

# **Systematic reviews of diagnostic test accuracy**

Karen R Steingart, MD, MPH

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# 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 2×2 table of test vs reference standard

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

# Test accuracy may not capture clinical impact

**Table 1| Attributes of the test-treatment pathway that affect patient health**

Pathway component and mechanism	Definition
<b>(1) Diagnostic test delivered</b>	
Timing of test	Speed with which a test is performed within the management strategy
Feasibility	Completion of test process. Reasons for non-completion are: patient acceptability (patient's refusal to have test), test was contraindicated (clinical reason not to administer test), and technical failure (ability of diagnostic equipment to produce data)
Test process	Patients' interaction with test procedure, potentially causing physical or psychological harms or benefits
<b>(2) Test result produced</b>	
Interpretability	Degree to which test data can be used to inform a diagnostic classification
Accuracy	Ability of a test to distinguish between patients who have disease and those who do not
Timing of results	Speed with which test results are available
<b>(3) Diagnosis made</b>	
Timing of diagnosis	Speed with which a diagnostic decision is made
Diagnostic yield	Degree to which the test contributes to a patient diagnosis in any form, including: provision of a definitive diagnosis, confirmation of a suspected diagnosis, ruling out a working diagnosis, and distinguishing between alternative diagnoses with different treatment implications. Diagnostic yield is different from accuracy because it also incorporates any other information used by a doctor to make a diagnosis (such as previous test results)
Diagnostic confidence	Degree of confidence that doctors and patients have in the validity or applicability of a test result
<b>(4) Management decided</b>	
Therapeutic yield	Degree to which diagnostic decisions affect treatment plans
Therapeutic confidence	Certainty with which doctors and patients pursue a course of treatment
<b>(5) Treatment implemented</b>	
Timing of treatment	Speed with which patients receive treatment
Treatment efficacy	Ability of the treatment intervention to improve patient outcomes
Adherence	Extent to which patients participate in the management plan, as advised by their doctor, to attain therapeutic goal



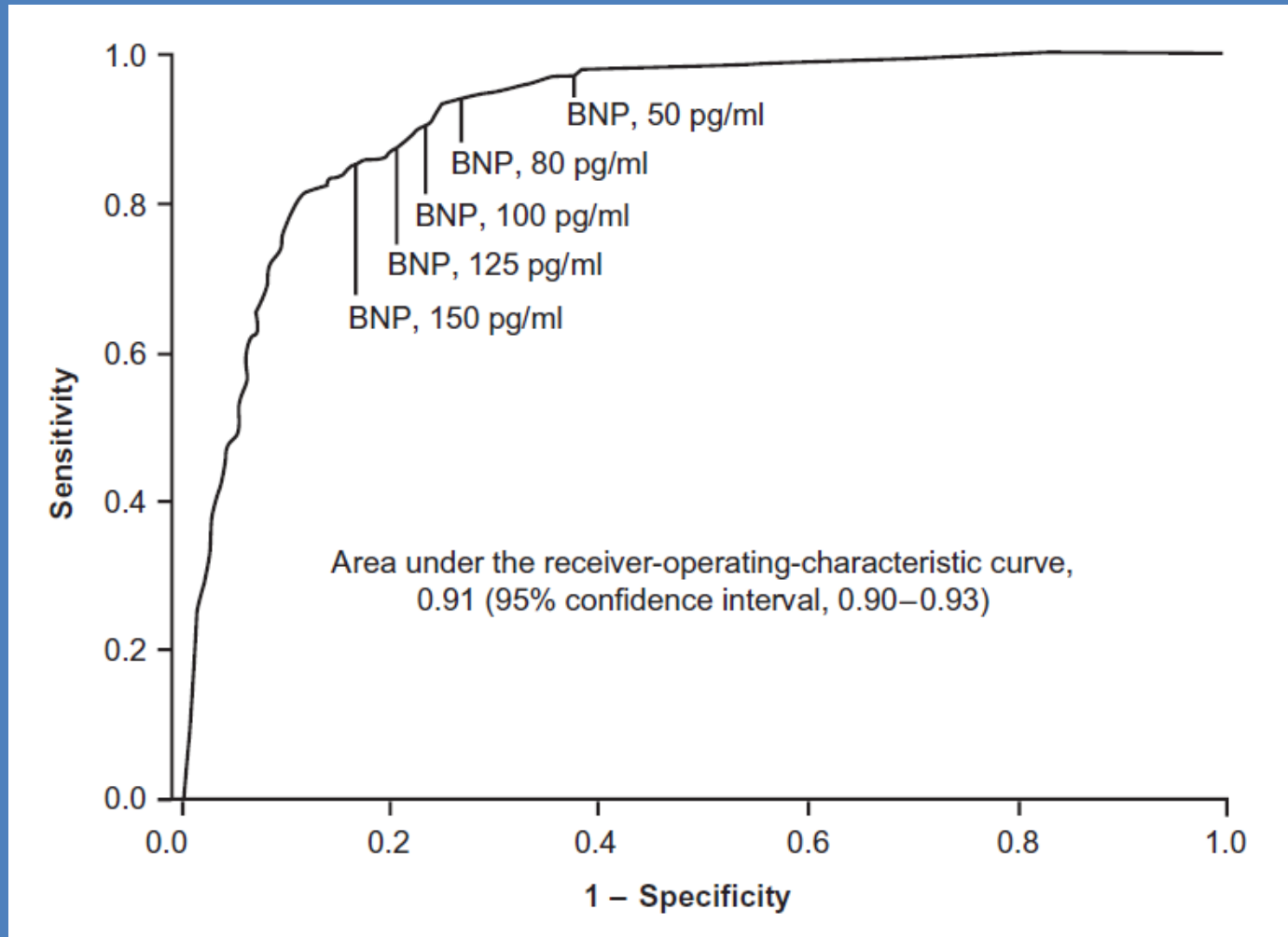
# Clinical impact of test results on diagnostic and treatment decisions, and eventually, patient outcomes



“Improved accuracy is not always a necessary prerequisite for improving patient health, nor does it guarantee other downstream improvements”

[di Ruffano et al. *BMJ* 2012;344:e686]

Accuracy vs Impact:  
Rapid measurement of B-type natriuretic peptide  
in the emergency diagnosis of heart failure



# B-Type Natriuretic Peptide Testing, Clinical Outcomes, and Health Services Use in Emergency Department Patients With Dyspnea

## A Randomized Trial

Hans-Gerhard Schneider, MBBS, MD; Louisa Lam, MPH; Amaali Lokuge, MBBS; Henry Krum, MBBS, PhD; Matthew T. Naughton, MBBS; Pieter De Villiers Smit, MBBS; Adam Bystrycki, MBBS; David Eccleston, MBBS, PhD; Jacob Federman, MBBS; Genevieve Flannery, MBBS; and Peter Cameron, MBBS, MD

**Background:** B-type natriuretic peptide (BNP) is used to diagnose heart failure, but the effects of using the test on all dyspneic patients is uncertain.

