporting the results of the review, we have suggested some shortcuts that are unlikely to increase the risk of bias but that were not tested in any of the included reviews or studies (see Table 2 in the results). These suggestions are based on recommendations from included reviews, from rapid review and systematic review guidance documents, and our own experience with rapid and systematic reviews. The statement that they are unlikely to increase the risk of bias is based on them not negatively affecting the rating of either the AMSTAR 2 or ROBIS tools.^{17,18}

3 | RESULTS

We identified 1976 records after duplicates were removed and assessed 162 full-text reports for eligibility (Figure 1). Of these, we excluded 71 reports, with the most common reasons for exclusion being wrong intervention/method ($n = 35$) or wrong study design ($n = 23$) (Figure 1 and Supporting Information File S4). We included 15 systematic reviews^{25–39} with 2 secondary references,^{40,41} 7 scoping reviews,^{42–48} and 65 primary studies^{49–113} with 2 secondary references.^{114,115} A small number of the included primary studies were also included in one or more of the included systematic reviews. Also, one systematic review of systematic reviews by Veginadu and colleagues³⁸ overlapped 100% with other included systematic reviews.^{14,25–28,33–35,37} We have highlighted these instances when presenting the results to prevent double counting.

3.1 | Characteristics of included studies

3.1.1 | Systematic reviews and scoping reviews

The countries where the reviews were most commonly conducted (based on the country affiliation of the first author) are Canada ($n = 7$), Germany ($n = 5$), United Kingdom ($n = 4$), and Australia ($n = 2$). The systematic reviews most commonly covered one particular review stage ($n = 12$), including aspects of the inclusion criteria,^{27,34,37} searching,^{27,34,37} selection,^{35,39} data extraction,^{31,36} or RoB assessment.³³ Three of the systematic reviews addressed rapid review methods in general.^{29,32,38} The scoping reviews all addressed rapid review methods in general,^{42–48} but with one also scoping the different definitions for rapid reviews,⁴⁴ one focusing on tools to support the automation

of systematic reviews,⁴⁵ and one on resource use during systematic review production.⁴⁶ The characteristics of the included reviews sorted by review stage and methodological shortcut can be found in Supporting Information File S5. The overall RoB is summarized in Figure 2 and the assessment for each study can be found in Supporting Information File S6.

3.1.2 | Primary studies

Of the 65 included primary studies, 6 were randomized controlled trials.^{51,55,62,63,75,83} The countries where studies were most commonly conducted (based on the country affiliation of the first author) are Canada ($n = 18$), United States of America ($n = 15$), United Kingdom ($n = 12$), and Australia ($n = 8$). The studies covered methods for one or more of the different review stages, including protocol preparation ($n = 1$), inclusion criteria ($n = 4$), searching ($n = 14$), selection ($n = 32$), data extraction ($n = 11$), and RoB assessment ($n = 2$). No studies specifically focused on question formulation separate to the inclusion criteria, or on synthesis or reporting. Four studies each covered two or three stages,^{58,76,90,110} and six studies tested a package of methods across all (or most) review stages.^{60,77,95,99,105,111} The characteristics of the included studies sorted by review stage and methodological shortcut can be found in Supporting Information File S5. The overall RoB is summarized in Figure 3 and the assessment for each study is in Supporting Information File S6.

The main results for each of the reviews and primary studies, sorted by review stage and methodological shortcut are in Supporting Information File S7. The findings are synthesized in Table 1 and also mapped to the AMSTAR 2, ROBIS, and MECIR criteria^{17,18,24} in Supporting Information File S8. When interpreting the impact of the shortcut on the potential increase in speed or RoB of the review (columns 3 and 4 in Table 1) greater weight was given to reviews and primary studies with the lowest RoB.

3.2 | Evidence for shortcuts

Shortcuts that can potentially speed up the review process with no, or minimal, impact on bias are shown in green in Table 2. Shortcuts that can potentially speed up the review process but are likely to increase the potential for bias, so are not recommended, are shown with white background in Table 2. If the shortcuts highlighted in white are used in a rapid review, the increased RoB should be acknowledged.

