

Systematic reviews of diagnostic test accuracy

Karen R Steingart, MD, MPH

karenst@uw.edu

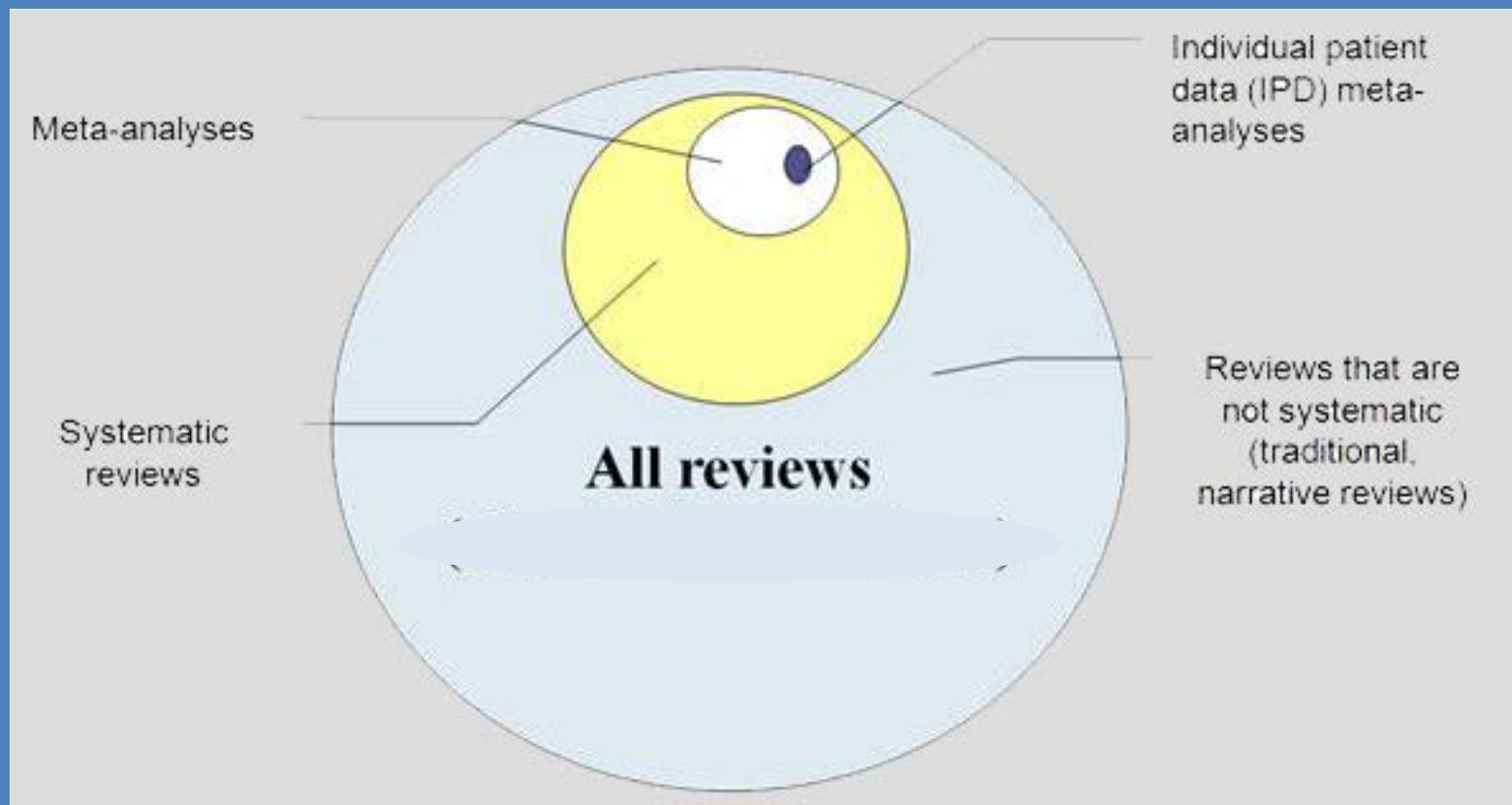
Montreal, July 2012

Conflicts of interest

- I am an Editor with the Cochrane Infectious Diseases Group
- I am a member of the GRADE Working Group
- I have no financial interests to declare

Overview

- Describe key steps in a systematic review/ meta-analysis of diagnostic test accuracy studies
- Demonstrate standard methods of meta-analysis of data from diagnostic studies
- Identify key references and tools for performing meta-analysis of diagnostic studies



A systematic review starts with a clearly formulated question and uses systematic and explicit methods to identify, select, and critically appraise relevant research, and to collect and analyse data from the studies that are included in the review

Egg slide adapted from Madhu Pai

Why systematic reviews?

The Ascent of Evidence
(and the exhaustion of Man)

Winnett



fig.1



fig.2



fig.3



fig.4

Why systematic reviews?

- Scientific summary of all available evidence
- Transparent and reproducible process
- Minimizes bias
- Studies can be formally compared to establish generalizability and consistency
- Heterogeneity can be identified & investigated
- Quantitative part (meta-analyses) may increase the precision of the overall result

Why systematic reviews of diagnostic test accuracy?

- Increasing number of available tests
- Patients ask more questions
- Higher costs and burden on society
- Difficult to choose among tests
- New hypotheses can be generated about particular subgroups

US Institute of Medicine, standards for systematic reviews

<http://www.iom.edu/Reports/2011/Finding-What-Works-in-Health-Care-Standards-for-Systematic-Reviews.aspx>

REPORT BRIEF  MARCH 2011

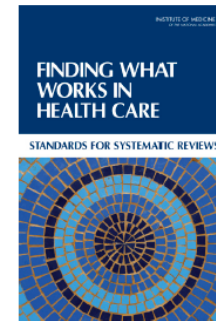
INSTITUTE OF MEDICINE
OF THE NATIONAL ACADEMIES

Advising the nation • Improving health

For more information visit www.iom.edu/srstandards

Finding What Works in Health Care

Standards for Systematic Reviews



Healthcare decision makers in search of reliable information comparing health interventions increasingly turn to systematic reviews for the best summary of the evidence. Systematic reviews identify, select, assess, and synthesize the findings of similar but separate studies and can help clarify what is known and not known about the potential benefits and harms of drugs, devices, and other healthcare services. Systematic reviews can be helpful for clinicians who want to integrate research findings into their daily practices, for patients to make well-informed choices about their own care, and for professional medical societies and other organizations that develop clinical practice guidelines.

In the *Medicare Improvement for Patients and Providers Act of 2008*, Congress directed the Institute of Medicine (IOM) to develop standards for conducting systematic reviews and to develop standards for clinical practice guidelines, which are evidence-based recommendations for clinicians to use when treating patients. The IOM formed two distinct committees to respond to this charge, and each committee assessed the relevant evidence and considered expert guidance to develop the standards. This report, *Finding What Works in Health Care: Standards for Systematic Reviews*, recommends standards for systematic reviews of the comparative effectiveness of medical or surgical interventions (see the insert for a list of the standards).

Systematic reviews ... can help clarify what is known and not known about the potential benefits and harms of drugs, devices, and other healthcare services.

Expectations

- *1,000 to 2,000 person hours to perform a review*
Allen IE, Olkin I. Estimating time to conduct a meta-analysis from number of citations retrieved. JAMA 1999;282:634–635.
- *21 standards and 82 elements of performance*
Institute of Medicine of the National Academies. Finding What Works in Healthcare. Standards for Systematic Reviews. National Academies Press: Washington, DC, 2011.
- *~ 100 elements of performance*
Cochrane performance expectations

Typical systematic reviewer



What is diagnostic test accuracy?

- Diagnosis
 - Does this patient have this disease at this point in time?
- Test accuracy
 - What proportion of those with the disease does the test detect? (sensitivity)
 - What proportion of those without the disease get negative test results? (specificity)
 - Requires 2x2 table of test vs reference standard

Diagnostic accuracy

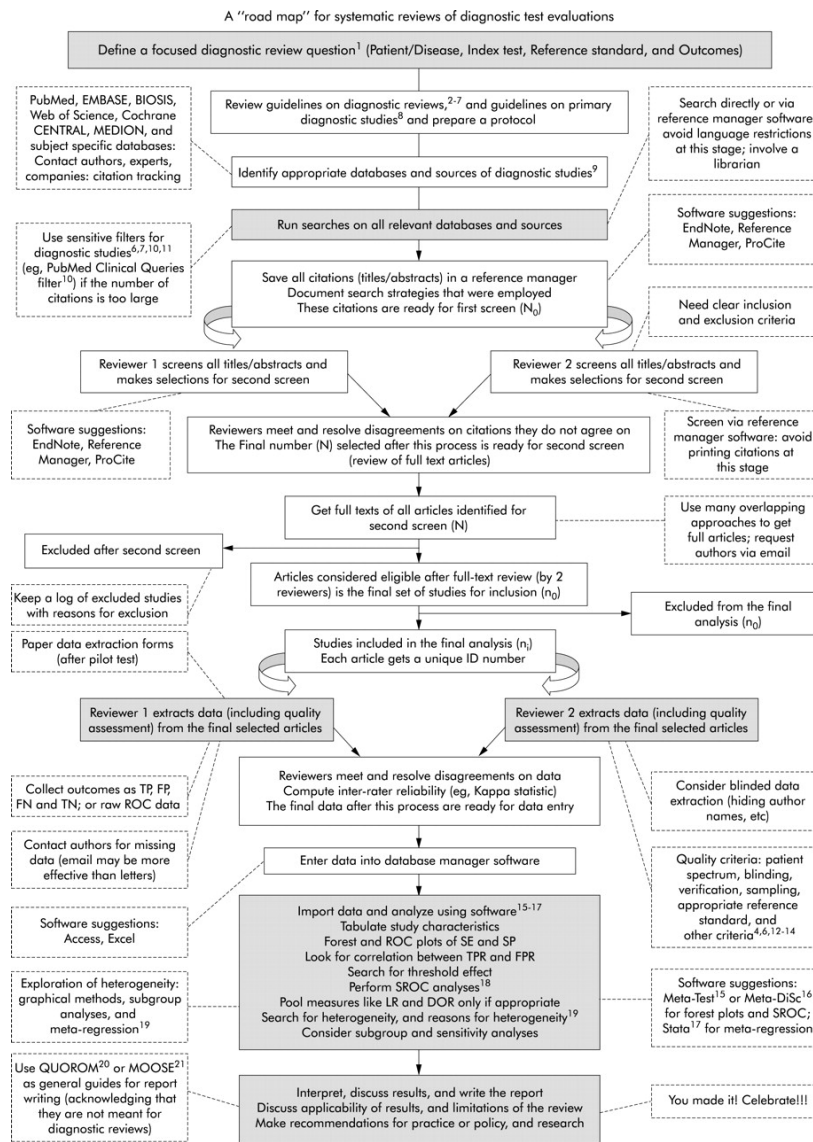
- Agreement between results of the index test and reference standard
- Many measures of agreement
- Focus on pairs of sensitivity & specificity

2x2 Table – sensitivity and specificity

		Disease (Reference test)		
		Present	Absent	
Index test	+	TP	FP	TP+FP
	-	FN	TN	FN+TN
		TP+FN	FP+TN	TP+FP+ FN+TN

sensitivity $TP / (TP+FN)$ specificity $TN / (TN+FP)$

Road map for diagnostic accuracy reviews



Key steps in a diagnostic test accuracy review

1. Framing focused questions
2. Searching for studies
3. Assessing study quality
4. Analyzing the data; undertaking meta-analyses
5. Drawing robust conclusions and informative presentation of results