**Objective:** To assess whether BNP testing alters clinical outcomes and health services use of acutely dyspneic patients.

**Design:** Randomized, single-blind study. Patients were assigned to a treatment group through randomized numbers in a sealed envelope. Patients were blinded to the intervention, but clinicians and those who assessed trial outcomes were not.

**Setting:** 2 Australian teaching hospital emergency departments.

**Patients:** 612 consecutive patients who presented with acute severe dyspnea from August 2005 to March 2007.

**Intervention:** BNP testing ( $n = 306$ ) or no testing ( $n = 306$ ).

**Measurements:** Admission rates, length of stay, and emergency department medications (primary outcomes); mortality and readmission rates (secondary outcomes).

**Results:** There were no between-group differences in hospital admission rates (85.6% [BNP group] vs. 86.6% [control group]; dif-

ference,  $-1.0$  percentage point [95% CI,  $-6.5$  to  $4.5$  percentage points];  $P = 0.73$ ), length of admission (median, 4.4 days [interquartile range, 2 to 9 days] vs. 5.0 days [interquartile range, 2 to 9 days];  $P = 0.94$ ), or management of patients in the emergency department. Test discrimination was good (area under the receiver-operating characteristic curve, 0.87 [CI, 0.83 to 0.91]). Adverse events were not measured.

**Limitation:** Most patients were very short of breath and required hospitalization; the findings might not apply for evaluating patients with milder degrees of breathlessness.

**Conclusion:** Measurement of BNP in all emergency department patients with severe shortness of breath had no apparent effects on clinical outcomes or use of health services. The findings do not support routine use of BNP testing in all severely dyspneic patients in the emergency department.

**Primary Funding Source:** Janssen-Cilag.

*Ann Intern Med.* 2009;150:365-371.

For author affiliations, see end of text.

ClinicalTrials.gov registration number: NCT00163709.

[www.annals.org](http://www.annals.org)



# Evaluation of the PIMA Point-of-Care CD4 Analyzer in VCT Clinics in Zimbabwe

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**Abstract:** Point-of-care (POC) CD4 testing was implemented at a stand-alone HIV voluntary testing and counseling centre in Harare, Zimbabwe. To validate the use of this new technology, paired blood samples were collected from 165 patients either by a nurse or a laboratory technician and tested using POC and conventional laboratory CD4 machines. Finger prick (capillary) blood was collected directly into the PIMA POC CD4 Analyzer cartridges and tested immediately, whereas venous blood collected into evacuated tubes was used for CD4 enumeration on a Becton Dickinson FACSCalibur. There was no significant difference in mean absolute CD4 counts between the POC PIMA and Becton Dickinson FACSCalibur platforms ( $+7.6$  cells/ $\mu$ L;  $P = 0.72$ ). Additionally, there was no significant difference in CD4 counts between the platforms when run by either a nurse ( $+18.0$  cells/ $\mu$ L;  $P = 0.49$ ), or a laboratory technicians ( $-3.1$  cells/ $\mu$ L;  $P = 0.93$ ). This study demonstrates that POC CD4 testing can be conducted in a voluntary testing and counseling setting for staging HIV-positive clients. Both nurses and laboratory technicians performed the test accurately, thereby increasing the human resources available for POC CD4 testing. By producing same-day results, POC CD4 facilitates immediate decision-making, patient management and referral and may help improve patient care and retention. POC CD4 may also alleviate testing burdens at traditional central CD4 laboratories, hence improving test access in both rural and urban environments.

**Key Words:** CD4, HIV, diagnosis, client-initiated testing, laboratory, PIMA, point-of-care, voluntary counseling and testing, VCT

(*J Acquir Immune Defic Syndr* 2010;55:1–7)

## BACKGROUND

CD4 T-lymphocyte count is an important qualifying test for antiretroviral treatment (ART) in HIV-positive individuals and is also used to monitor treatment efficacy.<sup>1–7</sup> The scale up of public ART programs globally<sup>8</sup> has led to an increased demand for CD4 count tests, especially to assess treatment eligibility. Despite expansion of laboratory infrastructure and services, access to CD4 testing remains a bottleneck to ART scale-up. In Zimbabwe, an estimated 380,000 adults are in need of ART<sup>9</sup> and, by the end of 2009, an estimated 215,000 were on ART within the public sector.<sup>10</sup> There is clearly a need to increase access to ART services and improving CD4 access may help.

In Zimbabwe, the “New Start” voluntary testing and counseling (VCT) centers (also known as client-initiated testing and counselling centers) are established by the Ministry of Health and Child Welfare in partnership with Population Services International (PSI) and provide free rapid HIV testing services to more than 360,000 clients nationwide on an annual basis. Clients testing positive at VCT centers are then referred to Opportunistic Infection (OI) clinics for HIV care and ART if eligible. After enrollment at the OI clinics, patients are scheduled for a CD4 count test. Due to high demand, delays in CD4 testing can occur for 2–3 weeks on average. There is substantial loss-to-follow-up of patients between HIV diagnosis and registration at the OI clinics and delays in CD4 testing can result in further loss of patients who do not return or who die before initiating treatment. The situation is exacerbated in rural areas where more limited CD4 access creates a significant bottleneck to the scale up of ART.



## Effect of point-of-care CD4 cell count tests on retention of patients and rates of antiretroviral therapy initiation in primary health clinics: an observational cohort study



Ilesh V Jani, Nádia E Sítos, Eunice R Alfai, Patrícia L Chongo, Jorge I Quevedo, Beatriz M Rocha, Jonathan D Lehe, Trevor F Peter

**Background** Loss to follow-up of HIV-positive patients before initiation of antiretroviral therapy can exceed 50% in low-income settings and is a challenge to the scale-up of treatment. We implemented point-of-care counting of CD4 cells in Mozambique and assessed the effect on loss to follow-up before immunological staging and treatment initiation.

**Methods** In this observational cohort study, data for enrolment into HIV management and initiation of antiretroviral therapy were extracted retrospectively from patients' records at four primary health clinics providing HIV treatment and point-of-care CD4 services. Loss to follow-up and the duration of each preparatory step before treatment initiation were measured and compared with baseline data from before the introduction of point-of-care CD4 testing.