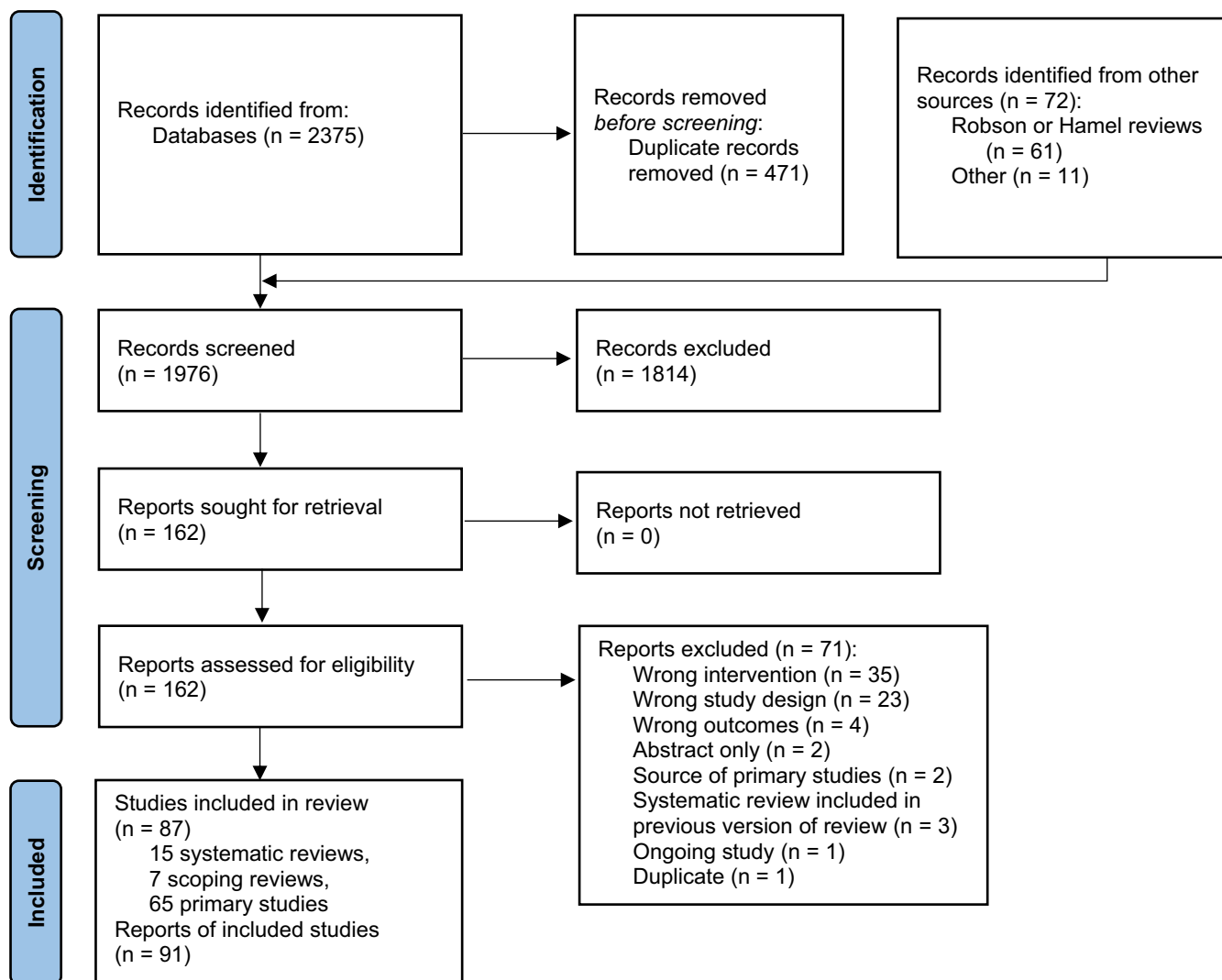


FIGURE 1 PRISMA flow diagram for the systematic review of rapid review methods. [Colour figure can be viewed at wileyonlinelibrary.com]

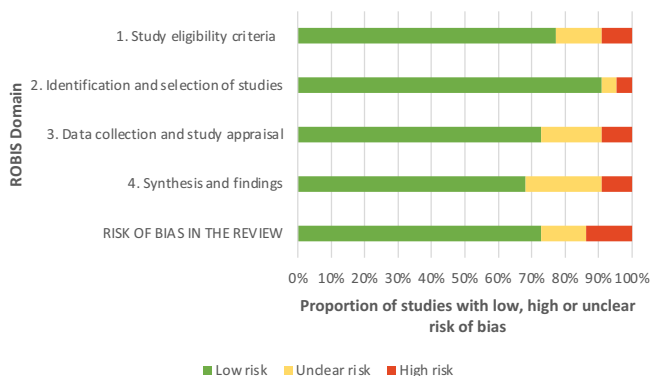


FIGURE 2 Overall risk of bias across the Risk of Bias Assessment Tool for Systematic Reviews domains for the 15 systematic reviews and 7 scoping reviews. [Colour figure can be viewed at wileyonlinelibrary.com]