1. Framing focused questions

Begin with a well-framed question, PICO

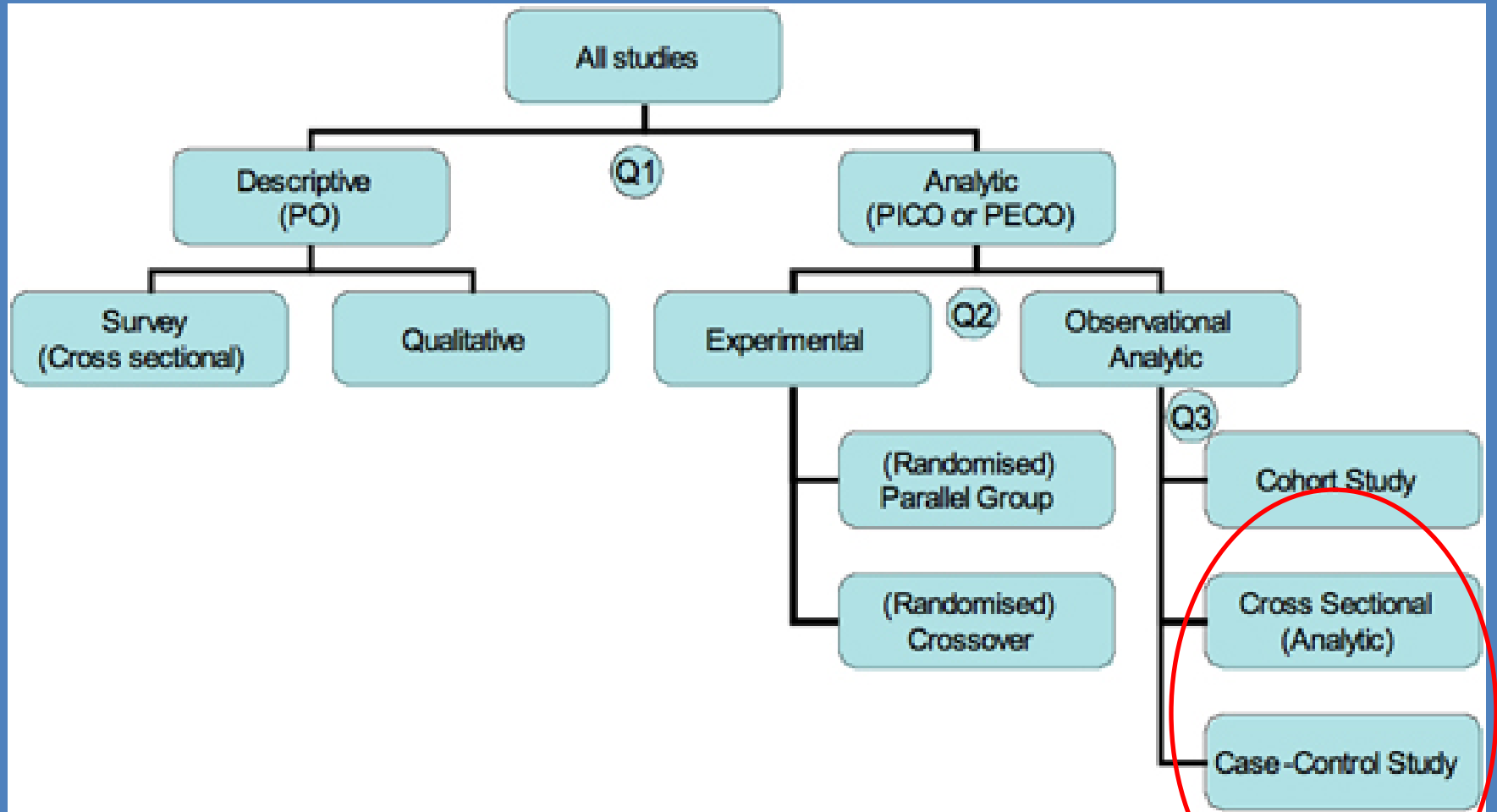
The objectives of the review



Population
Intervention
Comparison
Outcome

- + Study design
- + Purpose of the test/strategy
- + Reference standard

Overview of the study design tree



What is the purpose of the test?

- Triage
 - minimize use of invasive or expensive test
- Add-on
 - improve diagnosis beyond what is already done
- Replacement
 - replace test that is harmful or costly

What is the reference standard?

- ...is usually the best test currently available
- ...used to determine presence/absence of disease
- Indicators of diagnostic accuracy are calculated by comparing results of the index test with those of the reference standard
- Any discrepancy is assumed to arise from error in the index test
- The assumption of 100% accuracy for the reference standard rarely holds true in practice

PICO or PPPICPTR for systematic review of diagnostic test accuracy?

- **P**atients, **P**resentation, **P**rior tests
- **I**ndex test, **C**omparator tests
- **P**urpose: comparative question, role of test
- **T**arget condition, **R**eference standard

Title, objective...

- [Index test] vs [comparator] for [target condition] in [participant description]
 - The GenoType® MTBDRsl test for detecting resistance to second-line antituberculosis drugs and XDR-TB
- Objective: To obtain summary estimates of the diagnostic accuracy of GenoType® MTBDRsl for the detection of resistance to each fluoroquinolone and injectable drug and ethambutol in patient specimens confirmed as TB
- Purpose: Replacement test or an add-on test
- Target condition: Resistance to any of the fluoroquinolones, aminoglycosides and ethambutol
- Reference standard: Conventional drug susceptibility testing by solid or liquid culture +/- DNA sequencing

2. Searching for studies

Recommended standards for finding and assessing individual studies

Standard 3.1 Conduct a comprehensive systematic search for evidence

Required elements:

3.1.1 Work with a librarian or other information specialist trained in performing systematic reviews to plan the search strategy

3.1.2 Design the search strategy to address each key research question

3.1.3 Use an independent librarian or other information specialist to peer review the search strategy

3.1.4 Search bibliographic databases

3.1.5 Search citation indexes

3.1.6 Search literature cited by eligible studies

3.1.7 Update the search at intervals appropriate to the pace of generation of new information for the research question being addressed

3.1.8 Search subject-specific databases if other databases are unlikely to provide all relevant evidence

3.1.9 Search regional bibliographic databases if other databases are unlikely to provide all relevant evidence



Sources of studies for diagnostic accuracy reviews

- MEDLINE, EMBASE, the Cochrane Register of Diagnostic Test Accuracy Studies (under development)
- Search related diagnostic test accuracy reviews (for example HTA database, DARE etc)
- Check references of relevant studies/reviews
- Use a highly sensitive (broad) search strategy
- Use a wide variety of search terms, both text words and database subject headings (MeSH terms)
- Routine use of search filters should generally be avoided

Bossuyt PM, Leeflang MM. *Cochrane Handbook for Systematic Reviews of Diagnostic Test Accuracy Version 0.4 [updated September 2008]*. The Cochrane Collaboration, 2008

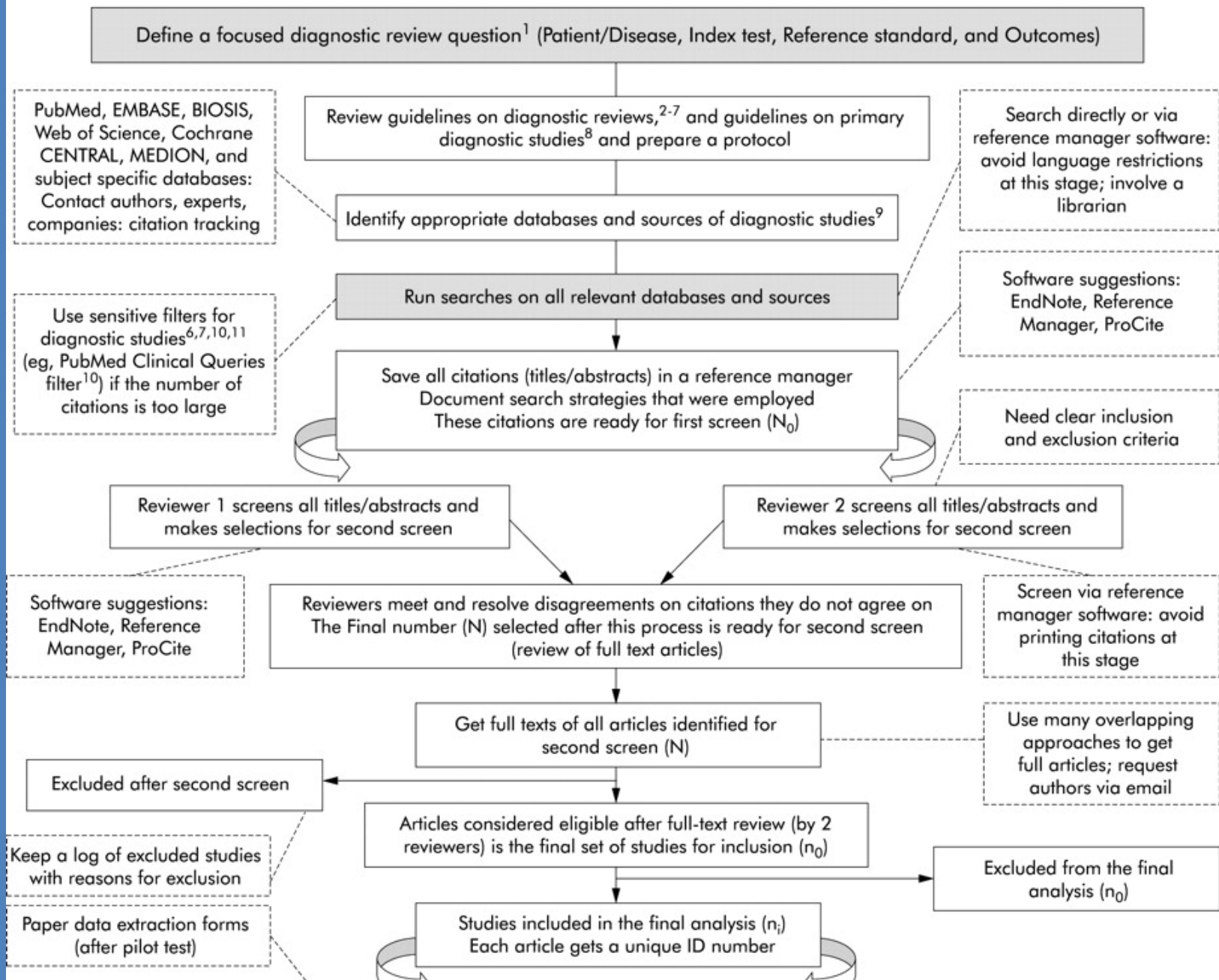
Database searching, example

- We searched the following databases on 25 September 2011 using the search terms and strategy described in Appendix 1: Cochrane Infectious Diseases Group Specialized Register; MEDLINE; EMBASE; ISI Web of Knowledge; MEDION; LILACS; BIOSIS; and SCOPUS. We will also search the metaRegister of Controlled Trials (mRCT) and the search portal of the WHO International Clinical Trials Registry Platform, to identify ongoing trials. We updated the search in MEDLINE on 15 December 2011. Searches were limited to 2007 onward and performed without language restriction.

Search terms, example

- The following search terms were used for MEDLINE searching (through PubMed):
.Xpert*[Text Word] OR GeneXpert**[Text Word]
OR Cepheid*[Text Word] OR near* patient in title
or abstract AND Tuberculosis/ OR exp
Tuberculosis, Pulmonary/ OR exp Tuberculosis,
Multidrug-Resistant/ OR Mycobacterium
tuberculosis/ OR TB[Text Word] OR
tuberculosis[Text Word]

