Precision in Meta-Analysis

Are we being misled by the play of chance?

Presentation objectives

Intro to precision in meta-analysis
 (MA), overview of current research

2. Debate on future challenges and required efforts



I: IntroII: Preparation for debateIII: Overview of current researchIV: Debate and questions

Part I:

Intro

We perform significance tests because we want to control the risk that we are being mislead by the play of chance

When a meta-analysis becomes 'statistically significant', authors, journal editors, policy makers, etc. typically put strong confidence in the estimated effect

This is all fine if the statistical methods do what they are supposed to do.

But if they don't....
how often are we actually being mislead by the play of chance?
how often will the implications for clinical practice be serious?

Statistical tests in meta-analysis

We typically use Z-statistics to test for 'statistical significance'

Z = Estimated treatment difference Standard Error

Z is subsequently transformed to a p-value

Statistical tests in meta-analysis

We conclude some treatment effect is statistically significant when our p-value crosses below the overall type I error, a (or when |Z| crosses above Z_a)

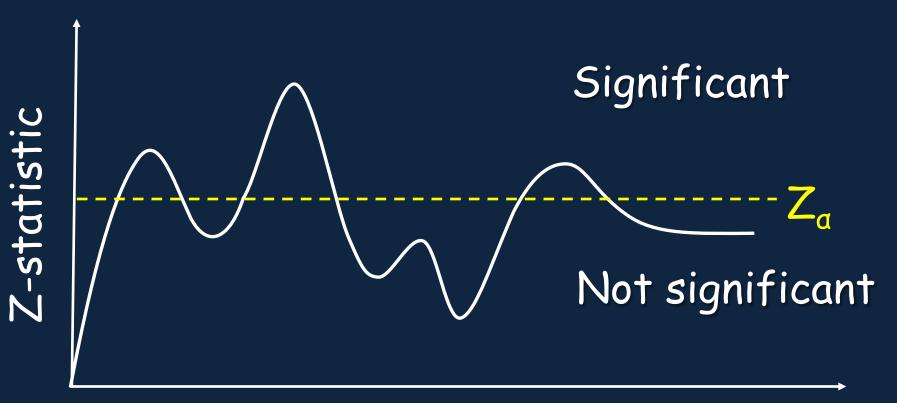
We assume the pooled treatment effect estimate is reliable

Statistical tests in meta-analysis

This conduct can lead to

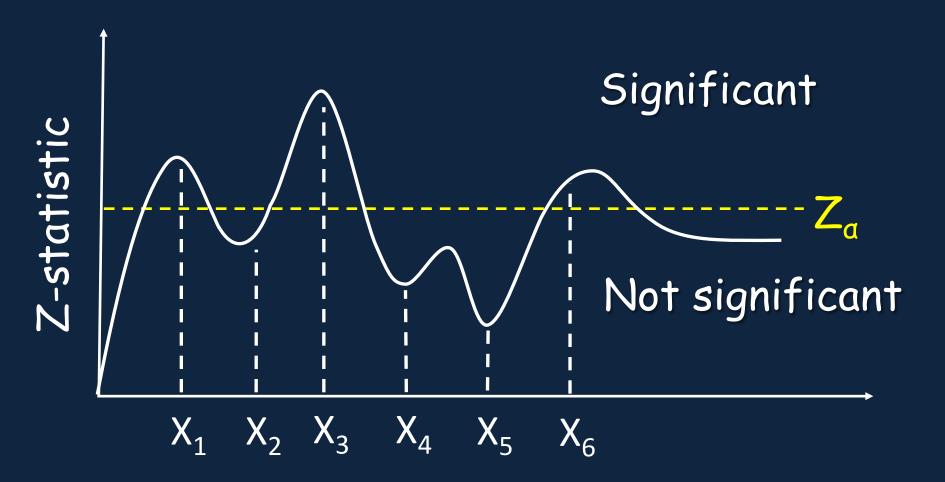
- 1. exaggeration of type I error (multiplicity)
- 2. overestimation of treatment effects