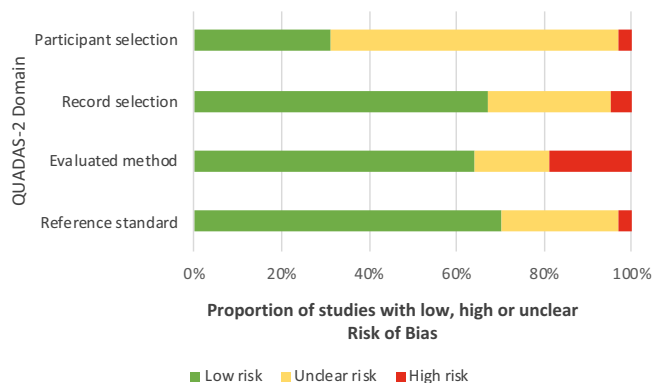


FIGURE 3 Overall risk of bias across the Quality Assessment of Diagnostic Accuracy Studies (QUADAS)-2 domains for the 65 primary studies. [Colour figure can be viewed at wileyonlinelibrary.com]

TABLE 1 Potential impact of shortcuts on speed, bias, and risk of bias assessment tools—by review stage and shortcut tested.

Review stage and shortcut tested	No. SR and PS (Refs.)	↑ speed	↑ bias	Comments
Protocol preparation				
Knowledge brokering to refine the rapid review proposal	1 PS (80)	?	?	Impact on speed or bias not known.
Inclusion criteria				
Limit the publication language to English	1 SR (34), 1 PS (86)	?	Min.	Some studies may be excluded but impact on meta-analyses and conclusions may be minimal. Inclusion of non-English studies may be needed for some research questions. Impact on speed not known.
Exclusion of gray literature	2 SR (27,37)	?	Min.	Likely to overestimate the treatment effects, but only the case in a minority of reviews. Unpublished studies may have smaller sample sizes and may be of lower quality. Impact on speed not known.
Restrictions based on publication date, including last 3, 5, 7, 10, 15, 20, 25, 35, and 40 years	2 PS (76,112)	Yes	Yes	Meta-analyses can lose significance and small chance of changing direction of effect.
Selecting only papers available electronically	1 PS (90)	Yes	Yes	Though the impact on meta-analyses and conclusions was minimal in this study, it may be more significant for reviews conducted with limited access to affluent libraries.
Searching				
Number and combinations of electronic databases, search of reference list of included studies, internet, handsearching, and trial registers	3 SR (25,26,28), 8 PS (61,69,76,87,90,97,107,109)	Yes	Yes	Impact minimal if at least two databases plus reference lists searched. For SRs of RCTs, combining MEDLINE, Embase, and CENTRAL with searches of references lists has the least chance of changing conclusions.
Search strategies				
Methodological filters for observational studies (30) or diagnostic studies	1 SR (30), 1 PS (58)	Yes	Yes	Filters with low sensitivity can lead to missed studies. Could be a good source of time savings if the particular filter is known to have good sensitivity (i.e., 99.5%–100%).
Search techniques to reduce the number of studies retrieved, including focused subject headings, subheadings, frequency operators and title only searches—in Medline, Embase, and PscINFO (all in Ovid)	1 PS (53),	Yes	Yes	Minimal impact, no change in conclusions. Could be a good source of time savings if studies aren't missed. But studies may be missed and evidence limited to one PS.
The use of the Polyglot Search Translator to translate database search strings across multiple databases	1 PS (55)	Yes	No	The PST can increase the speed of translation by 32% without increasing errors. Clear winner, RCT, low RoB.