A "road map" for systematic reviews of diagnostic test evaluations



Record judgments and 'reason for exclusion', example using EndNote



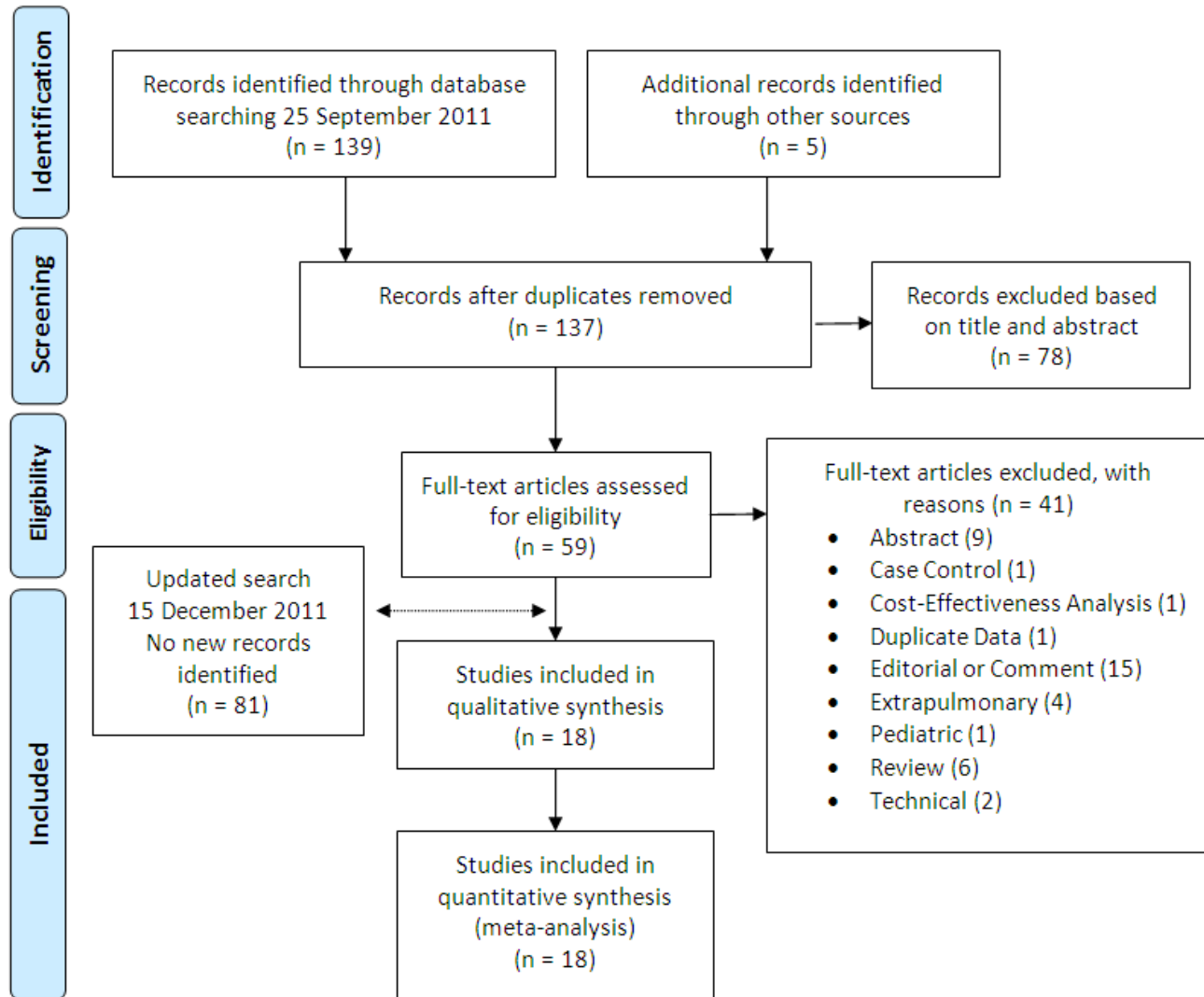
The screenshot shows the EndNote software interface. The main window displays a list of references with columns for Author, Year, Title, Journal/Secondary Title, Custom 2, Reason for exclusion, and Topic. A 'Term Lists' dialog box is open, showing a list of terms including Authors, Journals, Keywords, PDF, Screener1, Screener2, and Reason for exclusion. The 'Reason for exclusion' column in the main window contains various reasons such as 'Editorial and commentary', 'Case control', 'Technical', 'Duplicate data', 'Abstract', 'Review', 'Cost effectiveness', and 'Extrapulmonary'.

Author	Year	Title	Journal/Secondary Title	Custom 2	Reason for exclusion	Topic
	2009	Cepheid unveils fast TB test to aid d...	AIDS Reader	Exclude	Editorial and commentary	
Andersen, A...	2011	[Treatm...		Exclude	Editorial and commentary	
Armand, Syl...	2011	Compar...		Exclude	Case control	
Banada, Pa...	2010	Contain...		Exclude	Technical	
Blakemore, ...	2011	A Multi-...		Exclude	Duplicate data	
Blakemore, ...	2010	Evaluati...		Exclude	Technical	
Bodmer, T; ...	2010	Diagnos...		Exclude	Abstract	
Causse, Ma...	2011	Compar...		Exclude	Extrapulmonary	
Cavusolu, C....	2010	Evaluati...		Exclude	Abstract	
Chee, C. B. E.	2011	Recent...		Exclude	Editorial and commentary	
Chegou, N. ...	2011	Tubercu...		Exclude	Review	
Cuevas, Luis...	2011	The urg...		Exclude	Review	
Deforges, L; ...	2010	Applica...		Exclude	Abstract	
Dowdy, Davi...	2011	Is scale...		Exclude	Cost effectiveness	
Evans, Carl...	2011	GeneXp...		Exclude	Editorial and commentary	
Farga C, Vic...	2011	New cha...		Exclude	Review	
Fenner, L; B...	2011	In reply...		Exclude	Editorial and commentary	
Ferrara, Gio...	2011	Xpert M...		Exclude	Editorial and commentary	
Friedrich, S. ...	2011	Xpert M...		Exclude	Extrapulmonary	
Gotuzzo, E. ...	2011	Xpert M...		Exclude	Editorial and commentary	
Hesseling, A...	2011	Rapid molecular detection of tubercul...	N Engl J Med	Exclude	Editorial and commentary	
Hillemann, D...	2011	Rapid molecular detection of extrapul...	Journal of Clinical Microbi...	Exclude	Extrapulmonary	

How to create term lists linked to Screener and Reason for Exclusion

- 1) In EndNote go to Tools > 'Define Term Lists'
- 2) Click 'Create List' and type 'Screener1'
- 3) Make sure that the Screener1 list is highlighted and click on the 'Terms' tab
- 4) Click on 'New Term' to create the 3 terms in sequence: Include, Exclude, Further review
- 5) Click on the 'Link Lists' button
- 6) Find the custom field you are using as your Screener1 field and link it to the Screener1 list and click "OK"
- 7) Repeat these steps for 'Screener2' and the "Reason for Exclusion" list
- 8) As you begin to type in those fields, EndNote will auto-complete the rest of the text

Xpert MTB/RIF Test



The medical literature can be compared to a jungle. It is fast growing, full of deadwood, sprinkled with hidden treasure and infested with spiders and snakes.
Morgan. *Can Med Assoc J*, 134, Jan 15, 1986

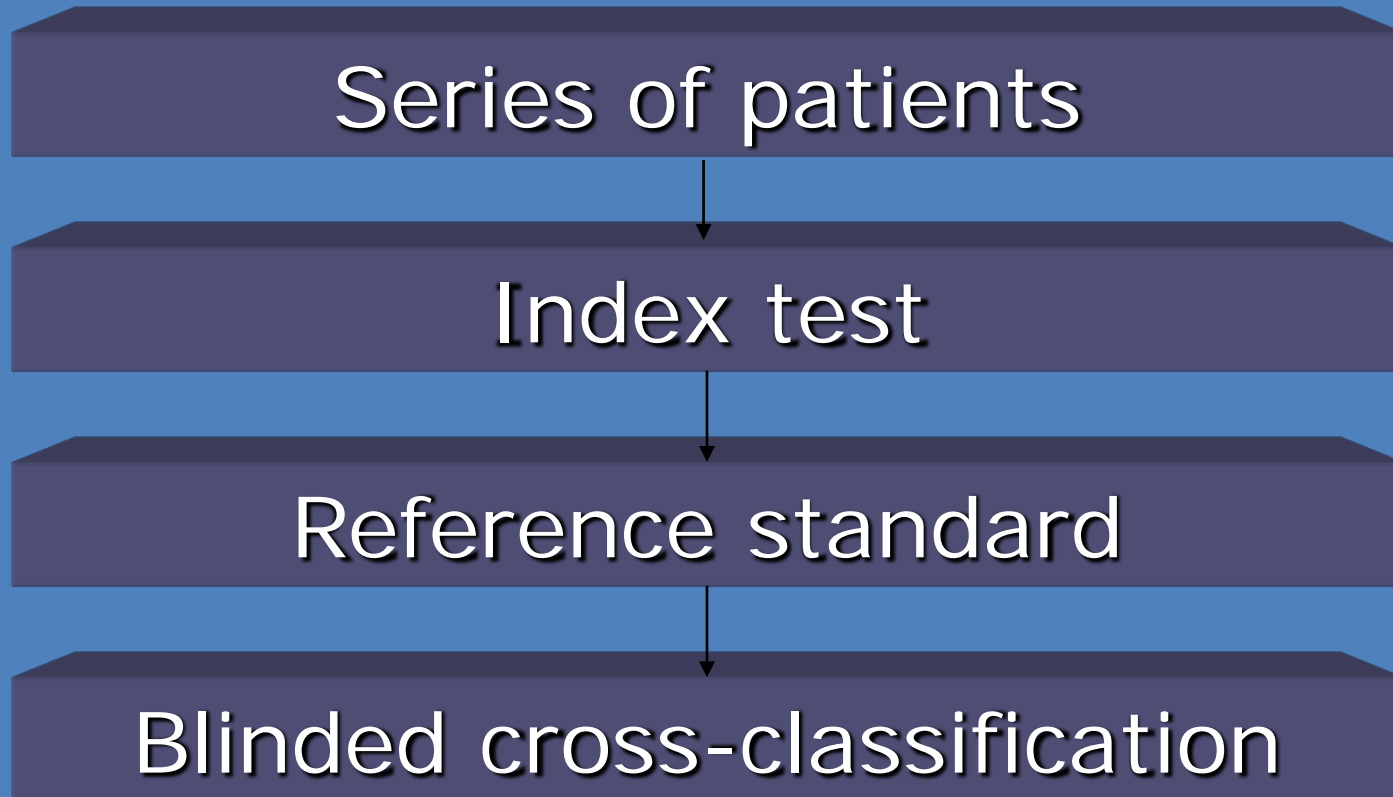


3. Assessing study quality

Quality definition in diagnostic reviews

- Methodological quality of a study: the degree to which the design and conduct of a study match the study objectives
- Two components: risk of bias and applicability

Diagnostic Accuracy Study: Ideal Design



NATURE | COLUMN: WORLD VIEW



Beware the creeping cracks of bias

Evidence is mounting that research is riddled with systematic errors. Left unchecked, this could erode public trust, warns [Daniel Sarewitz](#).

09 May 2012

Alarming cracks are starting to penetrate deep into the scientific edifice. They threaten the status of science and its value to society. And they cannot be blamed on the usual suspects — inadequate funding, misconduct, political interference, an illiterate public. Their cause is bias, and the threat they pose goes to the heart of research.

- [print](#)
- [email](#)
- [download pdf](#)
- [rights and permissions](#)

[Journal home](#)

[Current issue](#)

[For authors](#)

[Subscribe](#)

[E-alert sign up](#)

[RSS feed](#)



[E-alert](#)

[RSS](#)

[Facebook](#)

[Twitter](#)

Enjoy an exclusive
40% discount!
nature

Recent

Read

Commented

Emailed

Nothing will corrode public trust more than a creeping awareness that scientists are unable to live up to the standards that they have set for themselves. Useful steps to deal with this threat may range from reducing the hype from universities and journals about specific projects, to strengthening collaborations between those involved in fundamental research and those who will put the results to use in the real world.

What is bias?

Bias is **any process** at **any stage** of inference tending to produce **results** that **differ systematically** from the **true** values. Murphy EA. *The logic of medicine*, 1976)

Bias is **any trend** in the **collection, analysis, interpretation, publication or review** of data that can lead to **conclusions** that are **systematically different** from the **truth** (Last J. *A dictionary of epidemiology*, 2001)

A Catalogue of Bias, M. Tevfik Dorak (adapted from David Sackett)

<http://www.dorak.info/epi/bc.html>

Literature Review

- Foreign language exclusion bias
- Literature search bias
- One-sided reference bias
- Rhetoric bias

Study Design

- Selection bias
- Sampling frame bias
 - Berkson (admission rate) bias
 - Centripetal bias
 - Diagnostic access bias
 - Diagnostic purity bias
 - Hospital access bias
 - Migrator bias
 - Prevalence-incidence (Neyman / selective survival; attrition) bias
 - Telephone sampling bias
- Nonrandom sampling bias
 - Autopsy series bias
 - Detection bias
 - Diagnostic work-up bias
 - Door-to-door solicitation bias
 - Previous opinion bias
 - Referral filter bias
 - Sampling bias
 - Self-selection bias
 - Unmasking bias
- Noncoverage bias
 - Early-comer bias
 - Illegal immigrant bias
 - Loss to follow-up (attrition) bias
 - Response bias
 - Withdrawal bias
- Noncomparability bias
 - Ecological (aggregation) bias
 - Healthy worker effect (HWE)
 - Lead-time bias
 - Length bias
 - Membership bias
 - Mimicry bias
 - Nonsimultaneous comparison bias
 - Sample size bias

Study Execution

- Bogus control bias
- Contamination bias
- Compliance bias

Data Collection

- Instrument bias
 - Case definition bias
 - Diagnostic vogue bias
 - Forced choice bias
 - Framing bias
 - Insensitive measure bias
 - Juxtaposed scale bias
 - Laboratory data bias
 - Questionnaire bias
 - Scale format bias
 - Sensitive question bias
 - Stage bias
 - Unacceptability bias
 - Underlying/contributing cause of death bias
 - Voluntary reporting bias
- Data source bias
 - Competing death bias
 - Family history bias
 - Hospital discharge bias
 - Spatial bias
- Observer bias
 - Diagnostic suspicion bias
 - Exposure suspicion bias
 - Expectation bias
 - Interviewer bias
 - Therapeutic personality bias
- Subject bias
 - Apprehension bias
 - Attention bias (Hawthorne effect)
 - Culture bias
 - End-aversion bias (end-of-scale or central tendency bias)
 - Faking bad bias
 - Faking good bias
 - Family information bias
 - Interview setting bias
 - Obsequiousness bias
 - Positive satisfaction bias
 - Proxy respondent bias
- Recall bias
 - Reporting bias
 - Response fatigue bias
 - Unacceptable disease bias
 - Unacceptable exposure bias
 - Underlying cause (rumination bias)
 - Yes-saying bias

