

Meta-analysis when the numbers get tricky

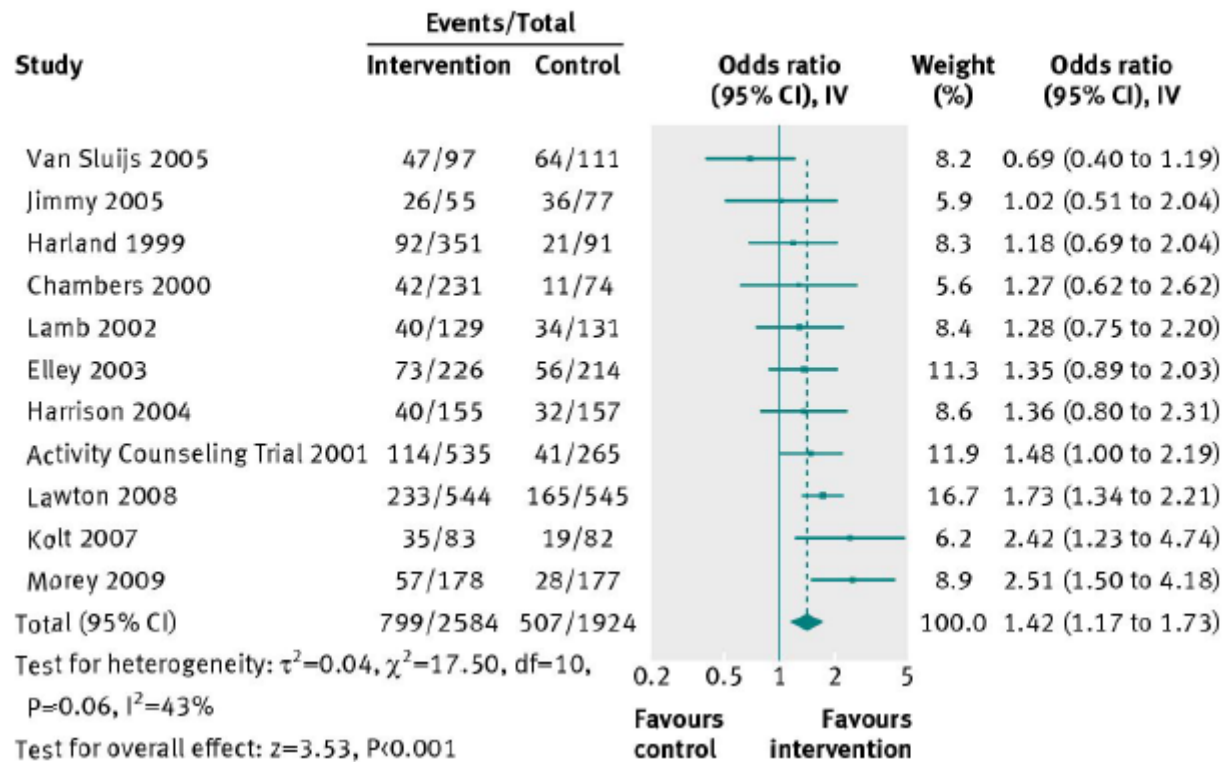
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Outline

- recap “standard” meta-analysis methods
 - effect measures for dichotomous data
 - for continuous data
- some examples of other types of data
- ways to handle these
- what we DON'T know how to handle
- summary

Dichotomous data

Proportion of participants achieving physical activity target



Effect measures for dichotomous data

- Risk ratio

$$\frac{\textit{Proportion with event in intervention}}{\textit{Proportion with event in control}}$$

- Odds ratio

$$\frac{\textit{Events in intervention} / \textit{No events in intervention}}{\textit{Events in control} / \textit{No events in control}}$$