(Continues)

TABLE 1 (Continued)

Review stage and shortcut tested	No. SR and PS (Refs.)	↑ speed	↑ bias	Comments
Simplified search strategy (SSS) for updates of SRs—involved streamlined Medline searches limited using clinical queries combined with a PubMed “similar articles” search.	1 PS (96)	Yes	Yes	Low sensitivity for topics that were not clinically focused. The reference searches used in this study are not true “gold standards.” <i>Only relevant for review updates.</i>
Peer-review of search strategies	1 PS (102)	No	?	While this strategy may reduce bias by finding more studies it is not a ‘shortcut’, and it is not known if it results in a change in effect or conclusions.
Machine learning RCT classifier within the Cochrane Evidence Pipeline workflow	1 PS (106)	Yes	Min.	The classifier had 99.5% sensitivity, may be acceptable. It is not known if it results in a change in effect or conclusions. <i>Only relevant for SRs of RCTs.</i>
Selection				
Removal of duplicates				
Seven electronic methods (one database, four reference managers, and two review softwares) for flagging and removing duplicate references	1 PS (79)	Yes	?	While the tools varied in their sensitivity and specificity it is not known if any false positives led to missed studies.
Number of reviewers				
Single versus double screening of titles/abstracts	1 SR (39), 7 PS—3 new (63,81,104) and 4 that are also included in the SR (58,59,90,101)	Yes	Yes	Single reviewer screening leads to missed studies and this can impact the conclusions. This is not a good shortcut as the potential for missed studies is too high.
Single versus double screening of full-texts: Complete dual review approach versus limited dual review approach (only in the full-text stage)	1 PS (104)	Yes	Yes	This is not a good shortcut as the potential for missed studies is too high.
Training/experience of reviewers				
English-speaking versus native-language speakers for screening of non-English-language articles for eligibility	1 PS (52)	Yes	Yes	May miss studies. English-speaking reviewers cannot fully replace native-language speakers
Use of nonprofessionals (vs. experts/professionals) to screen titles and abstracts	2 PS (56,83)	?	No	Potential to save time only if more availability of reviewers. Expertise not shown to improve performance and not necessarily faster.
Title/abstract screening				
Two stage method for title and abstract screening (titles then abstracts)	1 PS (78)	No	?	Not likely to speed up process.

TABLE 1 (Continued)

Review stage and shortcut tested	No. SR and PS (Refs.)	↑ speed	↑ bias	Comments
PiCo-based title-only screening using keyword searches based on participants, interventions, and comparators	1 PS (94)	Yes	Yes	May miss studies.
Crowdsourcing of selection				
Crowdsourcing of screening of titles/abstracts and full texts	3 PS (82,84,85)	Yes	?	Potential to save time. 100% sensitivity achieved for RCTs based on title/abstract screening but specific to Cochrane reviews. Mixed results for topic-based selection of title/abstract and full texts.
Dual computer monitors				
Use of dual computer monitors versus single computer monitors for abstract and full-text screening	1 PS (110)	No	No	No benefit.
Machine learning (ML) for screening				
Machine learning approaches to aid title/abstract screening. A range of software and techniques have been tested. Most require training by humans and can be combined with single or double human screening	1 SR (35) 15 PS (50,54,64–67,70,74,88,89,93,103,108,113)	Yes	Yes	Can lead to missed studies. Sensitivity varies. None of the tested approaches have 100% sensitivity. Future potential.
Data extraction				
Number of reviewers				
Single reviewer to extract data with verification by another reviewer versus double data extraction (independent extraction by two reviewers, discussion and consensus)	1 SR (31); 1 PS (51) also included in the SR	Yes	?	More errors but similar pooled estimates. Verification by a second author seems to be acceptable for the characteristics of included studies (MECIR C45) but not for outcomes. Acceptable for ROBIS assessment.
Reviewer experience				
More experienced versus less experienced data abstractors	2 PS (71,73)	No	No	No evidence for time saving or improved accuracy of more experienced abstractors from these two studies.
Use of software/automation for data extraction				
Data Abstraction Assistant to help human reviewers	1 SR (36) 2 PS (72,75)	No	Yes	No evidence yet of time savings or sufficient accuracy. Can impact AMSTAR 2 and ROBIS assessments if used to fully replace human reviewers and duplicate extraction.
Data extraction using the EXACT automatic data extraction tool that extracts results data from ClinicalTrials.gov (http://bio-nlp.org/EXACT)	1 PS (92)	Yes	No	100% accuracy but limited usefulness as only works for ClinicalTrials.gov .