- Data handling bias
 - Data capture error
 - Data entry bias
 - Data merging error
 - Digit preference bias
 - Record linkage bias

Analysis

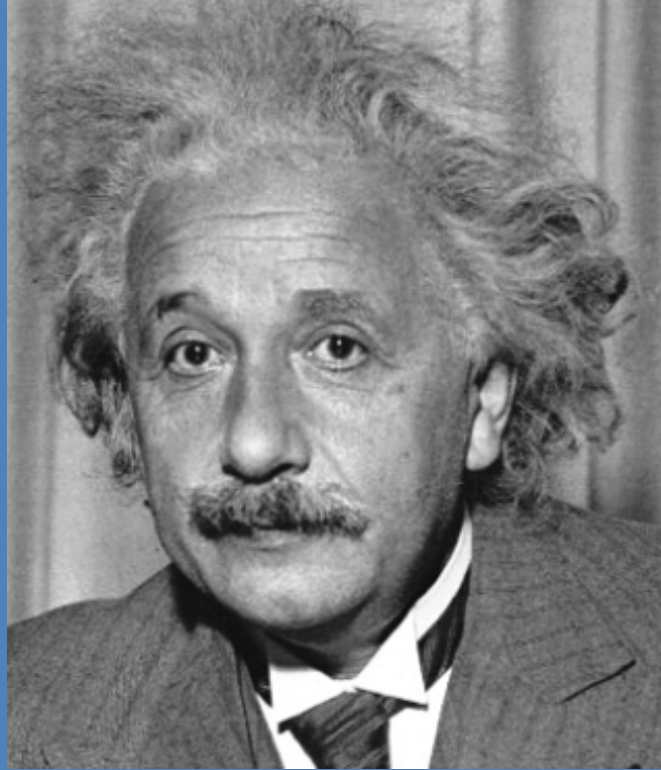
- Confounding bias
 - Latency bias
 - Multiple exposure bias
 - Nonrandom sampling bias
 - Standard population bias
 - Spectrum bias
- Analysis strategy bias
 - Distribution assumption bias
 - Enquiry unit bias
 - Estimator bias
 - Missing data handling bias
 - Outlier handling bias
 - Overmatching bias
 - Scale degradation bias
- Post hoc analysis bias
 - Data dredging bias
 - Post hoc significance bias
 - Repeated peeks bias

Interpretation of Results

- Assumption bias
- Cognitive dissonance bias
- Correlation bias
- Generalization bias
- Magnitude bias
- Significance bias
- Underexhaustion bias

Publication

- All's well literature bias
- Positive result bias
- Hot topic bias



“Everything should be made as simple as possible but not simpler.”

Sources of bias in diagnostic studies: 3 key issues

- Inclusion of right spectrum of patients
- Verification of patients
 - choice of reference standard
 - complete verification
- Independent assessment of index test and reference standard (blinding)

ACADEMIA AND CLINIC

Sources of Variation and Bias in Studies of Diagnostic Accuracy

A Systematic Review

Penny Whiting, MSc; Anne W.S. Rutjes, MSc; Johannes B. Reitsma, MD, PhD; Afina S. Glas, MD, PhD; Patrick M.M. Bossuyt, PhD; and Jos Kleijnen, MD, PhD

Background: Studies of diagnostic accuracy are subject to different sources of bias and variation than studies that evaluate the effectiveness of an intervention. Little is known about the effects of these sources of bias and variation.

Purpose: To summarize the evidence on factors that can lead to

Data Synthesis: The best-documented effects of bias and variation were found for demographic features, disease prevalence and severity, partial verification bias, clinical review bias, and observer and instrument variation. For other sources, such as distorted selection of participants, absent or inappropriate refer-

Empirical Evidence of Design-Related Bias in Studies of Diagnostic Tests

Jeroen C. Lijmer, MD

Ben Willem Mol, MD, PhD

Siem Heisterkamp, PhD

Gouke J. Bonsel, MD, PhD

Martin H. Prins, MD, PhD

Jan H. P. van der Meulen, MD, PhD

Patrick M. M. Bossuyt, PhD

Context The literature contains a large number of potential biases in the evaluation of diagnostic tests. Strict application of appropriate methodological criteria would invalidate the clinical application of most study results.

Objective To empirically determine the quantitative effect of study design shortcomings on estimates of diagnostic accuracy.

Design and Setting Observational study of the methodological features of 184 original studies evaluating 218 diagnostic tests. Meta-analyses on diagnostic tests were identified through a systematic search of the literature using MEDLINE, EMBASE, and DARE databases and the Cochrane Library (1996-1997). Associations between study

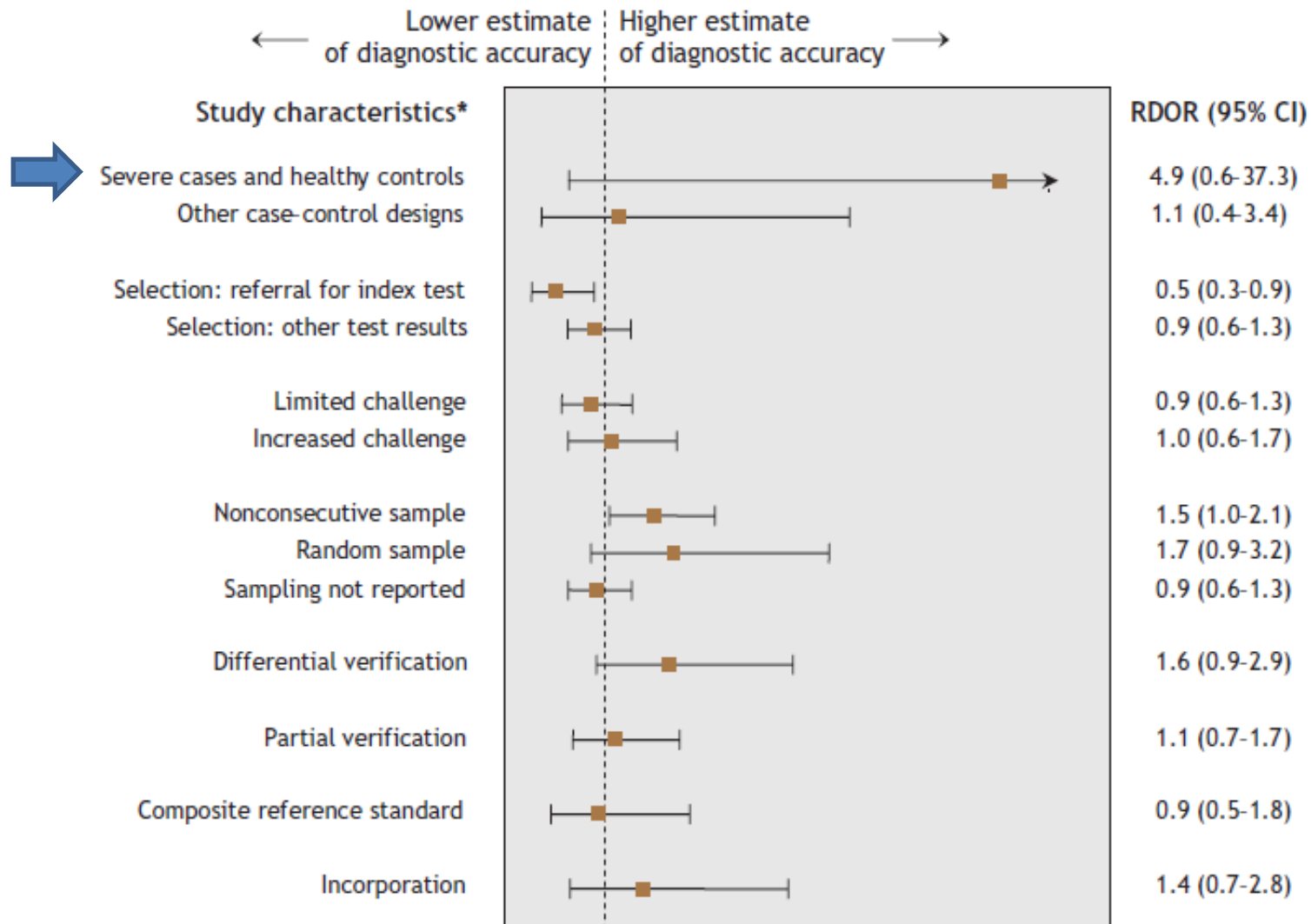
RESEARCH

Evidence of bias and variation in diagnostic accuracy studies

Anne W.S. Rutjes, Johannes B. Reitsma, Marcello Di Nisio, Nynke Smidt, Jeroen C. van Rijn, Patrick M.M. Bossuyt

An abridged version of this article appeared in the Feb. 14, 2006, issue of *CMAJ*.

Effects of study design, A Rutges CMAJ 2006



Diagnostic accuracy of nucleic acid amplification tests for tuberculous meningitis: a systematic review and meta-analysis

Madhukar Pai, Laura L Flores, Nitika Pai, Alan Hubbard, Lee W Riley, and John M Colford Jr

The Lancet Infect Dis 2003

Table 4. Stratified analyses for the evaluation of heterogeneity among studies with in-house tests

Subgroup	Number of studies	Summary diagnostic odds ratio* (95% CI)	Test for heterogeneity† p value
Study design			
Case-control	19	86.5 (39.3, 190.2)	0.03
Cross-sectional	16	43.3 (22.5, 83.3)	0.94
Blinded interpretation of test and/or reference standard results			
Yes	21	46.9 (24.9, 88.6)	0.16
No	14	82.3 (39.8, 170.2)	0.70
Consecutive or random sampling of participants			
Yes	18	63.3 (32.8, 122.4)	0.20
No	17	46.8 (23.6, 92.8)	0.42
Prospective data collection			
Yes	18	59.9 (28.1, 127.6)	0.12
No	17	55.2 (29.9, 101.6)	0.59

*Random effects model. † χ^2 test for heterogeneity. CI=confidence interval.

Case-control studies had a two-fold higher DOR than cross-sectional studies

Research article

Open Access

The development of QUADAS: a tool for the quality assessment of studies of diagnostic accuracy included in systematic reviews

Penny Whiting*¹, Anne WS Rutjes², Johannes B Reitsma²,
Patrick MM Bossuyt² and Jos Kleijnen¹

Address: ¹Centre for Reviews and Dissemination, University of York, England, UK and ²Department of Clinical Epidemiology and Biostatistics, Academic Medical Center, University of Amsterdam, The Netherlands

Email: Penny Whiting* - pfw2@york.ac.uk; Anne WS Rutjes - a.rutjes@amc.uva.nl; Johannes B Reitsma - j.reitsma@amc.uva.nl; Patrick MM Bossuyt - p.m.bossuyt@amc.uva.nl; Jos Kleijnen - jk13@york.ac.uk

* Corresponding author

Published: 10 November 2003

BMC Medical Research Methodology 2003, 3:25

This article is available from: <http://www.biomedcentral.com/1471-2288/3/25>

© 2003 Whiting et al; licensee BioMed Central Ltd. This is an Open Access article: verbatim copying and redistribution of this article is permitted in all media for any purpose, provided this notice is preserved along with the article's original URL.

Received: 14 July 2003

Accepted: 10 November 2003

Annals of Internal Medicine | RESEARCH AND REPORTING METHODS

QUADAS-2: A Revised Tool for the Quality Assessment of Diagnostic Accuracy Studies

Penny F. Whiting, PhD; Anne W.S. Rutjes, PhD; Marie E. Westwood, PhD; Susan Mallett, PhD; Jonathan J. Deeks, PhD; Johannes B. Reitsma, MD, PhD; Mariska M.G. Leeflang, PhD; Jonathan A.C. Sterne, PhD; Patrick M.M. Bossuyt, PhD; and the QUADAS-2 Group*

In 2003, the QUADAS tool for systematic reviews of diagnostic accuracy studies was developed. Experience, anecdotal reports, and feedback suggested areas for improvement; therefore, QUADAS-2 was developed. This tool comprises 4 domains: patient selection, index test, reference standard, and flow and timing. Each domain is assessed in terms of risk of bias, and the first 3 domains are also assessed in terms of concerns regarding applicability. Signalling questions are included to help judge risk of bias.