**Findings** After the introduction of point-of-care CD4 the proportion of patients lost to follow-up before completion of CD4 staging dropped from 57% (278 of 492) to 21% (92 of 437) (adjusted odds ratio [OR] 0.2, 95% CI 0.15–0.27). Total loss to follow-up before initiation of antiretroviral treatment fell from 64% (314 of 492) to 33% (142 of 437) (OR 0.27, 95% CI 0.21–0.36) and the proportion of enrolled patients initiating antiretroviral therapy increased from 12% (57 of 492) to 22% (94 of 437) (OR 2.05, 95% CI 1.42–2.96). The median time from enrolment to antiretroviral therapy initiation reduced from 48 days to 20 days ( $p < 0.0001$ ), primarily because of a reduction in the median time taken to complete CD4 staging, which decreased from 32 days to 3 days ( $p < 0.0001$ ). Loss to follow-up between staging and antiretroviral therapy initiation did not change significantly (OR 0.84, 95% CI 0.49–1.45).

**Interpretation** Point-of-care CD4 testing enabled clinics to stage patients rapidly on-site after enrolment, which reduced opportunities for pretreatment loss to follow-up. As a result, more patients were identified as eligible for and initiated antiretroviral treatment. Point-of-care testing might therefore be an effective intervention to reduce pretreatment loss to follow-up.

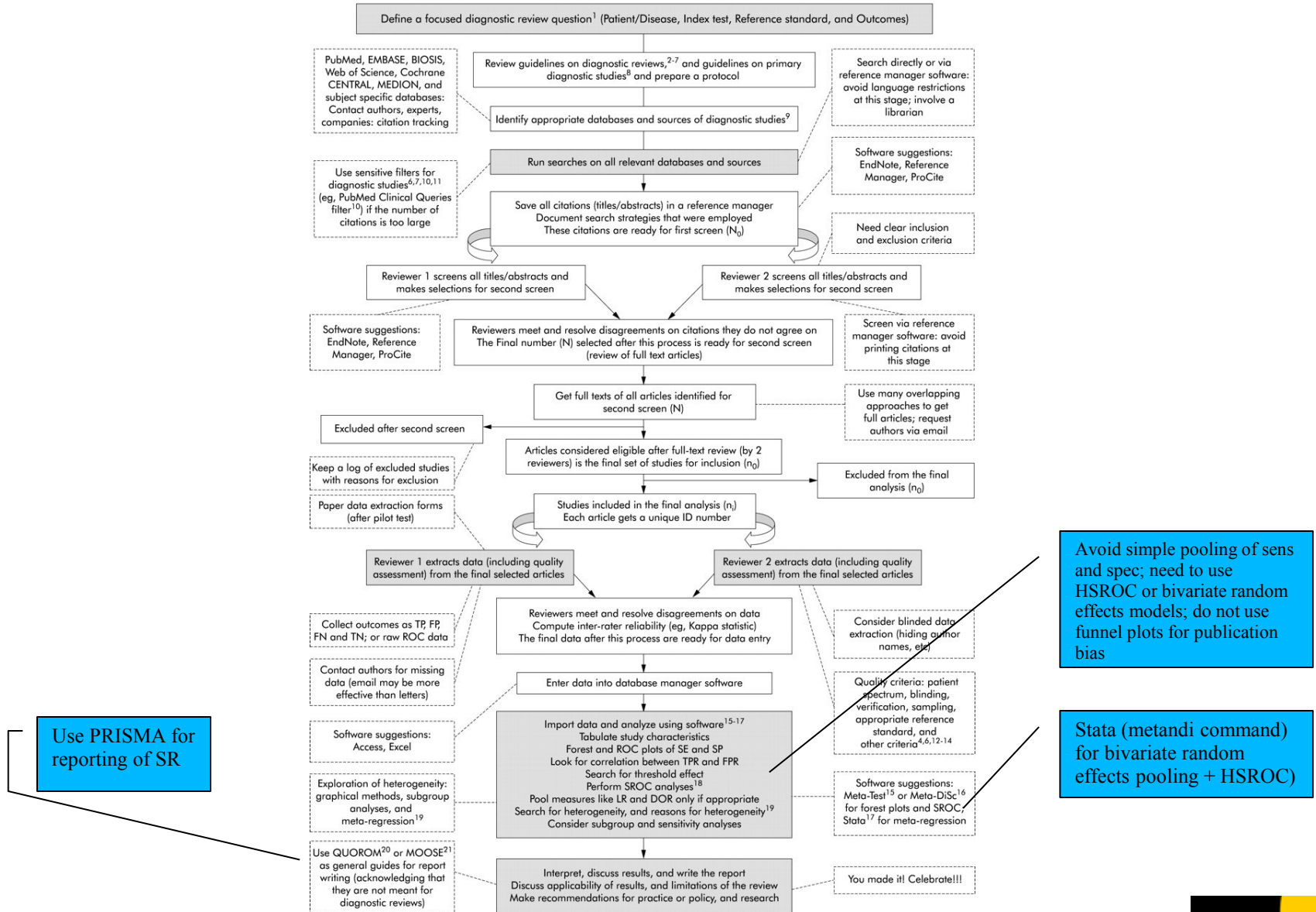
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# Road map for diagnostic accuracy reviews

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



# 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



**P**opulation  
**I**ntervention  
**C**omparison  
**O**utcome

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

# 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

## Accuracy of Rapid Influenza Diagnostic Tests

### A Meta-analysis

Caroline Chartrand, MD, MSc; Mariska M.G. Leeflang, DVM, PhD; Jessica Minion, MD, MSc; Timothy Brewer, MD, MPH; and Madhukar Pai, MD, PhD

**Purpose:** To examine the accuracy of rapid influenza diagnostic tests (RIDTs) in adults and children with influenza-like illness and evaluate factors associated with higher accuracy.

OPEN ACCESS Freely available online

PLOS MEDICINE

## Commercial Serological Tests for the Diagnosis of Active Pulmonary and Extrapulmonary Tuberculosis: An Updated Systematic Review and Meta-Analysis

Karen R. Steingart<sup>1</sup>, Laura L. Flores<sup>2,3</sup>, Nandini Dendukuri<sup>4</sup>, Ian Schiller<sup>4</sup>, Suman Laal<sup>5,6,7</sup>, Andrew Ramsay<sup>8</sup>, Philip C. Hopewell<sup>2,3</sup>, Madhukar Pai<sup>4\*</sup>

**Types of studies.** Diagnostic studies (with any study design) were included that evaluated serological tests for active TB (pulmonary and extrapulmonary TB) in patients who provided sera before or within 14 d of starting antituberculous treatment.

**Participants.** The participants constituted adults and children, with and without HIV infection, with suspected or confirmed active TB, from all clinical settings (clinic or hospital). The protocol for the current review included studies with at least ten TB cases. Studies could be performed in any country regardless of TB incidence or income status.