(Continues)

TABLE 1 (Continued)

Review stage and shortcut tested	No. SR and PS (Refs.)	↑ speed	↑ bias	Comments
Dual computer monitors				
Use of dual computer monitors versus single computer monitors for data extraction	1 PS (110)	Yes	No	Dual monitors shown to save time without increasing errors.
Contacting study authors				
Contacting study authors to obtain unique, missing or incomplete information from their studies	3 PS (56,68,100)	No	Min.	Contacting authors has low response and has small impact on the results. Does not save time.
Google Translate				
Use of Google Translate to allow translation and data extraction from non-English trials	1 PS (49)	Yes	Yes	Google Translate can help speed up process but more errors than native speakers.
Risk of bias assessment				
Blinded versus unblinded assessment	1 SR (33)	No	?	No clear advantage of using blinded RoB assessment.
Training/experience of reviewers				
Use of a pedagogical tool (an internet-based computer learning system—ICLS) to improve rate of correct answers (accuracy) when assessing RoB	1 PS (62)	No	No	Training made no difference.
Trained graduate students versus epidemiologist	1 PS (98)	Yes	No	Potential time savings if more availability of reviewers, though training is needed. Experience had minimal impact.

Note: The table only includes shortcuts tested in studies included in this systematic review. ↑, potential increase;?, unknown effect; Min., minimal. The orange shading is a heading for 'Review stage'. The yellow shading is to signify a sub-heading of 'Review stage'.

Abbreviations: PS, primary study; RCT, randomized controlled trial; RoB, risk of bias; SR, systematic review.

Other shortcuts that have future potential to speed up review production include techniques to minimize the number of records found when searching, provided very high sensitivity can be achieved (e.g., close to 100%). These include the use of better methodological search filters, and more precise search strategies. For screening of titles and abstracts, machine learning shows a lot of potential but is not yet sufficiently sensitive and there is a danger that relevant studies will be missed. The use of software to assist or automate data extraction shows future promise but is dependent on the quality and completeness of source data and ease of use. For example, data extraction from ClinicalTrials.gov using the EXACT automatic data extraction tool (<http://bio-nlp.org/EXACT>) was 100% accurate but its usefulness is limited

to only this trials register and to trials with results data included.⁹²

Shortcuts/methods where we were unable to make conclusions due to insufficient evidence of effect include the use of peer-review of search strategies and the most appropriate method for removal of duplicates. Further, there is no evidence to suggest that screening of titles and abstracts is done better by experts, or that data extraction is done better by experienced reviewers. Crowd sourcing of selection has potential but under specific conditions (e.g., to select RCTs based on title/abstract provided the agreement algorithm is robust⁸⁴) however, further work is required for topic-based screening.⁸⁵

In relation to data extraction, single-reviewer extraction clearly increases the RoB. The evidence for data

TABLE 2 Areas where “shortcuts” could be considered to reduce time to completion of rapid reviews without increasing the risk of bias are highlighted in green or gold; shortcuts in white are not recommended.

Possible “shortcuts” categorized by review step	Recommended (yes/no)	Potential impact on risk of bias if done	Supporting evidence from this review
Preparation of a protocol			
• Omit protocol	No	↑	×
• Prepare a protocol and make publicly available (e.g., PROSPERO or Open Science Framework). Limit included information to the key aspects that allow a complete assessment of risk of bias (with AMSTAR 2 or ROBIS).	Yes	↓	×
Question formulation			
• Limit the number of questions and sub-questions and limit the scope of the question/s	Yes	None expected	×
• Limit the number of outcomes	Yes	None expected	×
Inclusion criteria			
• Limit publication language to English but provide justification	Yes	Minimal	✓
• Exclude gray/unpublished literature but provide justification and note that it may lead to an overestimation of effect	Yes	Minimal	✓
• Restrictions based on publication date, for example, last 5, 10, or 20 years (unless it is known that relevant studies could only have been reported since a specific date)	No	↑	✓
• Limit to papers available electronically	No	↑	✓
• Restrict study types to systematic reviews	Yes	None expected	×
• Restrict study types to randomized controlled trials or controlled clinical trials	Yes	None expected	×
Searching			
• Use of a single database for searching (e.g., PubMed)	No	↑	✓
• Limit number of databases searched to at least two and one supplementary source (e.g., reference lists of included studies). If searching for RCTs search Medline/PubMed, EMBASE, and CENTRAL).	Yes	None	✓
• Use the Polyglot Search Translator to translate search strings across multiple databases, though noting its limitations: https://sr-accelerator.com/#/help/polyglot	Yes	None	✓
• Use of the Cochrane RCT Classifier for systematic reviews of RCTs to reduce screening burden (if available for your review) ^a	Yes	None	✓
• Use of existing methodological filters for observational or diagnostic studies	No	↑	✓
Selection			
• One reviewer screens titles and abstracts	No	↑	✓
• Title-only screening using keyword searches based on population, intervention, and comparison (PICO)	No	↑	✓
• One reviewer selects based on full text	No	↑	✓
• Use of dual computer monitors for selection	No	None	✓
Data extraction			
• One reviewer extracts data	No	↑	✓
• One reviewer extracts data with verification by a second reviewer	Yes	Minimal	✓ but evidence is limited
• Data extraction limited to key characteristics, results, conflicts of interest	Yes	None expected	×
• Use of dual computer monitors for data extraction	Yes	None	✓