The QUADAS-2 tool is applied in 4 phases: summarize the review question, tailor the tool and produce review-specific guidance, construct a flow diagram for the primary study, and judge bias and applicability. The tool will allow for more transparent rating of bias and applicability of primary diagnostic accuracy studies.

Ann Intern Med. 2011;155:529-536.

www.annals.org

For author affiliations, see end of text.

* For members of the QUADAS-2 Group, see the Appendix (available at www.annals.org).

Systematic reviews of diagnostic accuracy studies are often characterized by markedly heterogeneous results originating from differences in the design and conduct of included studies. Careful assessment of the quality of included studies is therefore essential. Since its publication in 2003, the QUADAS (Quality Assessment of Diagnostic Accuracy Studies) tool has been widely used (1, 2). More than 200 review abstracts in the Database of Abstracts of Reviews of Effects mention this tool, and it has been cited

Define the Scope

We established a steering group of 9 experts in the area of diagnostic research, most of whom participated in developing the original QUADAS tool. This group agreed on key features of the desired scope of QUADAS-2. The main decision was to separate "quality" into "risk of bias" and "concerns regarding applicability." We defined *quality* as "both the risk of bias and applicability of a study; 1) the degree to which estimates of diagnostic accuracy avoided

QUADAS, 2003

QUADAS-2, 2011

QUADAS ITEM	SCORE		
1. Was the spectrum of patients representative of the patients who will receive the test in practice?	Yes	No	Unclear
2. Is the reference standard likely to correctly classify the target condition?	Yes	No	Unclear
3. Is the time period between reference standard and index test short enough to be reasonably sure that the target condition did not change between the two tests?	Yes	No	Unclear
4. Did the whole sample or a random selection of the sample, receive verification using a reference standard of diagnosis?	Yes	No	Unclear
5. Did patients receive the same reference standard regardless of the index test result?	Yes	No	Unclear
6. Was the reference standard independent of the index test (i.e. the index test did not form part of the reference standard)?	Yes	No	Unclear
7. Were the index test results interpreted without knowledge of the results of the reference standard?	Yes	No	Unclear
8. Were the reference standard results interpreted without knowledge of the results of the index test?	Yes	No	Unclear
9. Were the same clinical data available when test results were interpreted as would be available when the test is used in practice?	Yes	No	Unclear
10. Were uninterpretable/ intermediate test results reported?	Yes	No	Unclear
11. Were withdrawals from the study explained?	Yes	No	Unclear

QUADAS-2

- Domain list
 - patient selection
 - index test
 - reference standard
 - flow and timing
- First 3 domains assessed for applicability
- *Signalling questions* used for judgments of risk of bias

Patient selection - 1

Risk of bias: Could the selection of patients have introduced bias?

- Was a consecutive or random sample of patients enrolled? Yes, No, Unclear
- Was a case control design avoided? Yes, No, Unclear
- *Could the selection of patients have introduced bias?*
Record Low, High, or Unclear

Patient selection - 2

Are there concerns regarding applicability?

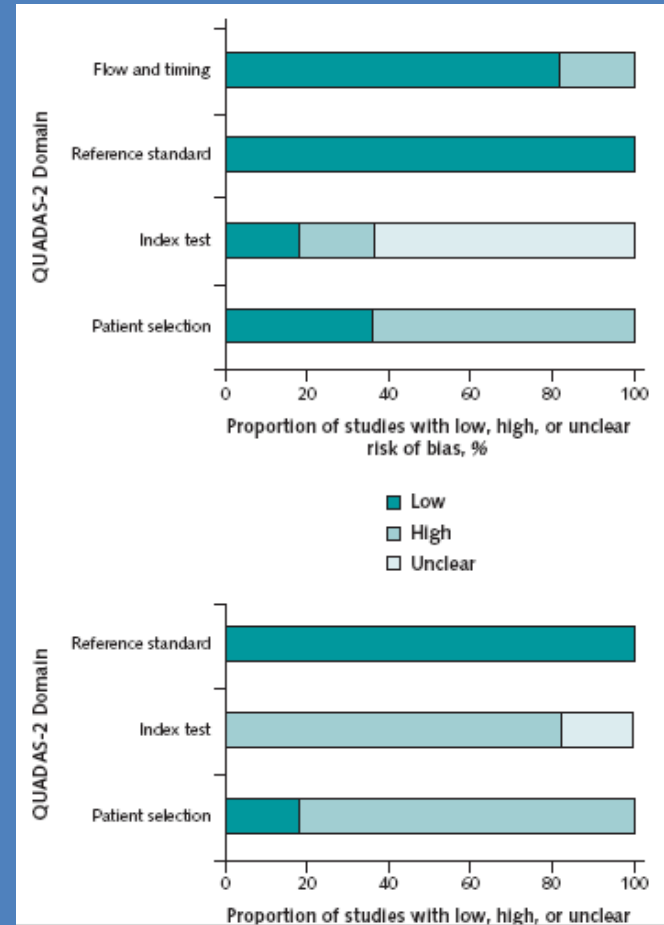
- Do the included patients and setting match the question? (demographic and clinical features, prior testing, intended use of index test, and setting)
- Record Low, High, or Unclear

Suggested displays – QUADAS-2

Table 2. Suggested Tabular Presentation for QUADAS-2 Results

Study	Risk of Bias				Applicability Concerns		
	Patient Selection	Index Test	Reference Standard	Flow and Timing	Patient Selection	Index Test	Reference Standard
1	😊	😊	😊	😊	😊	😊	😊
2	😊	😊	😊	😊	😊	😊	😊
3	😊	😊	😊	😊	😊	😊	😊
4	😊	😊	😊	😊	😊	😊	😊
5	😊	?	😊	😊	😊	😊	😊
6	😊	?	😊	😊	😊	?	😊
7	😊	?	😊	😊	😊	😊	😊
8	😊	?	😊	😊	😊	?	😊
9	😊	?	😊	😊	😊	😊	😊
10	😊	?	😊	😊	😊	😊	😊
11	😊	?	😊	😊	😊	😊	😊

😊 = low risk; 😞 = high risk; ? = unclear risk.



<http://www.bris.ac.uk/quadas/>

Problems with quality assessment

- Not as straightforward as it might seem
- No quality scores or cut-offs for 'good' quality
- Quality assessment is subjective
- Statistical incorporation of quality problematic with limited number of studies
- Hampered by poor reporting

Quality and Reporting of Diagnostic Accuracy Studies in TB, HIV and Malaria: Evaluation Using QUADAS and STARD Standards

Patricia Scolari Fontela¹, Nitika Pant Pai², Ian Schiller², Nandini Dendukuri², Andrew Ramsay³, Madhukar Pai^{1,4*}

1 Department of Epidemiology, Biostatistics and Occupational Health, McGill University, Montreal, Canada, **2** Department of Medicine, Division of Clinical Epidemiology, McGill University, Montreal, Canada, **3** Special Programme for Research and Training in Tropical Diseases, World Health Organization, Geneva, Switzerland, **4** Respiratory Epidemiology and Clinical Research Unit, Montreal Chest Institute, Montreal, Canada

Abstract

Background: Poor methodological quality and reporting are known concerns with diagnostic accuracy studies. In 2003, the QUADAS tool and the STARD standards were published for evaluating the quality and improving the reporting of diagnostic studies, respectively. However, it is unclear whether these tools have been applied to diagnostic studies of infectious diseases. We performed a systematic review on the methodological and reporting quality of diagnostic studies in TB, malaria and HIV.

Methods: We identified diagnostic accuracy studies of commercial tests for TB, malaria and HIV through a systematic search of the literature using PubMed and EMBASE (2004–2006). Original studies that reported sensitivity and specificity data were included. Two reviewers independently extracted data on study characteristics and diagnostic accuracy, and used QUADAS and STARD to evaluate the quality of methods and reporting, respectively.

Findings: Ninety (38%) of 238 articles met inclusion criteria. All studies had design deficiencies. Study quality indicators that were met in less than 25% of the studies included adequate description of withdrawals (6%) and reference test execution (10%), absence of index test review bias (19%) and reference test review bias (24%), and report of uninterpretable results (22%). In terms of quality of reporting, 9 STARD indicators were reported in less than 25% of the studies: methods for calculation and estimates of reproducibility (0%), adverse effects of the diagnostic tests (1%), estimates of diagnostic accuracy between subgroups (10%), distribution of severity of disease/other diagnoses (11%), number of eligible patients who did not participate in the study (14%), blinding of the test readers (16%), and description of the team executing the test and management of indeterminate/outlier results (both 17%). The use of STARD was not explicitly mentioned in any study. Only 22% of 46 journals that published the studies included in this review required authors to use STARD.

Conclusion: Recently published diagnostic accuracy studies on commercial tests for TB, malaria and HIV have moderate to low quality and are poorly reported. The more frequent use of tools such as QUADAS and STARD may be necessary to improve the methodological and reporting quality of future diagnostic accuracy studies in infectious diseases.

Quality of TB accuracy studies using QUADAS

Quality item	45 studies n (%)
Adequate spectrum composition	26 (58)
Adequate reference standard	44 (98)
Absence of disease progression bias	42 (93)
Absence of partial verification bias	44 (98)
Absence of differential verification bias	42 (93)
Absence of incorporation bias	45 (100)
Absence of index test review bias	6 (13)
Absence of reference test review bias	7 (16)
Absence of clinical review bias	14 (31)
Report of uninterpretable results	9 (20)
Description of withdrawals	3 (7)

Initiatives to improve quality and reporting

- STARD: reporting of diagnostic studies
- PRISMA: reporting of systematic reviews/meta-analyses of RCTs
- STROBE: reporting of observational studies
- MOOSE: reporting of meta-analyses of observational studies
- AMSTAR: assessing quality of systematic reviews



www.equator-network.org/

4. Analyzing the data; undertaking meta-analyses

Key steps

- Extract TP, FP, FN, and TN to determine paired estimates of sensitivity and specificity
- Visually examine results of individual studies
- Calculate overall summary estimates using HSROC/bivariate meta-analysis
- Look for and investigate possible reasons for heterogeneity

RevMan[About RevMan 5](#)

- ▶ [Licensing](#)
- ▶ [Download](#)
- ▶ [New Releases](#)
- ▶ [Documentation](#)
- ▶ [Troubleshooting](#)
- ▶ [Other Resources](#)

Other Resources

- ▶ [GRADEpro](#)
- ▶ [MeerKat](#)
- ▶ [RevBase](#)

News from the IMS

- [Pilot to standardise contact details in Archie](#)
- [IMS Bulletin No. 101 published](#)
- [Pilot of structured training and support for non-CRGs](#)
- [Archie 3.3 released](#)
- [Problems checking some reviews out via RevMan](#)

[[News room](#)]

RevMan

Review Manager (RevMan) is the software used for preparing and maintaining Cochrane Reviews.

You can use **RevMan** for protocols and full reviews. It is most useful when you have formulated the question for the review, and allows you to prepare the text, build the tables showing the characteristics of studies and the comparisons in the review, and add study data. It can perform meta-analyses and present the results graphically.

Together with **Archie**, RevMan forms the Cochrane Information Management System (IMS), which is designed to enable contributors to the Cochrane Collaboration to meet the demands of producing high quality systematic reviews of the evidence of the effects of healthcare and deliver these for publication in [The Cochrane Library](#) and elsewhere.

RevMan continues to be developed through an ongoing process of consultation with its users and if you have any suggestions for improvements, please [let us know](#).

RevMan 5 was released on 14 March 2008 (updated to 5.0.25 on 15 September 2010).