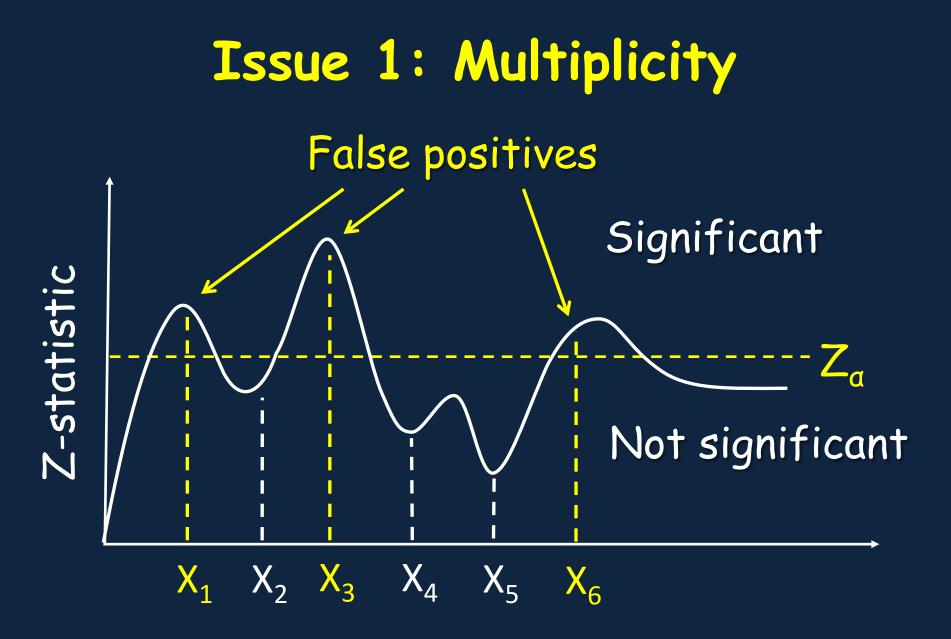
Issue 1: Multiplicity



Number of patients randomized

Issue 1: Multiplicity





Issue 1: Multiplicity

Every time we test for statistical significance over time we increase the risk of type I error (multiplicity) This problem occurs in meta-analysis

due to updating

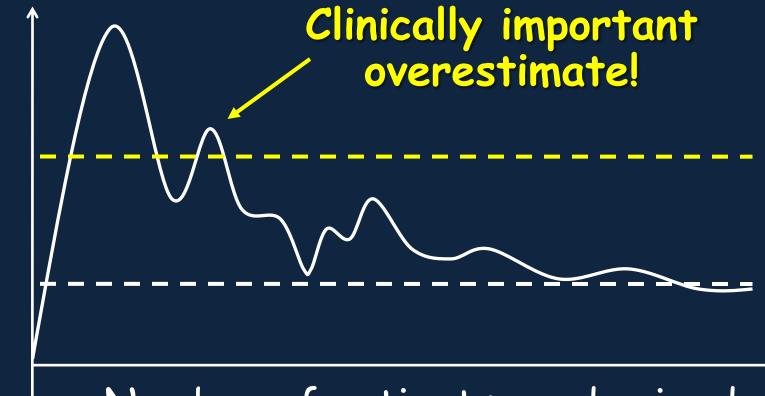
Issue 2: Overestimation

In 'early' meta-analysis precision is low, and thus, the standard error is large.

Z = Estimated treatment difference Standard Error

In 'early' statistically significant MAs, the est. treatment difference are large

Issue 2: Overestimation



Effect

Number of patients randomized

Part II: Preparation for debate

INPUT WANTED

As you will see in the following slides, both false positives (multiplicity) and overestimation is often problematic in metaanalysis that draw 'conventional' statistical inferences

INPUT WANTED

To the extent policy makers and clinicians rely on meta-analyses, the implications may be serious

INPUT WANTED

Potential solutions comprise
1. Adjustment of thresholds or tests
2. Setting some yardstick for when 'the answer is in' (e.g. required sample size)



Research efforts so far have focused on superiority testing for binary outcome meta-analysis...



- What more is needed to seal the deal?
- What about inferiority testing?
- What about other types of data?
 - continuous (HRQL)
 - time-to-event/survival

Questions?

Clinicians are notorious for relying on thresholds, statisticians for making things too complex.

For decision-making, can we meet in the middle?



How do we (largely) avoid misuse of proposed methodologies?