**Index test.** The index test was any commercial serological test for the diagnosis of active TB.

**Comparator tests.** There was either no test or smear microscopy used for comparison.

**Target conditions.** The target conditions were pulmonary and extrapulmonary TB.

**Reference standards.** Pulmonary TB required positivity on mycobacterial culture. (The previous review accepted positivity on either culture or smear microscopy as the reference standard [12].) Extrapulmonary TB required positivity on at least one of the following tests: culture, smear, or histopathological examination.

**Outcomes.** The outcomes were sensitivity and specificity.



## **2. Searching for studies**

# 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

# + Influenza rapid tests: Search strategy

Influenza, Human [Mesh]  
Influenza A virus [Mesh]  
Influenza B virus [Mesh]  
Influenza  
Flu  
grippe

Rapid test, rapid diagnos\*, rapid  
diagnostic test\*, point-of-care  
test\*, antigen detection test\*,  
antigen detection, rapid antigen  
test\*, immunoassay\*,  
immunochromatographic test\*  
Binax NOW, Directigen Flu, Flu  
OIA, QuickVue Influenza, Rapide  
detection Flu, SAS Influenza, TRU  
FLU, XPECT FLU, Zstat flu

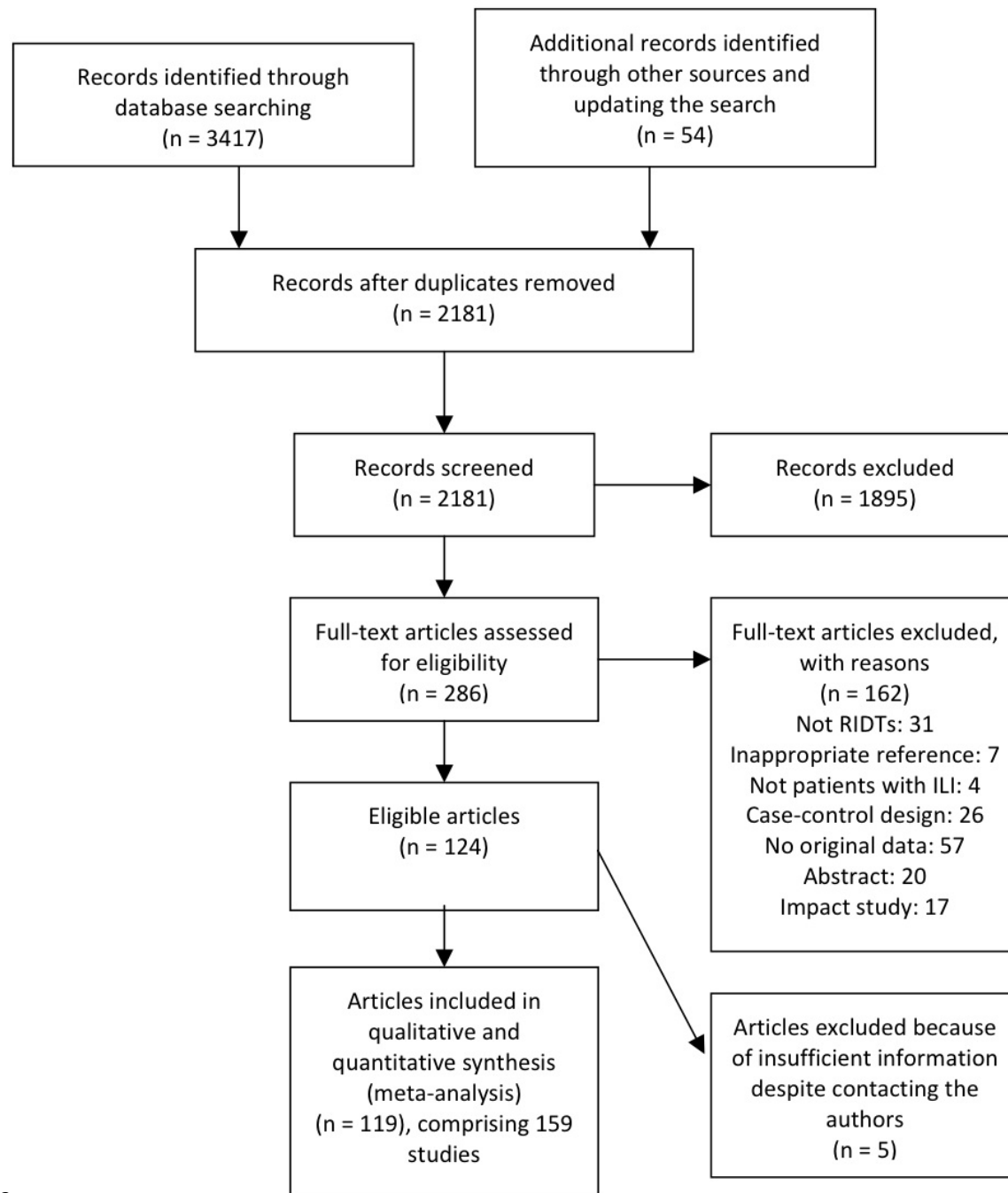
Databases: MEDLINE via Pubmed, EMBASE, Biosis et Web of Science  
March 2010, updated december 2011

Identification

Screening

Eligibility

Included



*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

# 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

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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