**Content available for RevMan**

Among other things, in this section you can find the following information:

- [What's New](#) - List of changes made in updates to RevMan 5
- [Wish List](#) for the next major version
- [FAQ](#) - Check if your question has been already answered
- [Suggestion Form](#) - Send suggestions for improving RevMan
- [Updates](#) - To update RevMan to the latest version
- [Next Release](#) - List of new features for next RevMan release

<http://ims.cochrane.org/revman>

Cut and paste data from excel

Review Manager 5

File Edit Format View Tools Table Window Help

[Commercial PTB_Anda_IgG_sp.rm5] Serological tests for pulmonary TB in adults and children

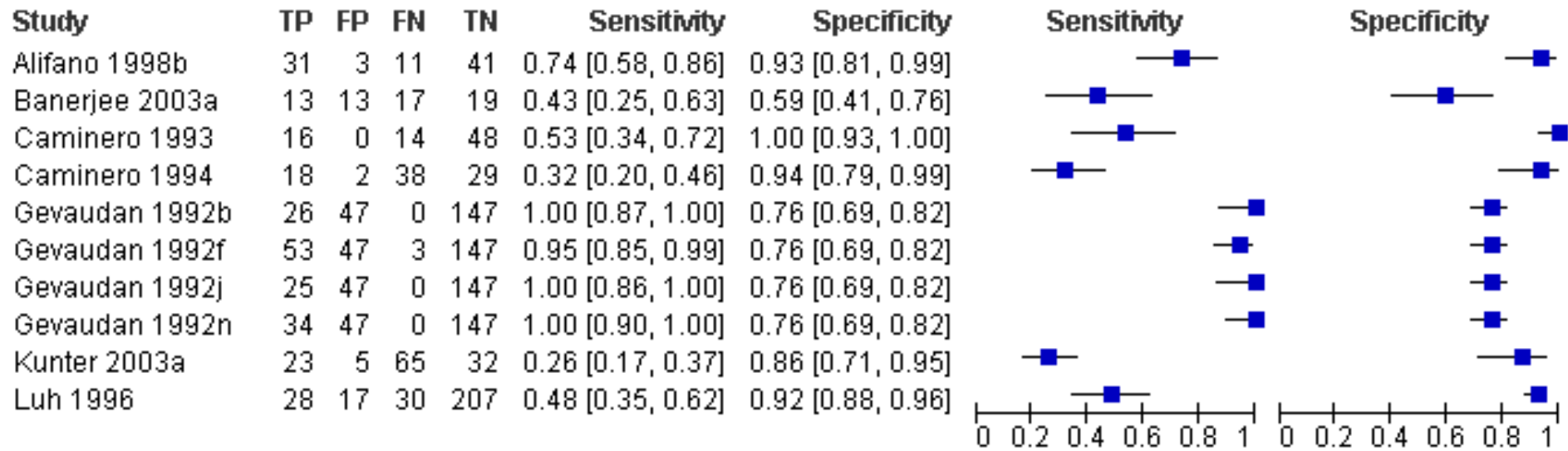
Text of Review **1 Anda-TB IgG**

Test: 1 Anda-TB IgG

Study	TP	FP	FN	TN	Sensitivity
Alifano 1994	35	2	7	92	0.83 [0.69, 0.97]
Alifano 1996 (a)	28	3	5	41	0.85 [0.68, 0.97]
Kalantri 2005 (a)	84	0	21	40	0.80 [0.71, 0.87]
Okuda 2004 (a)	28	10	6	101	0.82 [0.65, 0.97]
Traunmuller 2005	32	21	6	58	0.84 [0.69, 0.97]
Wu 2004 (a)	58	4	34	30	0.63 [0.52, 0.73]
Wu 2005	35	19	30	40	0.54 [0.41, 0.67]

Footnote:

Forest plot – diagnostic test accuracy review



One row is displayed for each study

Extracted data are presented as TP, FP, FN, TN

Data shown in the graph are also displayed numerically

Each study result is given a box for a point estimate

Horizontal line = confidence interval

Software

Open Access

Meta-DiSc: a software for meta-analysis of test accuracy data

Javier Zamora*¹, Victor Abraira¹, Alfonso Muriel¹, Khalid Khan² and
Arri Coomarasamy²

Address: ¹Clinical Biostatistics Unit, Ramón y Cajal Hospital, Madrid, Ctra. Colmenar km 9.100 Madrid 28034, Spain and ²University of Birmingham and Birmingham Women's Hospital, Edgbaston, Birmingham, UK

Email: Javier Zamora* - javier.zamora@hrc.es; Victor Abraira - Victor.abraira@hrc.es; Alfonso Muriel - Alfonso.muriel@hrc.es; Khalid Khan - k.s.khan@bham.ac.uk; Arri Coomarasamy - arricoomar@blueyonder.co.uk

* Corresponding author

Published: 12 July 2006

Received: 31 March 2006

BMC Medical Research Methodology 2006, **6**:31 doi:10.1186/1471-2288-6-31

Accepted: 12 July 2006

This article is available from: <http://www.biomedcentral.com/1471-2288/6/31>

© 2006 Zamora et al; licensee BioMed Central Ltd.

This is an Open Access article distributed under the terms of the Creative Commons Attribution License (<http://creativecommons.org/licenses/by/2.0>), which permits unrestricted use, distribution, and reproduction in any medium, provided the original work is properly cited.

Cut and paste data from excel worksheet into MetaDisc

Meta-DiSc - [Data - Untitled]

File Edit Analyze Window Help

File Edit Undo Redo Sort Asc Sort Desc Print Paste Paste with formatting Bold Italic Underline Bulleted List Numbered List Table

No.	Author	StudyId	Smear	TP	FP	FN	TN
1	Alifano 1994		1	35	2	7	92
2	Alifano 1996	a	1	28	3	5	41
3	Kalantri 2005	a	1	84	0	21	40
4	Okuda 2004	a	1	28	10	6	101
5	Traunmuller 2005		1	32	21	6	58
6	Wu 2004	a	1	58	4	34	30
7	Wu 2005		1	35	19	30	40
8							
9							
10							
11							
12							
13							
14							

Select plot and characteristics

Meta-DiSc - [Plots]

File Edit Analyze Window Help

Sensitivity

Symmetrical SROC Curve Sensitivity / Specificity Show Confidence Intervals
 Asymmetrical SROC Curve Positive LR / Negative L Show current options

	Sensitivity	95% CI
Alifano 1994	0.83	(0.69 - 0.93)
Alifano 1996	0.85	(0.68 - 0.95)
Kalantri 2005	0.80	(0.71 - 0.87)
Okuda 2004	0.82	(0.65 - 0.93)
Traunmuller 2005	0.84	(0.69 - 0.94)
Wu 2004	0.63	(0.52 - 0.73)
Wu 2005	0.54	(0.41 - 0.66)

Pooled Sensitivity = 0.73 (0.69 to 0.78)
Chi-square = 27.47; df = 6 (p = 0.0001)
Inconsistency (I-square) = 78.2%

Options
Redraw
Clear
Export

Zoom
+ -

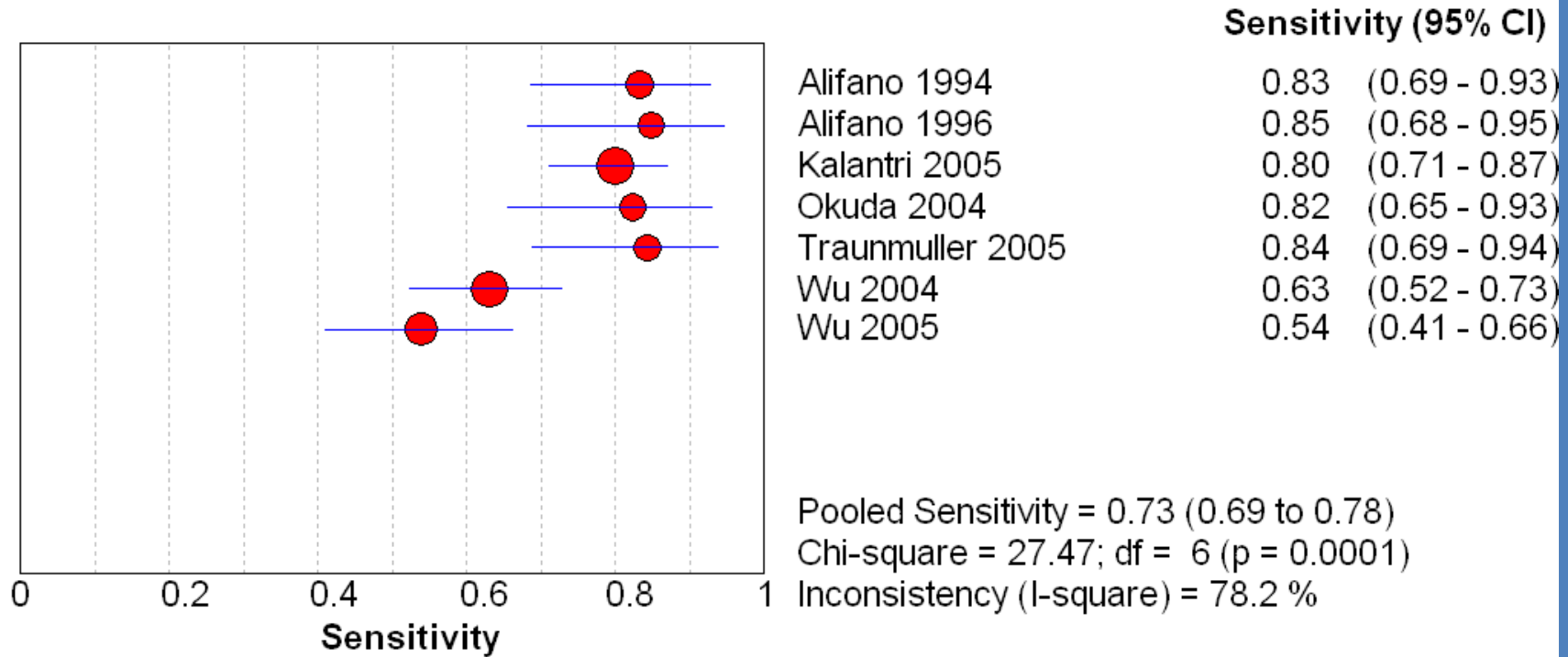
Close

Pooling Symbol
No Symbol

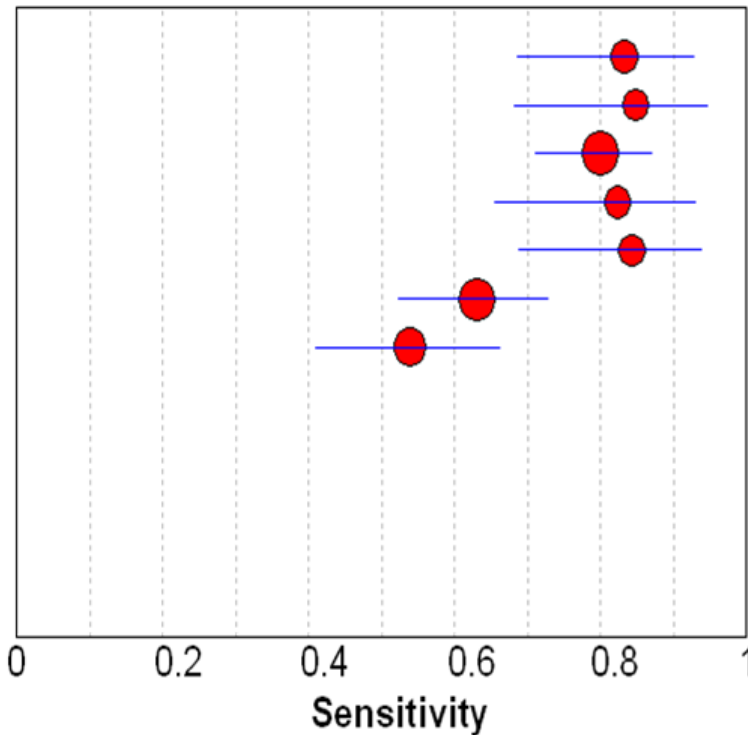
Individual study symbol
Circle (default)

Color selection: [Red]

Export plot



Edit the plot



Sensitivity (95% CI)

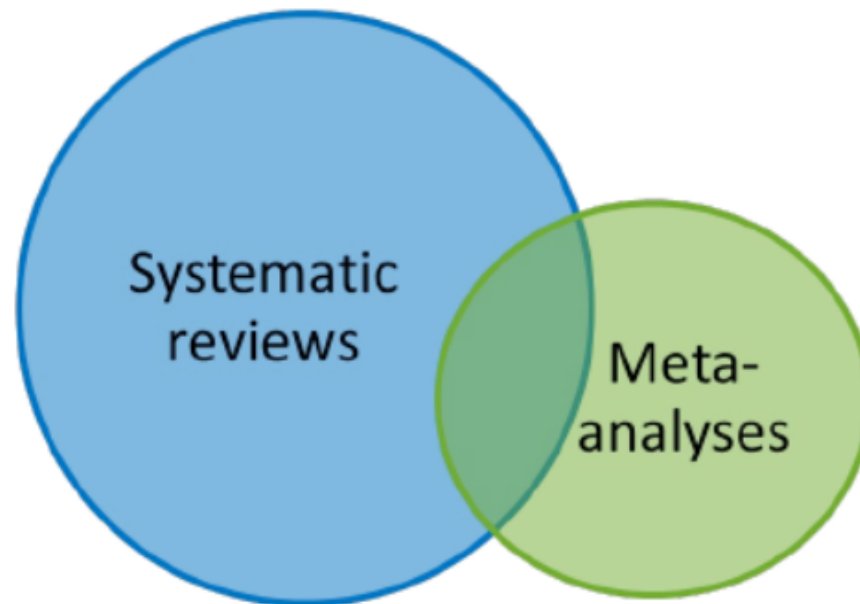
Alifano 1994	0.83	(0.69 - 0.93)
Alifano 1996	0.85	(0.68 - 0.95)
Kalantri 2005	0.80	(0.71 - 0.87)
Okuda 2004	0.82	(0.65 - 0.93)
Traunmuller 2005	0.84	(0.69 - 0.94)
Wu 2004	0.63	(0.52 - 0.73)
Wu 2005	0.54	(0.41 - 0.66)