(Continues)

TABLE 2 (Continued)

Possible “shortcuts” categorized by review step	Recommended (yes/no)	Potential impact on risk of bias if done	Supporting evidence from this review
• Use of Google Translate for data extraction from non-English trials	No	↑	✓
• Not contacting study authors to obtain missing information	Yes	Minimal	✓
Risk of bias assessment			
• Limit or omit risk of bias assessment	No	↑	×
• Not using blinding for risk of bias assessment	Yes	None	✓
• One reviewer does risk of bias assessment	No	?	×
• One reviewer does risk of bias assessment with checking by a second reviewer	?	?	×
• Omit assessment of reporting/publication bias	?	?	×
Synthesis			
• Narrative synthesis only (no meta-analysis)	Yes	None	×
• If data and time permit conduct a meta-analysis of one primary outcome only	Yes	None	×
Reporting			
• Include a methods section, even if added as an appendix	Yes	↓	×
• Limit included information to the key aspects that allow a complete assessment of risk of bias (with AMSTAR 2 or ROBIS)	Yes	None	×
• Consider using a checklist to report the key aspects of the methods rather than narrative reporting	Yes	None	×

Note: Shortcuts highlighted in green or gold can potentially speed up the review process with no, or minimal, impact on bias. Shortcuts in green have supporting evidence included in this review, while those in gold do not. Shortcuts highlighted in white may increase the risk of bias and, if used, this increased risk should be acknowledged. ↑, potential for increased risk of bias; ↓, potential for decreased risk of bias; ×, no evidence to support this recommendation found and included in this systematic review; ✓, evidence to support this recommendation found and included in this systematic review.

^aThis is currently exclusively available for Cochrane Reviews and/or reviews conducted using Covidence software.

extraction conducted by one reviewer with verification by a second reviewer is limited to a single primary study⁵¹ that was also included in a systematic review³¹; the study showed a small increase in errors compared to dual extraction but similar pooled estimates as the reference standard (duplicate data extraction). Given the potential to save time, this shortcut would benefit from further research. An alternative, suggested in the Cochrane Handbook,⁵ of single reviewer extraction with verification by a second reviewer for study characteristics and duplicate data extraction for outcome data was not specifically tested in any included studies. Other alternatives, such as the use of dual extraction for a proportion of studies (e.g., 20%) with discussion, and then single reviewer extraction for the remaining studies have also been suggested but not tested.

There are also a range of shortcuts or techniques that can be used to reduce the time needed to complete the review that are unlikely to increase the RoB, as assessed

by the AMSTAR 2 or ROBIS tools.^{17,18} These were not tested in any of the included studies but are listed in Table 2 and highlighted in gold.

3.3 | Results of primary studies and reviews that assessed rapid review methods in general

Results are presented in Table 3 by theme for the three systematic reviews,^{29,32,38} seven scoping reviews,^{42–48} and six primary studies^{60,77,95,99,105,111} that tested a range of methods—see also Supporting Information File S7 for further details for each review or study, and Supporting Information File S8 for a narrative summary of the results by theme. The results from the systematic review by Veginadu and colleagues³⁸ that examined different methods are not reported here separately due to the 100% overlap with other systematic reviews already reported.^{14,25–28,33–35,37}

TABLE 3 Themes from the primary studies and reviews that assessed rapid review methods in general.