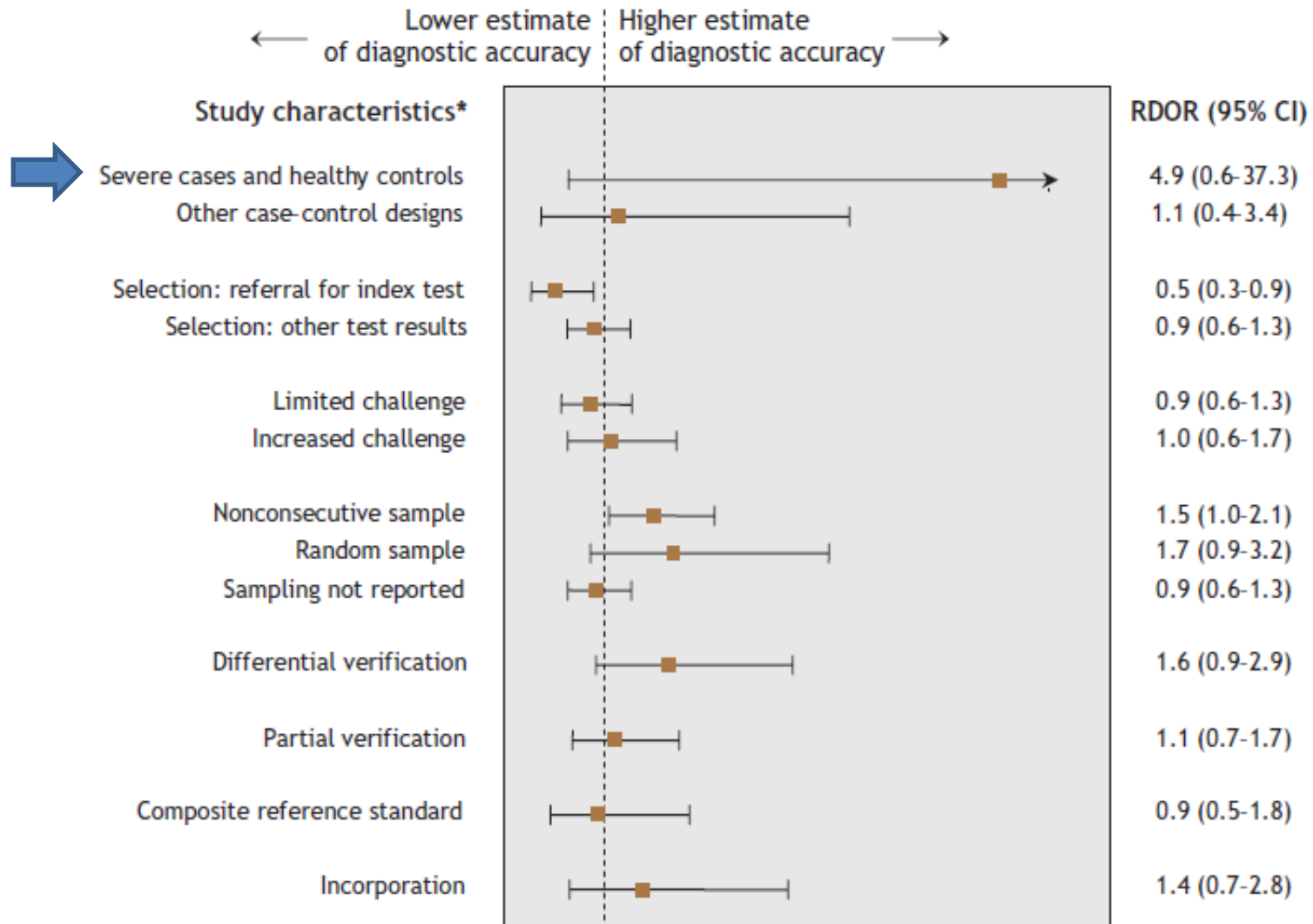
RESEARCH

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
<b>Study design</b>			
Case-control	19	86.5 (39.3, 190.2)	0.03
Cross-sectional	16	43.3 (22.5, 83.3)	0.94
<b>Blinded interpretation of test and/or reference standard results</b>			
Yes	21	46.9 (24.9, 88.6)	0.16
No	14	82.3 (39.8, 170.2)	0.70
<b>Consecutive or random sampling of participants</b>			
Yes	18	63.3 (32.8, 122.4)	0.20
No	17	46.8 (23.6, 92.8)	0.42
<b>Prospective data collection</b>			
Yes	18	59.9 (28.1, 127.6)	0.12
No	17	55.2 (29.9, 101.6)	0.59

\*Random effects model. † $\chi^2$  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<sup>\*1</sup>, Anne WS Rutjes<sup>2</sup>, Johannes B Reitsma<sup>2</sup>,  
Patrick MM Bossuyt<sup>2</sup> and Jos Kleijnen<sup>1</sup>Address: <sup>1</sup>Centre for Reviews and Dissemination, University of York, England, UK and <sup>2</sup>Department of Clinical Epidemiology and Biostatistics, Academic Medical Center, University of Amsterdam, The NetherlandsEmail: Penny Whiting<sup>\*</sup> - 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<sup>\*</sup> Corresponding author

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# QUADAS, 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<sup>\*</sup>

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. This 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](http://www.annals.org)

For author affiliations, see end of text.

<sup>\*</sup> For members of the QUADAS-2 Group, see the Appendix (available at [www.annals.org](http://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-2, 2011

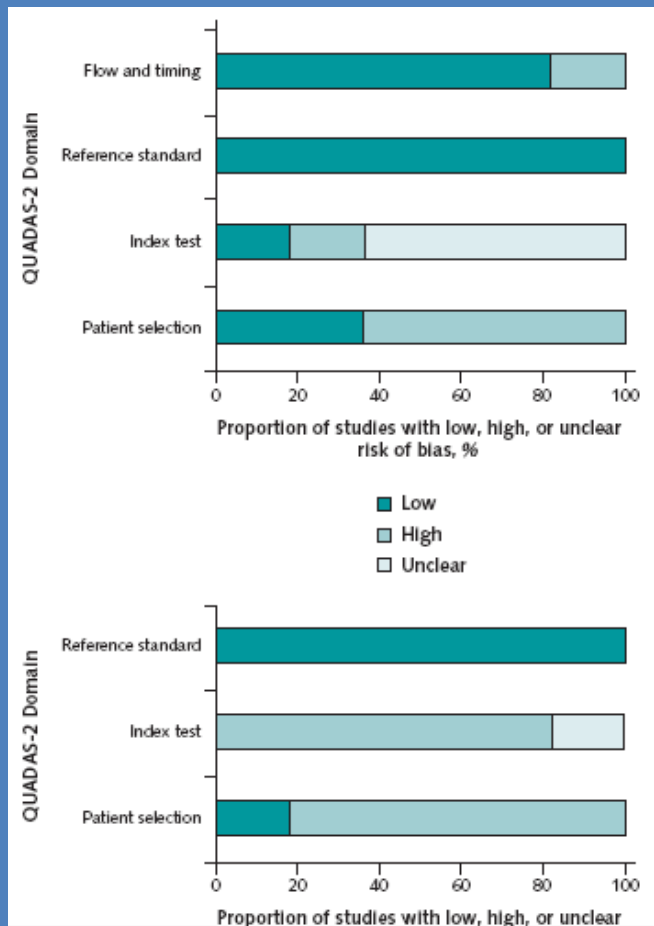
# Suggested displays – QUADAS-2

## RESEARCH AND REPORTING METHODS | QUADAS-2: A Revised Tool

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/>

# In general, diagnostic studies are poorly done and reported (contacting authors is helpful)

OPEN ACCESS Freely available online



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

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### 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.