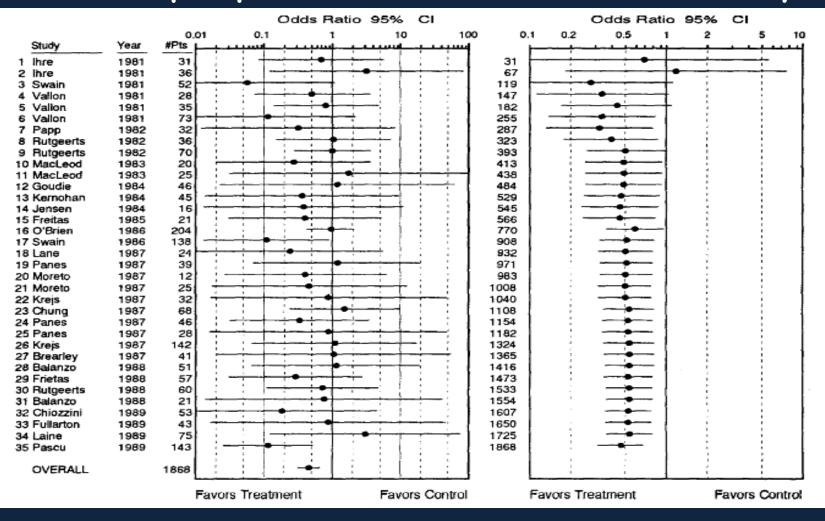
Part III: Overview of current research Part III: From 1992 to present day



Lau et al. proposed cumulative meta-analysis

NEJM 1992; 327(4):248-254 *J Clin Epi* 1995; 48(1):357-371

Lau et al. proposed cumulative meta-analysis



Maybe this is where is went wrong...

"The cumulative aspect of the [frequentist] meta-analysis ... are calculated and interpreted in through the Bayesian paradigm"

Conclusion: no need to adjust for multiplicity

Berkey et al simulated 'uncertainty of first time to significance' with associated power and type I error using real MA data (MCMC)

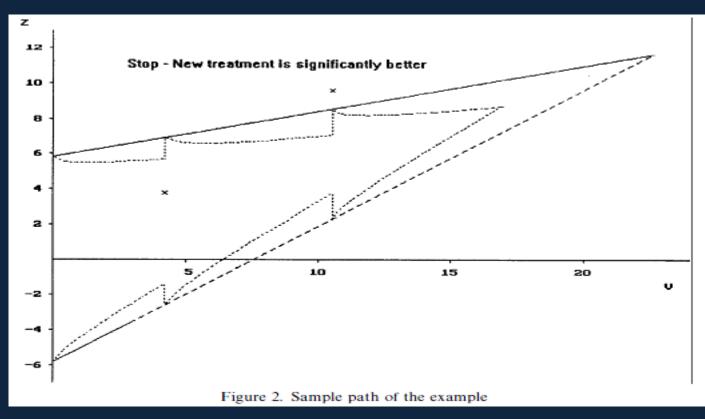
- With a=5%, the actual type I error after 15 trials was 15% (update for every trial)

Contr Clin Trials 1996; 17(5):357-371

Pogue&Yusuf and Whitehead seperately proposed use of 'information size' considerations and sequential monitoring boundaries to control type I error

> *Stat Med* 1997; 16(24):2901-2913 *Contr Clin Trials* 1997; 18(6):580-93 *Lancet* 1998; 351(9095):47-52

Whitehead proposed use of the *triangular test* for random-effects meta-analysis and performed a simulation study