Pooled Sensitivity = 0.83 (95% CI = 0.75 - 0.91)
I-squared = 27.47, (p = 0.001, 95% CI = 0.00 - 50.00)
Heterogeneity: consistency (I-squared) = 78.2 %



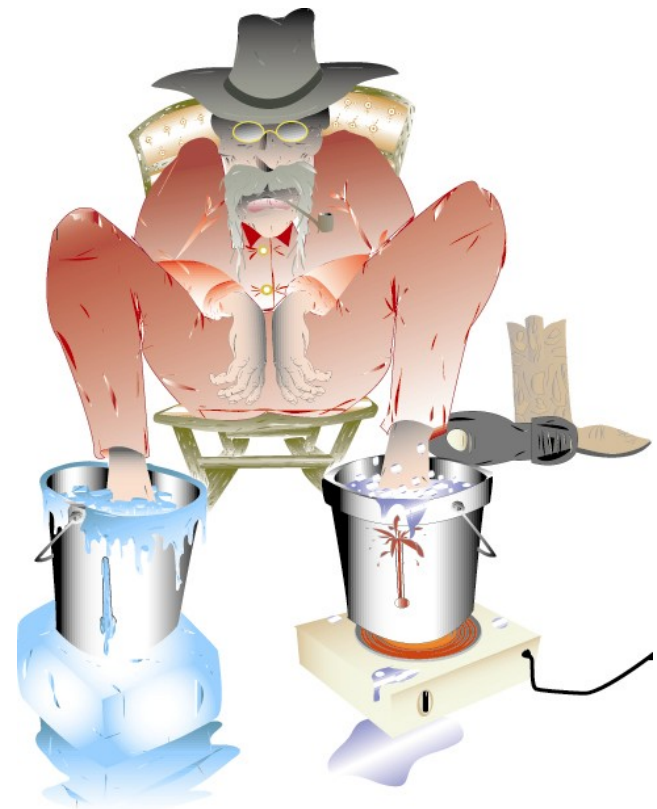
What is a meta-analysis?

- combines the results from two or more studies
- estimates an 'average' or 'common' effect
- optional part of a systematic review



"Il a été dit que un homme avec une jambe dans la glace et l'autre jambe dans l'eau bouillante est à l'aise - en moyenne. JM Yancey

It has been said that a fellow with one leg frozen in ice and the other leg in boiling water is comfortable - on average.



Improved statistical models for meta-analysis of diagnostic studies

- To repair statistical shortcomings
 - hierarchical structure
 - random effects model
- Two different approaches:
 - hierarchical summary ROC model (HSROC, Gatsonis and Rutter 2001)
 - bivariate regression of sensitivity and specificity (Bivariate, Reitsma 2005)

The models are 'hierarchical' because they involve statistical distributions at two levels

- At the lower level, they model the cell counts in the 2x2 tables extracted from each study using binomial distributions and logistic (log-odds) transformations of proportions
- At the second (higher) level, the models assume random study effects to account for heterogeneity in diagnostic test accuracy between studies beyond that accounted for by sampling variability at the lower level

A hierarchical regression approach to meta-analysis of diagnostic test accuracy evaluations

Carolyn M. Rutter^{1,*†} and Constantine A. Gatsonis²

¹*Group Health Cooperative, Center for Health Studies, 1730 Minor Avenue, Suite 1600, Seattle, WA 98101, U.S.A.*

²*Center for Statistical Sciences, Brown University, Box G-H, Providence, RI 02912, U.S.A.*



ELSEVIER

Journal of Clinical Epidemiology 58 (2005) 982–990

**Journal of
Clinical
Epidemiology**

Bivariate analysis of sensitivity and specificity produces informative summary measures in diagnostic reviews

Johannes B. Reitsma^{a,*}, Afina S. Glas^a, Anne W.S. Rutjes^a, Rob J.P.M. Scholten^b,
Patrick M. Bossuyt^a, Aeilko H. Zwinderman^a

^a*Department of Clinical Epidemiology and Biostatistics, Academic Medical Center, University of Amsterdam, PO Box 22700, 1100 DE Amsterdam, The Netherlands*

^b*Dutch Cochrane Centre, Academic Medical Center, University of Amsterdam, The Netherlands*

Accepted 21 February 2005

Biostatistics (2007), **8**, 2, pp. 239–251
doi:10.1093/biostatistics/kxl004
Advance Access publication on May 11, 2006

A unification of models for meta-analysis of diagnostic accuracy studies

ROGER M. HARBORD*

*Medical Research Council Health Services Research Collaboration,
Department of Social Medicine, University of Bristol, Canynge Hall,
Whiteladies Road, Bristol BS8 2PR, UK
roger.harbord@bristol.ac.uk*

JONATHAN J. DEEKS

Centre for Statistics in Medicine, Oxford, UK

MATTHIAS EGGER

Department of Social and Preventive Medicine, University of Berne, Switzerland

PENNY WHITING, JONATHAN A. C. STERNE

*Medical Research Council Health Services Research Collaboration,
Department of Social Medicine, University of Bristol, UK*

Bivariate model vs HSROC model

- Where studies report a common threshold for a positive result, use the bivariate model
- Where studies report several different thresholds, use the HSROC model

Stata command, metandi

The Stata Journal (2009)
9, Number 2, pp. 211–229

metandi: Meta-analysis of diagnostic accuracy using hierarchical logistic regression

Roger M. Harbord
Department of Social Medicine
University of Bristol
Bristol, UK
roger.harbord@bristol.ac.uk

Penny Whiting
Department of Social Medicine
University of Bristol
Bristol, UK

Abstract. Meta-analysis of diagnostic test accuracy presents many challenges. Even in the simplest case, when the data are summarized by a 2×2 table from each study, a statistically rigorous analysis requires hierarchical (multilevel) models that respect the binomial data structure, such as hierarchical logistic regression. We present a Stata package, `metandi`, to facilitate the fitting of such models in Stata. The commands display the results in two alternative parameterizations and produce a customizable plot. `metandi` requires either Stata 10 or above (which has the new command `xtmelogit`), or Stata 8.2 or above with `gllamm` installed.

Keywords: `st0163`, `metandi`, `metandiplot`, diagnosis, meta-analysis, sensitivity and specificity, hierarchical models, generalized mixed models, `gllamm`, `xtmelogit`, receiver operating characteristic (ROC), summary ROC, hierarchical summary ROC

Import/paste data from excel worksheet into Stata data editor

The screenshot displays the Stata 10.1 Special Edition interface. The top window shows the Stata logo and version information, along with copyright details and contact information for StataCorp. Below this, the license information is displayed, indicating a single-user perpetual license for Adithya Cattamanchi at UCSF Department of Medicine.

The Data Editor window is open, showing a table with the following data:

	id	subid	author	year	site	laboratory	typeoftest	drug	tp	fp	fn	tn							
1	3	a	Brossier	2010	France	Indirect	OFL		21	1	3	27							
2	6	a	Hillemann	2009	Germany	Indirect	OFL		29	0	3	74							
3	7	a	Huang	2011	Taiwan	Indirect	OFL		63	0	11	160							
4	9	a	Kiet	2010	Vietnam	Indirect	OFL		31	0	10	21							
5	10	a	Kontsevaya	2011	Russia	Indirect	OFL		25	0	4	19							
6	13	a	Miotto	2012	Italy	Indirect	OFL		42	1	15	116							
7	15	a	Said	2012	NHLS/ Pretoria	Indirect	OFL		26	7	11	292							

Close the data editor, type/paste metandi syntax in lower right 'Command' screen

The screenshot displays the Stata 10.1 interface. The main window shows the Stata logo and version information. The Command window at the bottom right contains the following text:

```
Command
gen sens = tp/(tp +fn)
gen spec = tn/(tn+fp)
label variable sens "Sensitivity"
label variable spec "Specificity"
metandi tp fp tn, nolog
```

A red circle highlights the Command window content. The Variables window on the left lists the following variables:

Name	Label	Type	Format
id	ID	byte	%8.0g
subid	Subid	str1	%s
author	AUTHOR	str10	%10s
year	YEAR	int	%8.0g
site	Site	str14	%14s
typeoftest	Type of test	str8	%s
drug	Drug	str3	%s
tp	TP	byte	%8.0g
fp	FP	byte	%8.0g
fn	FN	byte	%8.0g
tn	TN	int	%8.0g

Hit enter and receive Stata output

```

1 Command
2 edit
3 gen sens = tp/(tp +fn)
4 gen spec = tn/(tn+fp)
5 label variable sens "Sensitivity"
6 label variable spec "Specificity"
7 metandi tp fp fn tn, nolog

```

Name	Label	Type	Format
id	ID	byte	%8.0g
subid	Subid	str1	%9s
author	AUTHOR	str10	%10s
year	YEAR	int	%8.0g
site	Site	str14	%14s
laboratory	Laboratory	str8	%8s
typeoftest	Type of test	str3	%3s
drug	Drug	byte	%8.0g
tp	TP	byte	%8.0g
fp	FP	byte	%8.0g
fn	FN	byte	%8.0g
tn	TN	int	%8.0g
sens	Sensitivity	float	%9.0g
spec	Specificity	float	%9.0g

```

2. (//* option or -set maxvar=) 5000 maximum variables
3. New executable previously downloaded; type -update swap- to install
4. New update available; type -update all-

. edit
(11 vars, 7 obs pasted into editor)

. gen sens = tp/(tp +fn)

. gen spec = tn/(tn+fp)

. label variable sens "Sensitivity"

. label variable spec "specificity"

. metandi tp fp fn tn, nolog

Meta-analysis of diagnostic accuracy

Log likelihood = -24.584852          Number of studies = 7


```

	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
Bivariate						
E(logitSe)	1.440036	.1928134		1.062128	1.817943	
E(logitSp)	4.932998	.7298992		3.502422	6.363574	
Var(logitSe)	.0896943	.1221508		.0062165	1.294136	
Var(logitSp)	.6298585	.9793345		.0299057	13.26577	
Corr(logitSe)	1	.		.	.	
HSROC						
Lambda	5.37453	.597532		4.203389	6.545672	
Theta	-.343075	1.143215		-2.583736	1.897586	
beta	.9745442	.9394819	1.04	0.300	-.8668065	2.815895
s2alpha	.9507444	1.064443		.105942	8.532168	.
s2theta	0
Summary pt.						
Se	.8084602	.0298576		.7430971	.8603191	
Sp	.9928467	.0051839		.9707566	.9982798	
DOR	585.8326	468.7682		122.0854	2811.145	
LR+	113.0187	83.0438		26.77351	477.0848	
LR-	.1929198	.0303402		.1417437	.2625692	
1/LR-	5.1835	.8151999		3.808519	7.054688	

```

covariance between estimates of E(logitSe) & E(logitSp) .0351746

.