## 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

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- [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:

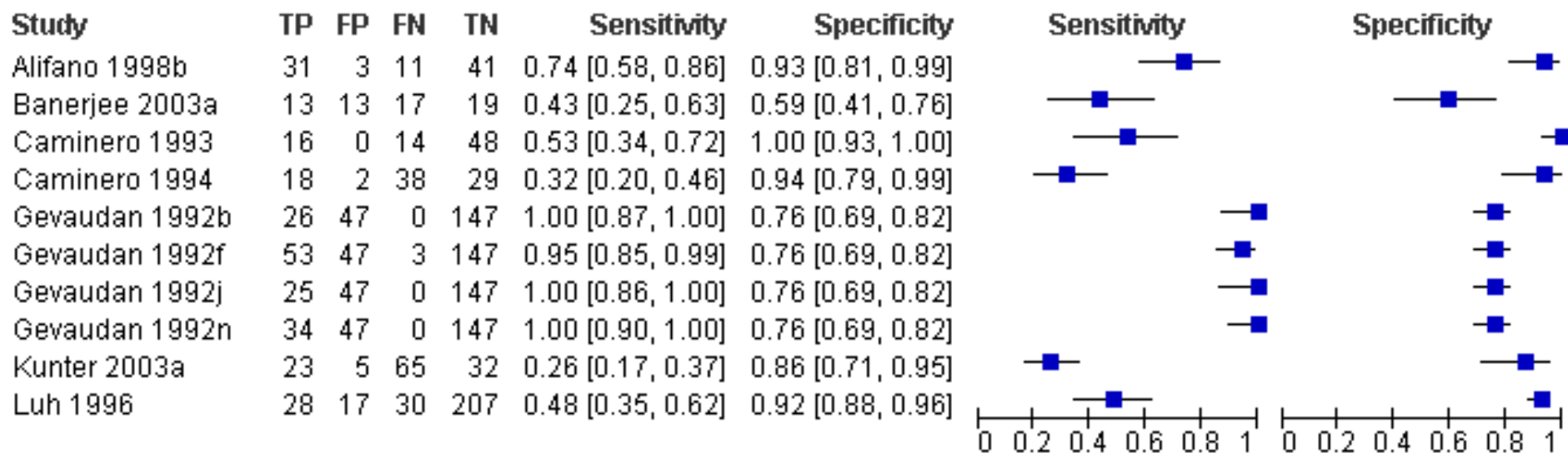
- [What's New](#) - List of changes made in updates to RevMan 5
- [Wish List](#) for the next major version
- [FAQ](#) - Check if your question has been already answered
- [Suggestion Form](#) - Send suggestions for improving RevMan
- [Updates](#) - To update RevMan to the latest version
- [Next Release](#)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



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



# 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

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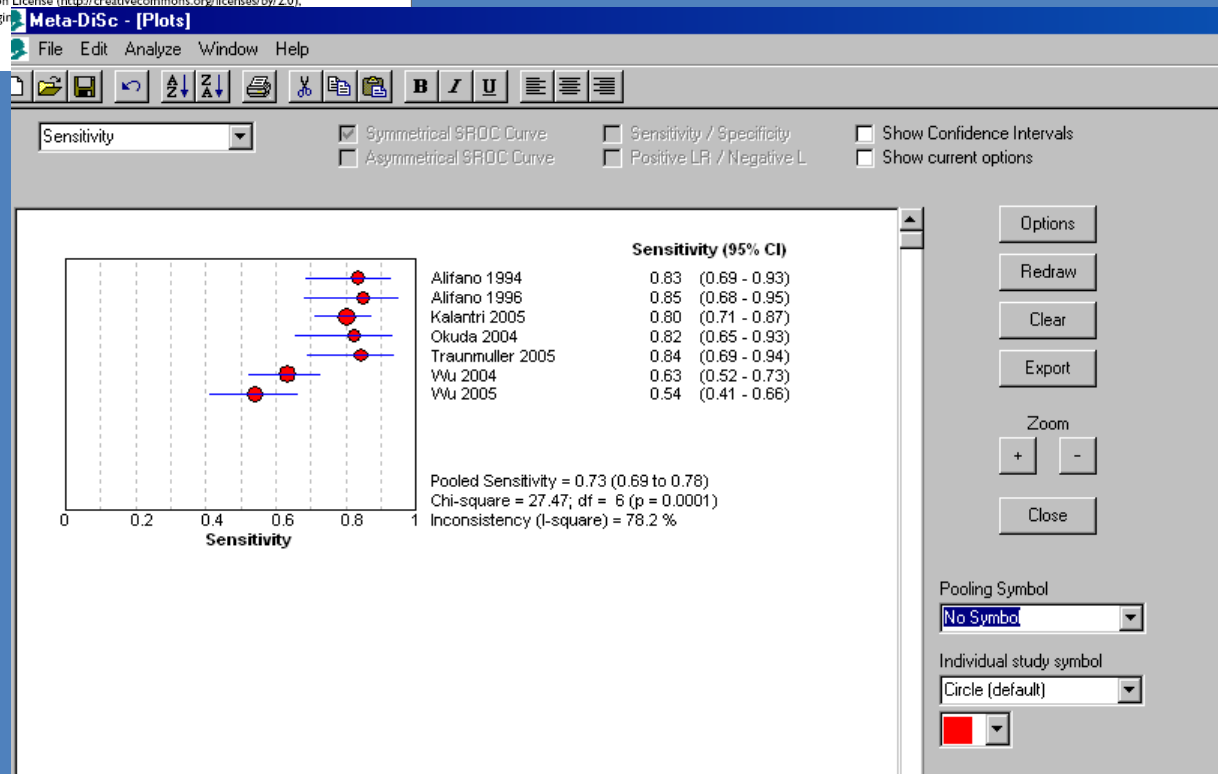
Accepted: 12 July 2006

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

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# Statistical models for meta-analysis of diagnostic studies

- Simple, separate pooling of sens and spec should not be done
- Two recommended 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 2×2 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

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*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

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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