Theme	Main findings	Supporting references
Definition of a rapid review	Eight key themes were identified (in order of frequency of reporting): 1. compare and contrast to a full traditional systematic review 2. variation in methods shortcuts 3. accelerated/rapid process or approach ^a 4. resource efficiency rationale ^a 5. stakeholder rationale 6. systematic approach 7. focus/depth/breadth of scope 8. bias/limitations A definition was proposed based on the four most commonly reported themes	1 ScR (44)
Methods used	A range of methods were reported to have been used in rapid reviews	1 SR (32), 4 ScR (42,43,47,48), 3 PS (77,95,111)
Quality of reporting	The quality of reporting of methods for rapid reviews is poor, but this is not unique to rapid reviews (29)	1 SR (29), 2 ScR (43,48), 1 PS (111)
Time to complete	Estimated median production time is 2 months (IQR 1–6.8 months) for those published in a public repository and 8 months (IQR 2.8–12.5 months) for those published in journals (43)	1 SR (32), 2 ScR (42,43)
Team size	The median rapid review team size is 7 authors (IQR 3–9 authors)	1 ScR (43)
Report length	The median report length is 8 pages (IQR, 6–12 pages; range 2–120 pages)	1 ScR (43)
Resource use	The most resource intensive stages (mostly time) are administration and project management, study selection, data extraction, and critical appraisal, which also vary with the number of included studies, while protocol development, literature search, and study retrieval take less time	1 ScR (46)
Tools for automation	There was a perception that automation tools saved time and increased accuracy (99) but there are limitations with the existing tools (45,99) Further work is required in validation, making them more user-friendly, training, and dissemination (45,99) The cost of the tools was a barrier to 45% of respondents, suggesting potential equity of access issues (99)	1 ScR (45) (99)
Comparison of findings between rapid and systematic reviews	While the conclusions have generally been found to be consistent, the variety in methods used for both rapid and systematic reviews make comparisons difficult. A prospective comparison is needed (95)	2 ScR (42,48), 4 PS (60,95,105,111)

Abbreviations: IQR, interquartile range; PS, primary study; ScR, scoping review; SR, systematic review.

^aEqual third place.

4 | DISCUSSION

Several commonly used shortcuts in rapid reviews⁴⁷ are likely to increase the RoB in the results. These include restrictions based on publication date, and the use of a single reviewer for the main tasks of screening titles and abstracts, selecting studies based on the full-text, and for extracting data from included studies. These shortcuts

have been shown to result in missing relevant studies and to introduce extra errors in data extraction that may impact on the results, including a small chance of changing the conclusions.^{31,39,51,58,59,63,76,81,90,101,104,112} The use of a single electronic database as a source of studies is also not justified by the evidence.^{25,61,69,76,87,90} Thus, authors of rapid reviews should be transparent by reporting their use of these shortcuts and

acknowledging the possibility of them causing bias in the results. Some currently used shortcuts are much less likely to introduce bias or to change the conclusions of the review, including limiting the publication language to English and excluding gray or unpublished literature,^{27,34,37,86} though with some caveats depending on the review question. Other shortcuts that can be used relatively safely to save time and resources include limiting searching to two to three main databases plus one Supplementary source,^{25,26,28,61,69,76,87,90,97,107,109} not contacting study authors for missing information,^{56,68,100} and not using blinding for the RoB assessment.³³

The review has also highlighted areas where the evidence to support methodological choices for both systematic reviews and rapid reviews is lacking or limited. For example, very little attention has been given to testing different methods for data extraction and for RoB assessments and how these can be accelerated. It has also highlighted areas that may have future potential to speed up review production but are not yet sufficiently developed to enable full confidence. These include techniques to minimize the number of records found when searching but with high sensitivity (e.g., close to 100%), such as the use of methodological search filters, and more precise search strategies.

For screening of titles and abstracts, machine learning shows a lot of potential but is not yet sufficiently sensitive to replace human reviewers and there is a danger that relevant studies will be missed. However, there are ways that machine learning can be used to create efficiencies in title and abstract screening without introducing bias, as suggested by Hamel and colleagues in their useful guidance document on the topic.¹¹⁶ Given the rapid developments in machine learning, it may not be too long before it can be used with greater confidence to accelerate the review process.¹¹⁷ There are also ways to use automation tools and an experienced review team to speed up the systematic review process but without increasing the RoB, as demonstrated by Clark and colleagues^{118,119} with their methodology and automation tools for completing full systematic reviews in around 2 weeks. The use of the living systematic methodology is another way to speed up the process.¹²⁰

Three of the most resource-intensive stages (mostly time) during systematic review production are study selection, data extraction, and critical appraisal,⁴⁶ which are most likely influenced by the number of records found when searching and the number of studies included in the review. Thus, a particular area for time efficiencies that may have no impact on bias at all is in the question formulation and setting of the PICO (population, intervention, comparison, and outcomes) aspects of the inclusion criteria. Significant time savings could be

achieved by setting a very focused review question and by limiting the included study designs to the highest level of evidence available—preferably systematic reviews or to RCTs if no systematic reviews are available.