- Risk difference

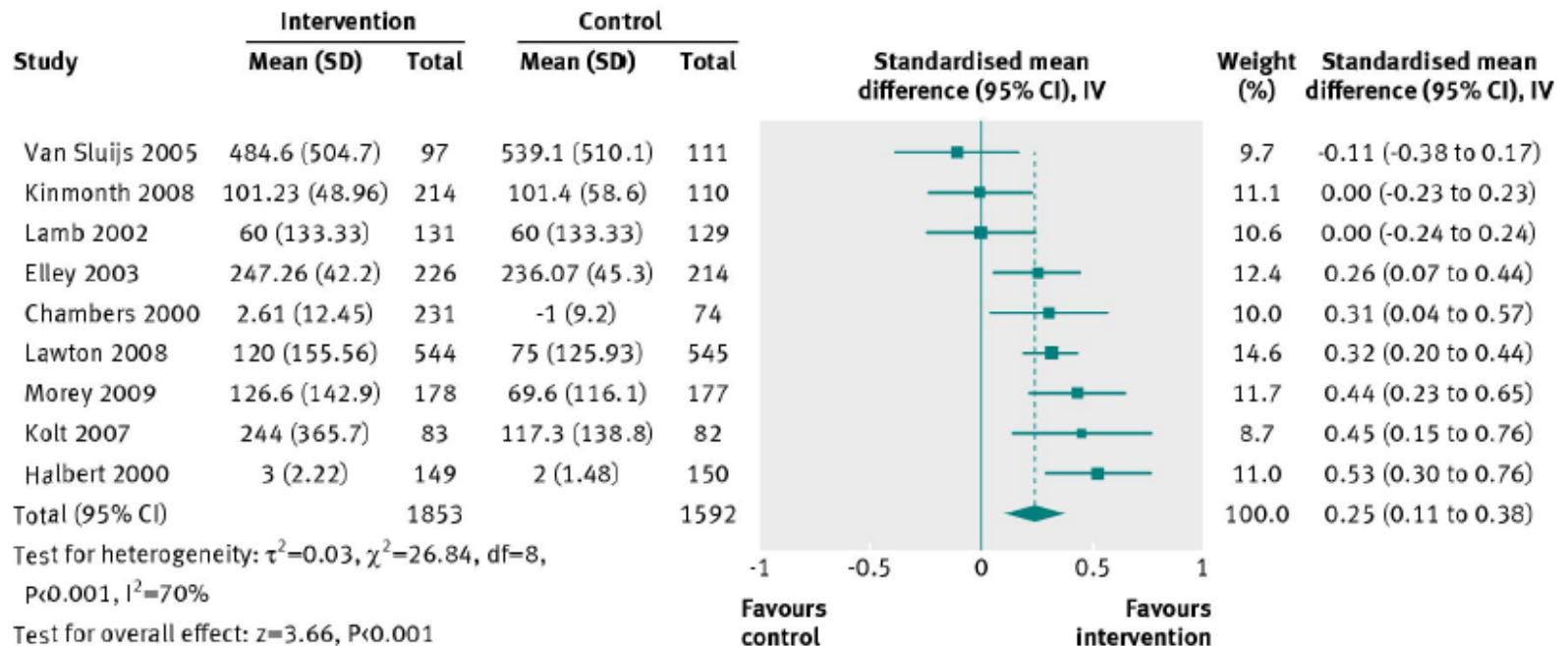
$$\textit{Proportion with event in intervention} \\ - \textit{Proportion with event in control}$$

Effect measures for continuous data

- mean difference
 - need mean and SD in each group
 - assumes approximately normally-distributed
- standardised mean difference
 - different measurement scales

Continuous data

Fig 4 Individual study and pooled effects of physical activity promotion on self reported physical activity at 12 months (continuous data). Random effects model used. SD=standard deviation; 95% CI=95% confidence intervals; IV=inverse variance



What do we need from the study report?

- measure of effect in each group (e.g. mean, proportion of events)
 - with estimate of variability (e.g. SD, SE, 95%CI)

OR

- ratio of effects in the groups (for dichotomous)
 - with variability (e.g. 95% CI)
- difference in effects (for continuous)
 - with variability (e.g. 95% CI)

Other data types

- time to event
- multiple episodes of the event
- cluster and crossover randomised trials

What can we do here?