```

```

Command

```

Read Stata output

Review

```

1  edit
2  gen sens = tp/(tp +fn)
3  gen spec = tn/(tn+fp)
4  label variable sens "Sensitivity"
5  label variable spec "Specificity"
6  metandf tp fp fn tn, nolog
        
```

Variables

Name	Label	Type	Format
id	ID	byte	%8.0g
subid	Subid	str1	%9s
author	AUTHOR	str10	%10s
year	YEAR	int	%8.0g
site	Site Laboratory	str14	%14s
typeoftest	Type of test	str8	%9s
drug	Drug	str3	%9s
tp	TP	byte	%8.0g
fp	FP	byte	%8.0g
fn	FN	byte	%8.0g
tn	TN	int	%8.0g
sens	Sensitivity	float	%9.0g
spec	Specificity	float	%9.0g

Results

```

2.  (/v# option or -set maxvar-) 5000 maximum variables
3.  New executable previously downloaded; type -update swap- to install
4.  New update available; type -update all-

. edit
(11 vars, 7 obs pasted into editor)

. gen sens = tp/(tp +fn)

. gen spec = tn/(tn+fp)

. label variable sens "Sensitivity"

. label variable spec "Specificity"

. metandf tp fp fn tn, nolog

Meta-analysis of diagnostic accuracy

Log likelihood = -24.584852          Number of studies = 7
        
```

	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]
Bivariate					
E(logitse)	1.440036	.1928134			1.062128 1.817943
E(logitssp)	4.932998	.7298992			3.502422 6.363574
var(logitse)	.0896943	.1221508			.0062165 1.294136
var(logitssp)	.6298585	.9793345			.0299057 13.26577
corr(logits)	1	.			.
HSROC					
Lambda	5.37453	.597532			4.203389 6.545672
Theta	-.343075	1.143215			-2.583736 1.897586
beta	.9745442	.9394819	1.04	0.300	-.8668065 2.815895
s2alpha	.9507444	1.064443			.105942 8.532168
s2theta	0	.			.
Summary pt.					
Se	.8084602	.0298576			.7430971 .8603191
Sp	.9928467	.0051839			.9707566 .9982798
OR	113.0187	83.0438			26.77351 477.0848
LR+	.11929108	.0303402			.1417457 .2625692
LR-	5.1835	.8151999			3.808519 7.054888
1/LR-					
Covariance between estimates of E(logitse) & E(logitssp) .0351746					

Command

Pooled sensitivity = 80.8% (95% CI 74.3, 86.0)
 Pooled specificity = 99.3% (95% CI 97.1, 99.8)

Heterogeneity

- Refers to variation in results among studies
- May be caused by variation in
 - test thresholds (unique to meta-analyses of diagnostic tests)
 - prevalence of disease
 - patient spectrum
 - study quality
 - chance variation

Variation due to threshold differences

- Explicit threshold differences
 - studies have used different cut-off values to define positive test results
- Implicit threshold differences
 - differences in observers
 - differences in equipment
- Consequence: negative correlation arises between sensitivity and specificity

Exploring heterogeneity

- Subgroup analysis
- Meta-regression analysis

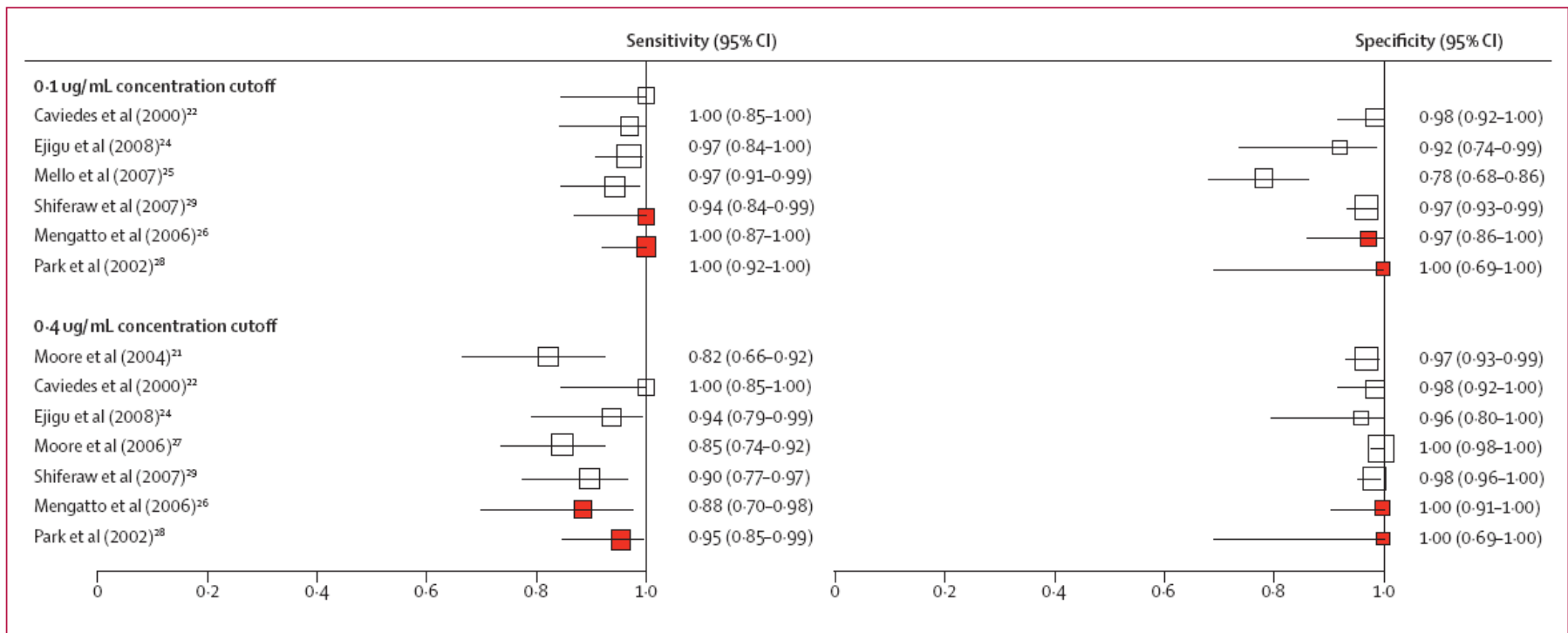


Figure 4: Forest plot of accuracy of the microscopic-observation drug susceptibility assay for detection of isoniazid resistance

Open squares represent studies using direct inoculation with specimens from patients; red squares represent studies using indirect inoculation with isolates. Size of the square is proportionate to the size of the study.

Use of the 0.1 $\mu\text{g}/\text{mL}$ isoniazid cutoff was associated with higher sensitivity, but lower specificity than the 0.4 $\mu\text{g}/\text{mL}$ cutoff. J Minion, TLID 2010

5. Drawing robust conclusions and informative presentation of results
 - summary of findings tables

Some questions

- What are the consequences of using the test in terms of the numbers of TP, FP, FN, and TN?
- How applicable are the results?
- To what extent were the primary studies biased? If serious study limitations were identified, could these impact the results?
- What are the implications for future research?

GRADE Summary of Findings Table for Xpert MTB/RIF Assay

Review question: What is the diagnostic accuracy of Xpert MTB/RIF assay for diagnosis of pulmonary TB and detection of rifampicin resistance?
 Patients/population: Adult pulmonary TB suspects (for diagnosis of pulmonary TB); Confirmed TB cases (for detection of rifampicin resistance)
 Setting: Clinical centers and laboratories
 Index test: Xpert MTB/RIF assay
 Importance: Compared with sputum smear microscopy and conventional drug susceptibility testing, near point-of-care tests, such as Xpert MTB/RIF assay, have considerable advantages for scaling up programmatic management by offering speed of diagnosis, standardized testing, potential for high throughput, and fewer requirements for laboratory bio-safety
 Reference standard: Conventional drug susceptibility testing by solid or liquid culture
 Studies: Cross-sectional or cohort

Outcomes: TP, TN, FP, FN	Effect % (95% CI)	No. of Participants (Studies)	What do these results mean given 5% prevalence among suspects being screened for TB?	What do these results mean given 15% prevalence among suspects being screened for TB?	What do these results mean given 30% prevalence among suspects being screened for TB?	Quality of Evidence
Diagnostic accuracy for diagnosis of pulmonary TB						
All patients	Pooled sensitivity ##.# (95% CI ##.#, ##.#) and pooled specificity ##.##% (95% CI ##.#, ##.#)	#### (18)	With a prevalence of 5%, 50/1000 will have pulmonary TB. Of these, ## (TP) will be identified; ## (FN) will be missed. Of the 950 patients without TB, ## (TN) will not be treated; ## (FP) will be unnecessarily treated	With a prevalence of 15%, 150/1000 will have pulmonary TB. Of these, ## (TP) will be identified; ## (FN) will be missed. Of the 850 patients without TB, ## (TN) will not be treated; ## (FP) will be unnecessarily treated	With a prevalence of 30%, 300/1000 will have pulmonary TB. Of these, ## (TP) will be identified; ## (FN) will be missed. Of the 700 patients without TB, ## (TN) will not be treated; ## (FP) will be unnecessarily treated	Moderate ⊕⊕⊕○
Smear positive patients	Pooled sensitivity ##.##% (95% CI ##.#, ##.#) and pooled specificity ##.##% (95% CI ##.#, ##.#)	#### (##)	With a prevalence of 5%, 50/1000 will have pulmonary TB. Of these, ## (TP) will be identified; ## (FN) will be missed. Of the 950 patients without TB, ## (TN) will not be treated; ## (FP) will be unnecessarily	With a prevalence of 15%, 150/1000 will have pulmonary TB. Of these, ## (TP) will be identified; ## (FN) will be missed. Of the 850 patients without TB, ## (TN) will not be treated; ## (FP) will be unnecessarily	With a prevalence of 30%, 300/1000 will have pulmonary TB. Of these, ## (TP) will be identified; ## (FN) will be missed. Of the 700 patients without TB, ## (TN) will not be treated; ## (FP) will be unnecessarily	Moderate ⊕⊕⊕○

Summary

- Diagnostic tests should be evaluated in consecutively or randomly selected patients with diagnostic uncertainty who are representative of those in whom the test will be used in practice; an appropriate reference standard should be used to establish diagnosis
- Currently, searching electronic databases is challenging. Searches based upon index test and target condition, which are designed to maximize sensitivity, are recommended
- Test accuracy studies are often poorly reported, hampering data extraction and quality assessment
- Though sometimes unable to provide a definitive estimate of test accuracy, systematic reviews can highlight gaps in the evidence base and aid in the design of future studies

References and tools for meta-analysis

- Leeflang. Ann Intern Med. 2008;149:889-897
- Rutter and Gatsonis. Stat Med. 2001; 20:2865–2884
- Reitsma. J Clin Epidemiol. 2005; 982–990
- Zamora. BMC Medical Research Methodology 2006, 6:31
- Cochrane Diagnostic Test Accuracy Working Group
<http://srdta.cochrane.org/>
- <http://www.teachepi.org/> Dr Pai's website for learning and teaching epidemiology
- <http://www.tbevidence.org/> Evidence-based TB diagnosis
- RevMan <http://ims.cochrane.org/revman>
- Meta-analysis in Stata... Ed. Jonathan Sterne 2009

With special thanks to

- Madhu Pai
- Hans Reitsma
- Penny Whiting
- Mariska Leeflang
- Many others



Diagnostic Test Accuracy Review Training,
Montreal, May 2009

Meta-regression

- Is a form of linear regression in which studies are the unit of analysis
- Aims to relate the size of effect to one or more characteristics of the studies involved
- DOR is the dependent variable
- Covariates that might be associated with the variability in DOR are the independent variables
- Tip: Specify covariates that you want to explore in advance

The threshold effect (-0.21) was significant ($p = 0.01$). This was also seen in the SROC plot, Ling 2008.

Table 6. Results from Meta-Regression Analysis Using the Restricted Maximum Likelihood Method

Comparison	Model Coefficient	Relative Diagnostic Odds Ratio (95% CI)	P value
Threshold Effect (S)	-0.21	—	0.01
Retrospective/Both (17) vs Prospective Design (108)	0.13	1.14 (0.56, 2.33)	0.71
Some Convenient Sampling/NR (80) vs Consecutive/Random Sampling (45)	0.38	1.46 (0.87, 2.43)	0.15
No Blinding/NR (105) vs Any Blinding (20)	0.25	1.29 (0.65, 2.58)	0.47
FDA-Approved NAATs (92) vs Not FDA-Approved NAATs (33)	-0.06	0.95 (0.53, 1.68)	0.85
Respiratory Specimens (95) vs Sputum Specimens (30)	0.64	1.89 (1.01, 3.52)	0.05
Culture Reference Standard (105) vs Clinical Reference/Both (20)	0.34	1.40 (0.70, 2.81)	0.34
Resolved Data (37) vs Unresolved Data (88)	-0.05	0.95 (0.54, 1.66)	0.86

doi:10.1371/journal.pone.0001536.t006

Determined using 'Metareg' command in Stata

Diagnostic Odds Ratio (DOR) and Relative DOR

- DOR = odds of a positive result in diseased individuals versus odds of a positive result in non-diseased individuals
- Combines both likelihood ratios $DOR = LR+ / LR-$
- $DOR = 1$ means the test cannot discriminate between people with and without disease
- RDOR (relative DOR) = ratio of 2 DORs
- $RDOR = 1$ means a particular covariate (e.g. blinded study design) does not affect the overall DOR