The time used for data extraction could also be reduced by extraction of a much more limited amount of information about study characteristics and results. Sharing and reuse of data extraction and risk of bias assessments from previous systematic reviews could also facilitate faster review production and prevent duplication of effort; a goal towards which Cochrane is moving.¹²¹ Comparison of data extraction and risk of bias assessments for similar systematic reviews could also save time and improve accuracy; as done for the rapid review of COVID-19 therapeutics by the Pan American Health Organization.⁹ These aspects have not been tested in any of the included studies but are unlikely to increase the RoB and may actually reduce the RoB by improving the quality of the data.

Our recommendations for rapid review methods (Table 2) are mostly consistent with those made by the Cochrane rapid review group,⁶ though our recommendations are strengthened by explicitly linking them with the supporting evidence and making it clear when there is no supporting evidence. In contrast to the Cochrane recommendations, we have not made recommendations relating to involvement of an information specialist or peer review of the search strategy as there is no supporting evidence of impact, and not all reviewers have access to information specialists, especially in low- and middle-income countries. The work of the Pan American Health Organization in partnership with the Latin American and Caribbean Center on Health Sciences Information to develop a guided search tool of evidence (EVID@EASY) that uses validated search filters may help to support reviewers without access to information specialists.^{122,123}

The strengths of this systematic review include the use of a pre-registered protocol,²⁰ coverage of all of the systematic review stages, a comprehensive search strategy, dual independent screening of titles/abstracts and full-text to determine inclusion, and the conduct of a RoB assessment of included studies. It updates and extends previous reviews on the topic that were limited to specific systematic review stages¹⁴ or did not include a RoB assessment.¹⁵ Limitations include our reliance on these two previous reviews as a source of prior studies and the lack of time and resources to conduct more supplementary searching, as we had previously planned. While we did not conduct dual, independent data extraction and RoB assessment, we minimized the risk of errors by ensuring that both were checked and verified by a second reviewer, with resolution by discussion and

consensus. We believe that it is unlikely that these limitations will have affected the main findings and conclusions of this review.

5 | CONCLUSION

This review reinforces the findings of the previous reviews, including our own, that there is a limited evidence base for many of the systematic review stages, and that can also be used to inform possible shortcuts for rapid reviews.^{12,14,15,20} However, it does show that there are some shortcuts that are currently widely used for rapid reviews that cannot be justified, such as the use of a single electronic database (e.g., PubMed) as a source of studies, and use of a single reviewer for selection of studies and data extraction. There are also areas where efficiencies can be achieved without increasing bias, such as using a focused review question to reduce the scope of the review. There is also future potential to create efficiencies through the careful use of machine learning to ensure that it does not increase the risk of bias, development of more efficient search strategies, and improved techniques for accurate data extraction, though attention should be given to making these tools equitably accessible.

AUTHOR CONTRIBUTIONS

Michelle M. Haby and Ludovic Reveiz initiated and conceptualized the review. Michelle M. Haby and Ludovic Reveiz developed the methodology with input from Jorge Otávio Maia Barreto, Marcela Torres, and Sasha Peiris. Michelle M. Haby, Jorge Otávio Maia Barreto, Sasha Peiris, Cristián Mansilla, Jenny Yeon Hee Kim, Marcela Torres, and Diego Emmanuel Guerrero-Magaña conducted the selection of studies, data extraction and risk of bias assessment. Michelle M. Haby conducted the analysis with input into the interpretation of the data from all authors. All authors contributed to writing, reviewing, and editing the manuscript.

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CONFLICT OF INTEREST STATEMENT

The authors declare that they have no competing interests. Authors hold sole responsibility for the views expressed in the manuscript, which may not necessarily reflect the opinion or policy of PAHO.

DATA AVAILABILITY STATEMENT

The data that supports the findings of this study are available in the supporting information of this article.

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