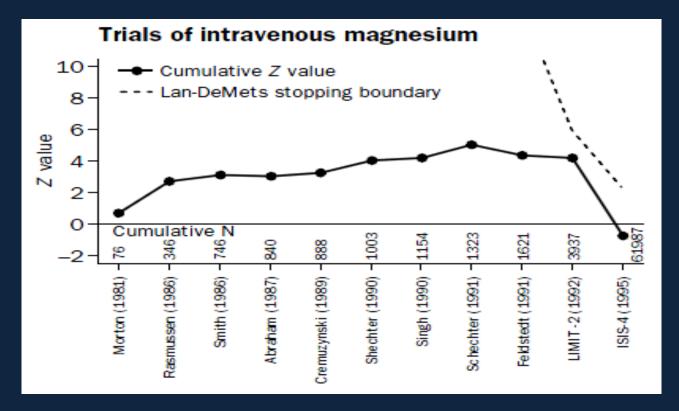


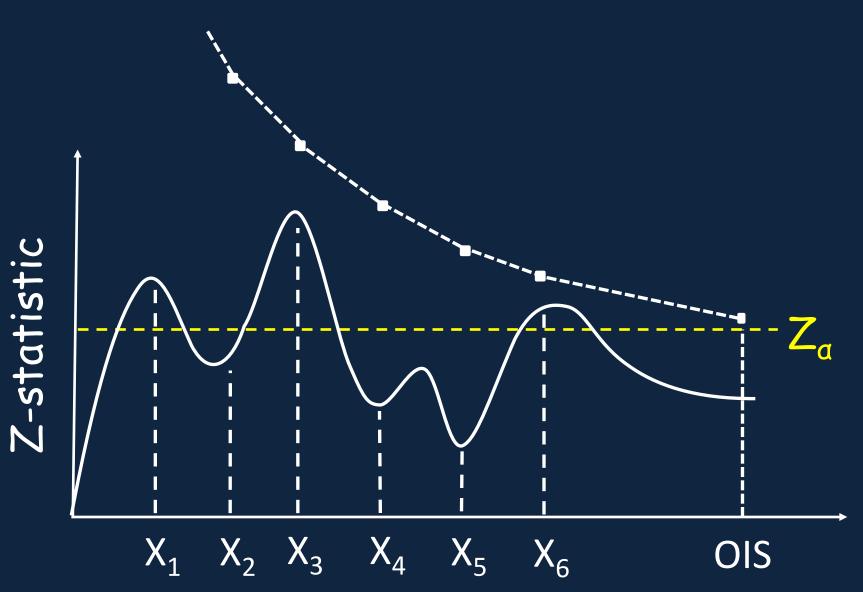
Whitehead proposed use of the *triangular test* for random-effects meta-analysis and performed a simulation study

 Only found satisfactory control of type I error when heterogeneity was mild

 Information scale (Fischer) is only reliable with good heterogeneity estimates

Pogue&Yusuf proposed use of Lan-DeMets a-spending monitoring boundaries and the Optimum Information Size (OIS)





First signs of empirical evidence First actual application

> Proc Natl Acad Sci USA 2001; 98(3):831-836 J Clin Epi 2004; 57(11):1124-1130 BMJ 2005; 331(7512):313-321

First signs of empirical evidence:

Ioannidis et al looked at the 'Evolution of treatment effects over time' in 60 metaanalyses on interventions in perinatal medicine and for myocardial infarction

Ioannidis 60 meta-analyses: relative change in estimates per added trial

Table 1. 95% prediction intervals for the relative change in the treatment effect (odds ratio) for different numbers of accumulated patients (cumulative sample size, *N*)

	Pregnancy/perinatal		Myocardial infarction	
Patients, N	Fixed effects	Random effects	Fixed effects	Random effects
100	0.37-2.78	0.32-3.13	0.18-5.51	0.23-4.43
500	0.59-1.71	0.56-1.71	0.60-1.67	0.63-1.58
1,000	0.67-1.49	0.65-1.53	0.74-1.35	0.76-1.32
2,000	0.74-1.35	0.73-1.37	0.83-1.21	0.84-1.20
15,000	0.85-1.14	0.86–1.15	0.96-1.05	0.96-1.05

Ioannidis 60 meta-analyses: relative change in estimates per added trial

"More than 10,000 patients are required to relieve uncertainty about the first decimal point in the odds ratio of a treatment effect reported by a meta-analysis."

First sign of empirical evidence:

Trikalinos et al looked at the evolution of treatment effects and statistical significance in 100 mental health metaanalyses

100 patients: subsequent changes in odds ratios of 3- to 5-fold were common

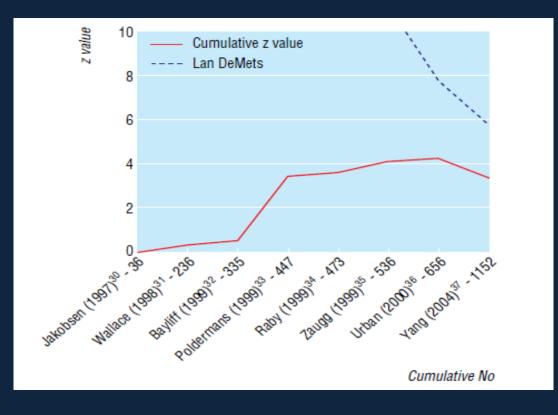
500 patients, changes >1.5-fold were only observed in 5% of the meta-analyses