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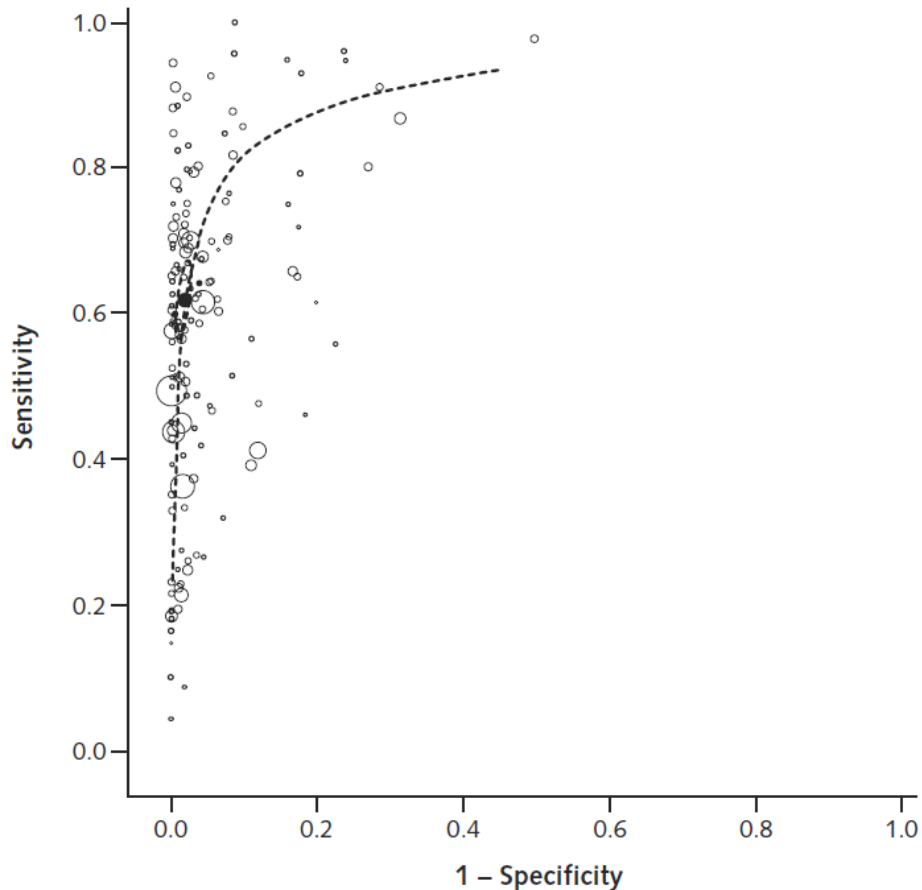
# Bivariate model vs HSROC model

- Where studies report a common threshold (or cut-off) for a positive result, use the bivariate model
- Where studies report several different thresholds, use the HSROC model



# Influenza rapid tests

*Figure 2. Hierarchical summary receiver-operating characteristic curve plot of rapid influenza diagnostic test studies.*



Sensitivity: 62.3% (57.9 – 66.6)  
Specificity: 98.2% (97.5 – 98.7)  
LR+: 34.5 (23.8 – 45.2)  
LR-: 0.38 (0.34 – 0.43)



# Stata command, metandi

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

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

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**Abstract.** Meta-analysis of diagnostic test accuracy presents many challenges. Even in the simplest case, when the data are summarized by a  $2 \times 2$  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

# Stata output

```

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  metandi tp fp fn tn, nolog
        
```

Name	Label	Type	Format
id	ID	byte	%8.0g
subid	Subid	str1	%3s
author	AUTHOR	str10	%10s
year	YEAR	int	%8.0g
site	Site Laboratory	str14	%14s
typeoftest	Type of test	str8	%3s
drug	Drug	str3	%3s
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"

. 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]
<b>Bivariate</b>					
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(logits)	1	.			.
<b>HSROC</b>					
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	.			.
<b>Summary pt.</b>					
Se	.8084602	.0298576			.7430971 .8603191
Sp	.9928467	.0051839			.9707566 .9982798
OR	126.8236	168.7682			12.80834 2811.145
LR+	113.0187	83.0438			26.77351 477.0848
LR-	.1929198	.0303402			.1417457 .2625692
1/LR-	5.1835	.8151999			3.808519 7.054888
Covariance between estimates of E(logitse) & E(logitSp) .0351746					

Command

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

# Heterogeneity: very common in diagnostic SRs

- 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

# Example: subgroup analysis

*Table 2. Accuracy Estimates From Subgroup Analyses*

Characteristic	Pooled Sensitivity (95% CI), %	P Value	Pooled Specificity (95% CI), %	P Value
<b>Population</b>				
Children (60 studies)	66.6 (61.6–71.7)	<0.001	98.2 (97.5–99.0)	0.135
Adults (33 studies)	53.9 (47.9–59.8)	Reference	98.6 (98.0–98.9)	Reference
<b>Virus type</b>				
Influenza A (72 studies)	64.6 (59.0–70.1)	0.62	99.1 (98.7–99.4)	<0.001
Influenza B (27 studies)	52.2 (45.0–59.3)	0.050	99.8 (99.7–99.9)	<0.001
Influenza A and B (47 studies)	62.3 (55.2–69.4)	Reference	96.1 (94.4–97.8)	Reference
<b>Study conducted during the H1N1 pandemic</b>				
Yes (41 studies)	56.3 (48.7–63.9)	0.065	98.9 (98.3–99.5)	0.022
No (74 studies)	65.0 (59.7–70.4)	Reference	97.5 (96.6–98.5)	Reference
<b>Index test*</b>				
BinaxNOW (17 studies)†	57.0 (45.9–67.5)	0.028‡	98.6 (96.9–99.3)	0.057‡
Directigen Flu A (10 studies)	76.7 (63.8–86.0)	0.49‡	97.2 (92.6–99.0)	0.62‡
Directigen Flu A+B (30 studies)	57.2 (48.8–65.2)	0.011‡	99.3 (98.8–99.6)	<0.001‡
QuickVue Influenza (16 studies)	69.0 (58.1–78.2)	0.66‡	95.8 (91.3–98.0)	0.82‡
QuickVue Influenza A+B (21 studies)	48.8 (39.0–58.8)	<0.001‡	98.4 (96.8–99.2)	0.064‡
<b>Reference standard</b>				
RT-PCR (67 studies)	53.9 (48.2–59.6)	<0.001	98.8 (98.3–99.3)	0.002
Culture (48 studies)	72.3 (66.8–77.9)	Reference	96.7 (95.2–98.3)	Reference