>2000 patients randomised, subsequent changes were unlikely

In the interim, 8 of the 44 meta-analyses showing no effect were temporarily statistically significant (-18.2%) may type T error

(=18.2% max type I error)

First application by PJ Devereaux et al: Meta-analysis of periop beta-blockade in non-cardiac surgery



 More compelling evidence from empirical and simulation studies

- Methodological proposals

- Increasing awareness (e.g. GRADE)

- Applications in systematic reviews

Copenhagen Trial Unit group: Applied monitoring boundaries and heterogeneity adjusted OIS to all Cochrane neonatal and other meta-analyses

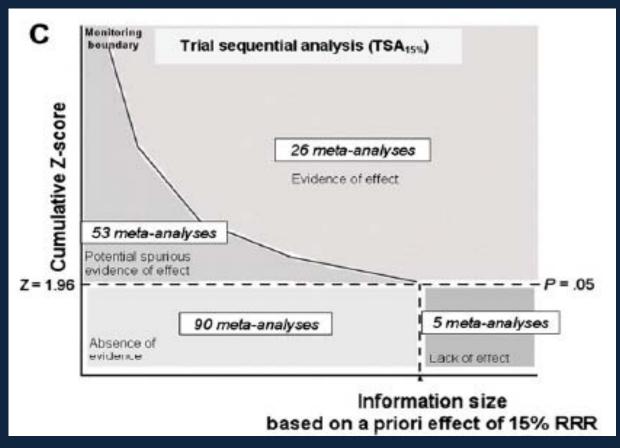
> *J Clin Epi* 2008; 61(1):64-75 *J Clin Epi* 2008; 61(8):763-769 *Int J Epi* 2009; 38(1):276-86 *Int J Epi* 2009; 38(1):287-98 *BMC Med Res Meth* 2009; 9:86

Copenhagen Trial Unit group: Proposed heterogeneity adjusted Optimum Information Size (OIS)

OIS=N/(1-H)

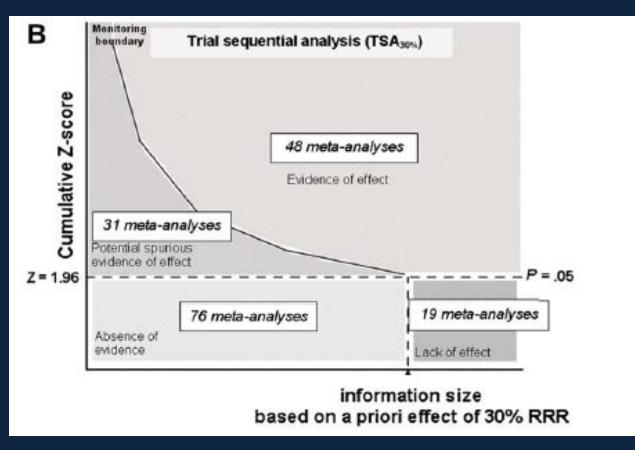
N is the required sample size for a RCT H is the degree of heterogeneity (0-100%)

Copenhagen Trial Unit group: Application to Cochrane neonatal reviews





Copenhagen Trial Unit group: Application to Cochrane neonatal reviews



Copenhagen Trial Unit group: Application to Cochrane neonatal reviews

Conclusion: Most meta-analysis are inconclusive and potentially spurious

The recommendations in current guidelines is discrepant with the strength of evidence

Copenhagen Trial Unit group: Application to 33 meta-analysis surpassing their heterogeneity adjusted OIS

6 out of 21 'positive' meta-analysis yielded important overestimates at *first time of statistical significance*

Copenhagen Trial Unit group: Application to 33 meta-analysis surpassing their heterogeneity adjusted OIS

3 out of 12 'negative' meta-analysis yielded false positive results in the interim

Copenhagen Trial Unit group: Application to 33 meta-analysis surpassing their heterogeneity adjusted OIS

Monitoring boundaries eliminated all spurious inferences



Compelling evidence from simulations and other proposed methods

Clin Trials 2007; 4:329 *J Clin Epi* 2009; 62:825-830

Hu et al proposed a *penalized* Z-statistic based on *the law of the iterated logarithm* and simulation

$Z^{*}(k) = Z(k)/(\sqrt{\lambda} \cdot \log(\log(I(k))))$

Where k is the number of trials and I(k) is the inverse of the pooled variance

Hu et al simulations

- Repeated significance testing (with $\alpha = 5\%$) yields an actual type I error of 15-35\%
- The penalized Z-statistics exhibits good control of the type I error (reasonable values for λ is provided in the paper)

Borm et al proposed a *k-fail-safe* adjustment of the p-value based on the max number of MA updates and simulation

Regression on Type I error = $\alpha \cdot f$, where f is some function of the max number of updates in the meta-analysis

Borm et al proposed a *k-fail-safe* adjustment of the p-value based on the max number of MA updates and simulation

 $P^* = P \cdot \sqrt{6 \cdot Max no. updates - 1.5}$

Borm et al simulations

- Repeated significance testing (with one-sided α =2.5%) yields 2- to 7-fold inflation of the type I error
- The adjusted P-value exhibits good control of the type I error

GRADEProfiler recommends downgrading of the quality of the overall evidence if a meta-analysis does not surpass its OIS (or has less than 300 events)

For that reason, a good number of systematic reviews now consider the meta-analysis sample size

Copenhagen Trial Unit group: A number of applications to systematic reviews as well as empirical studies

Copenhagen Trial Unit group - applications:

Whenever one of us was invited to co-author a SR we have used OIS and monitoring boundaries to achieve reliable statistical inferences

More simulation studies More empirical studies More applications Abuse of the methodology?

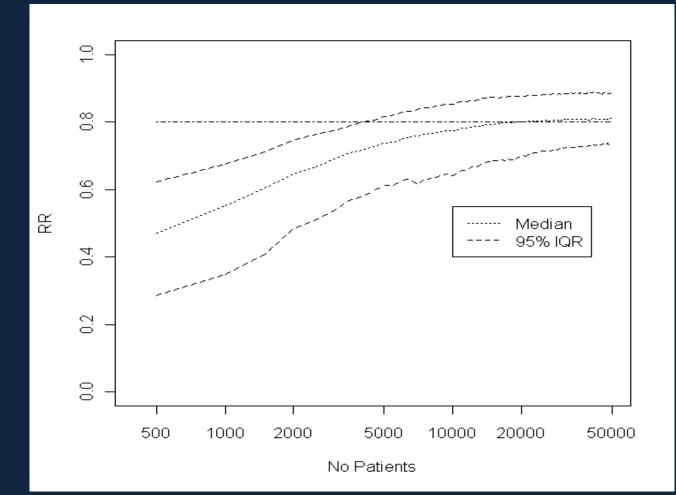
Simulation study sneak peak

Plausible cardiology meta-analysis scenario (based on survey of Cochrane Heart Group meta-analysis on mortality)

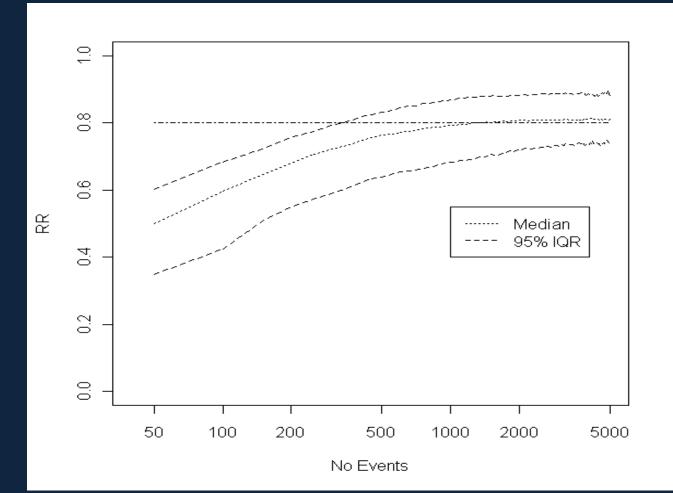
Plausible cardiology meta-analysis scenario (random-effects model simulation)

True effect: RR=0.80 Event rate: 1%-15% Moderate heterogeneity (RR, 0.60-1.05) Trial sizes: 40-400(25%), 401-1000(65%) and 1001-10000(10%)