# 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 D et al. PLoS ONE 2008.

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

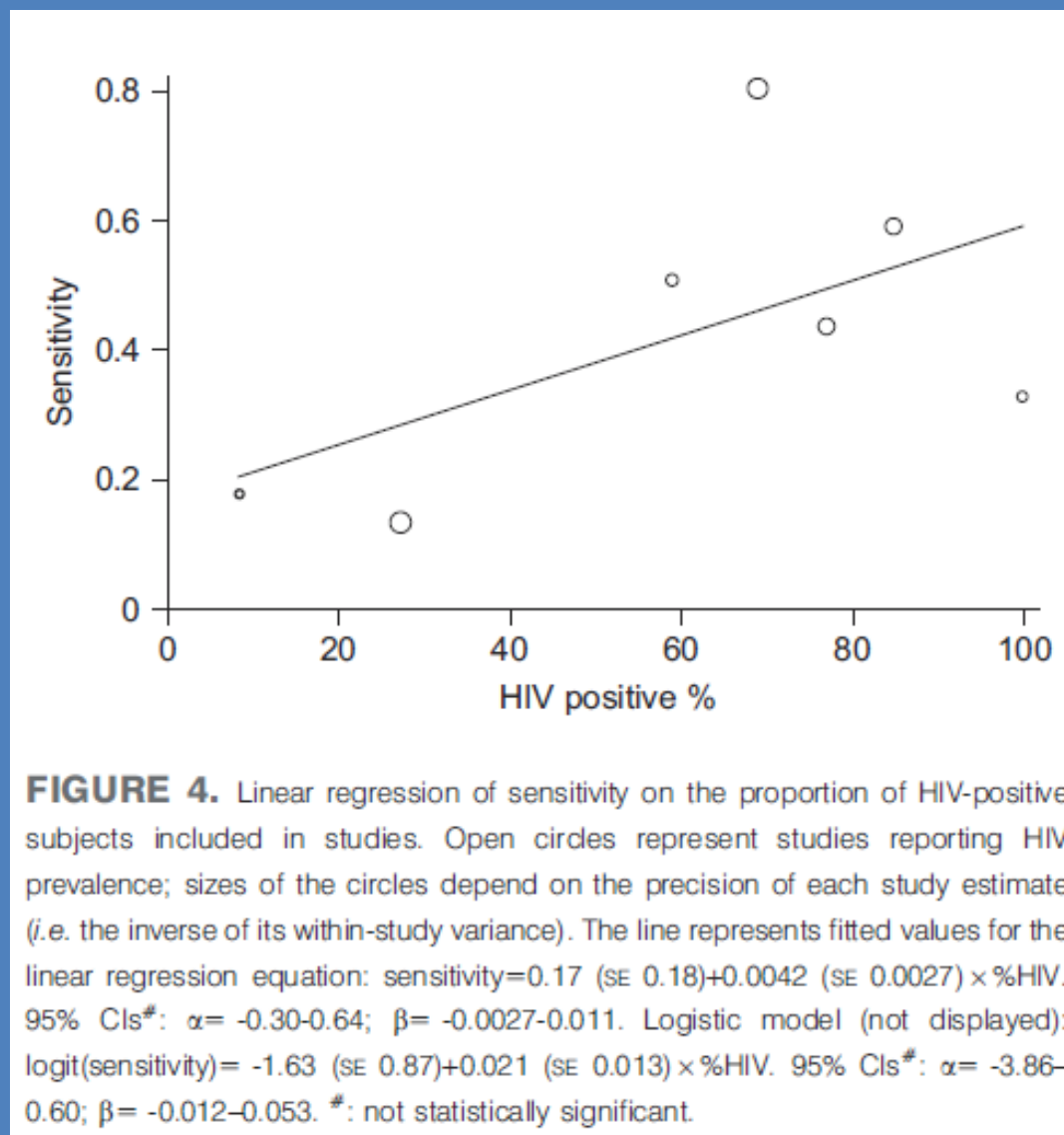
Comparison	Model Coefficient	Relative Diagnostic Odds Ratio (95% CI)	P value
Threshold Effect (S)	-0.21	—	0.01
Retrospective/Both (17) vs Prospective Design (108)	0.13	1.14 (0.56, 2.33)	0.71
Some Convenient Sampling/NR (80) vs Consecutive/Random Sampling (45)	0.38	1.46 (0.87, 2.43)	0.15
No Blinding/NR (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



# Exploration of heterogeneity – urine LAM ELISA for TB



# Publication bias

- Formal assessment of publication bias using methods such as funnel plots or regression tests is not recommended for diagnostic test accuracy studies

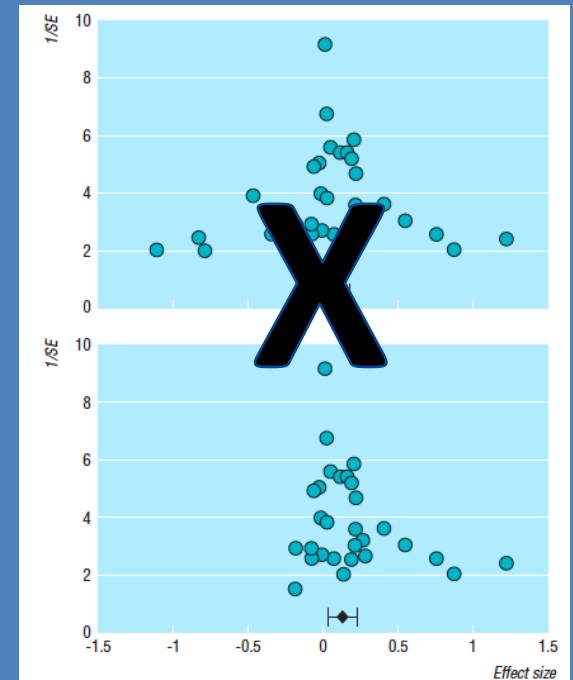


Fig 1 Typical funnel plot generated from 35 simulated studies (top) and same data with five missing studies showing a typical manifestation of publication bias (bottom)

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

# Issues to discuss

- 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 were the limitations of the SR itself?
- 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
<b>Diagnostic accuracy for diagnosis of pulmonary TB</b>						
<b>All patients</b>	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 ⊕⊕⊕○
<b>Smear positive patients</b>	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 ⊕⊕⊕○

# Some general limitations of diagnostic SRs

- Literature search strategies are imperfect and studies can be missed
- Publication bias is always a concern
- Poor quality studies or poorly reported studies
- Unexplained heterogeneity
- Not enough studies on clinical impact of tests
- Industry supported studies or COI of study authors
- COI of systematic reviewers
- Keeping up to date in rapidly evolving fields

# Keeping systematic reviews updated!

## **Interferon- $\gamma$ assays in the immunodiagnosis of tuberculosis: a systematic review**

2004

Madhukar Pai, Lee W Riley, and John M Colford Jr

**Annals of Internal Medicine**

ARTICLE

## **Meta-analysis: New Tests for the Diagnosis of Latent Tuberculosis Infection: Areas of Uncertainty and Recommendations for Research**

2007

Dick Menzies, MD, MSc; Madhukar Pal, MD, PhD; and George Comstock, MD, DrPH

**Annals of Internal Medicine**

REVIEW

## **Systematic Review: T-Cell–based Assays for the Diagnosis of Latent Tuberculosis Infection: An Update**

2008

Madhukar Pal, MD, PhD; Alice Zwerling, MSc; and Dick Menzies, MD, MSc

## **Predictive value of interferon- $\gamma$ release assays for incident active tuberculosis: a systematic review and meta-analysis**

2012

Molebogeng X Rangaka, Katalin A Wilkinson, Judith R Glynn, Daphne Ling, Dick Menzies, Judith Mwansa-Kambafwile, Katherine Fielding, Robert J Wilkinson, Madhukar Pai

# References and Tools

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