Plausible cardiology meta-analysis scenario



Plausible cardiology meta-analysis scenario

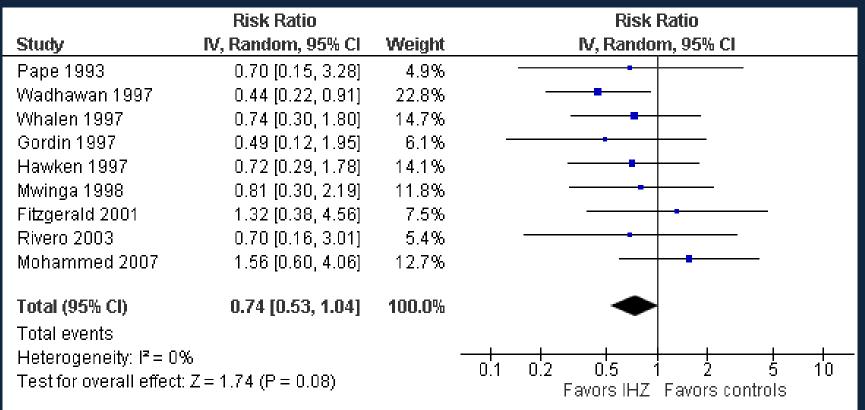


Alternative applications: Isoniazid chemoprophylaxis (IHZ) for preventing tuberculosis (TB) among purified protein derivative negative (PPD-) HIV infected individuals

IHZ: antibiotic PPD+/-: test for active/past TB

Thorlund K, Aranka A, Mills E. *Clin Epi* (in press)

Alternative applications: 9 trials, 2911 patients, event rate 2-12%



Alternative applications: 9 trials, 2911 patients, event rate 2-12%

Relative risk 0.74 (95% CI, 0.53 to 1.04) OIS=10500 patients (α=5%, β=80%, PC=5%, RRR=25%)

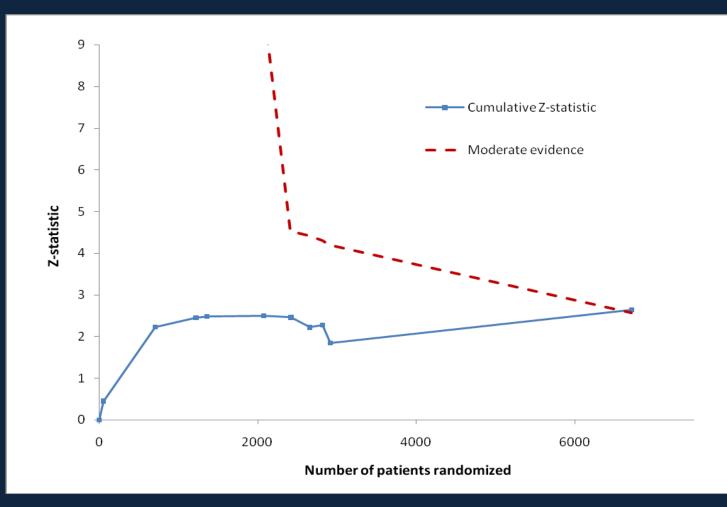
The answer is not in!

Need an additional 7500 patients (too much?)

Applied monitoring boundaries and played around with numbers to approximate:

'How large must a new trial be for the MA to cross the monitoring boundaries?' (assuming RRR=25% and PC=5%)

Topping up the sample size (3800 pts)



Summary

- Type I error due to multiplicity
 - Simulation studies: 2- to 7-fold
 - Empirical studies 4 to 5 fold

Summary

- Treatment effect estimates
 - 'Early' = unrealiable
 - 'Early' significant = overestimate
 - Definition of 'early' differs across

medical areas



- Methodology
 - OIS and monitoring boundaries
 - adjustment/penalization of
 - Z-statistics and p-values



OIS and monitoring boundaries have some support from studies and applications

Other methods have never been applied to real MA data

Part IV: Questions and debate



Research efforts so far have focused on superiority testing for binary outcome meta-analysis...



- What more is needed to seal the deal?
- What about inferiority testing?
- What about other types of data?
 - continuous (HRQL)
 - time-to-event/survival
- Other...



Clinicians are notorious for relying on thresholds, statisticians for making things too complex.

For decision-making, can we meet in the middle?



How do we (largely) avoid misuse of proposed methodologies?

Bring it on...