Congestive Heart Failure

Effects of carvedilol early after myocardial infarction: Analysis of the first 30 days in Carvedilol Post-Infarct Survival Control in Left Ventricular Dysfunction (CAPRICORN)

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... report contained the term *death*, *death*, or *death*. A composite end point of death or nonfatal MI, which was an end point used in the primary CAPRICORN study,⁶ was also evaluated in this analysis. We also assessed the

Cumulative incidence curves for the occurrence of these events were constructed by the Kaplan-Meier method, using a time-to-first-event approach. The analyses of major outcome

Time-to-event - possibilities

- use hazard ratio (and 95% CI)
 - similar interpretation to risk ratio

OR

- convert to dichotomous at a time point
 - e.g. proportion surviving at 36 months
 - (can read from survival curves if necessary)

BUT

- do NOT use mean time to event

Example – time-to-event

Table 3. Adjusted Risk (Hazard Ratios) for Primary and Secondary Outcomes in HF Patients Within CKD and Non-CKD Groups, Based on Treatment With Carvedilol

Outcome	CKD, eGFR ≤ 60 mL/ min/1.73 m ² (Carvedilol Versus Placebo) (n=1293 Versus 1273)		Non-CKD, eGFR >60 mL/ min/1.73 m ² (Carvedilol Versus Placebo) (n=822 Versus 829)	
	HR*	95% CI	HR*	95% CI
All-cause mortality	0.76	0.63–0.93	0.59	0.43–0.81
Cardiovascular mortality	0.77	0.62–0.94	0.59	0.42–0.82
HF mortality	0.68	0.52–0.88	0.58	0.34–0.99
First hospitalization for HF	0.74	0.62–0.88	0.83	0.63–1.09
Composite of cardiovascular mortality or hospitalization for HF	0.75	0.65–0.87	0.78	0.63–0.98
Sudden cardiac death	0.76	0.56–1.05	0.58	0.37–0.92

*HR is based on the Cox model after adjusting for treatment arm and study type.

There was no significant interaction of treatment and CKD/non-CKD for any of the outcomes.

What can we do here?

	All Exercisers (N = 53)	All Stretchers (N = 62)
Number of colds over 12 mo	n (%)	n (%)
0	26 (49)	19 (31)
1	11 (21)	15 (24)
2	3 (6)	13 (21)
3	2 (4)	2 (3)
Number of colds between	RR (95% CI)	
0-3 mo	0.62 (0.24-1.59)	
3-6 mo	1.26 (0.56-2.86)	
6-9 mo	0.51 (0.20-1.28)	
9-12 mo	0.32 (0.13-0.81)	
P value†	0.02	
Number of upper respiratory tract infections‡ over 12 mo	n (%)	n (%)
0	14 (26)	11 (18)
1	9 (17)	18 (29)
2	11 (21)	11 (18)
3	1 (2)	6 (10)
4	5 (9)	3 (5)
5	1 (2)	0 (0)
6	1 (2)	0 (0)
Number of upper respiratory tract infections‡ between	RR (95% CI)	
0-3 mo	0.95 (0.49-1.86)	
3-6 mo	2.19 (1.04-4.61)	
6-9 mo	0.94 (0.58-1.55)	
9-12 mo	0.71 (0.39-1.28)	
P value†	0.16	

Multiple episodes - possibilities

- use rate ratio (and 95% CI)
 - similar interpretation to risk ratio
- calculate rate ratio
 - need follow-up time (average in each group)
 - assumes constant risk over time
 - estimate standard error using number of events

OR

- use mean number of episodes if common event
 - methods for continuous data
 - need standard deviation, 95%CI etc.

What can we do here?

Table 5 Incidence and Duration of URTI *

Variable	EG (N = 16)	CG (N = 16)
URS (No.)	4	4
duration of incidence (days)	5.3 ± 1.5	6.3 ± 2.2

M ± SD

*p > 0.05

What can we do here?

Effectiveness of multifaceted educational programme to reduce antibiotic dispensing in primary care: practice based randomised controlled trial

OPEN ACCESS

Table 3 | Effects of interventions on antibiotic prescribing at index consultation and antibiotic prescribing and reconsultation during 28 days' follow-up

Variables	Intervention groups		Control groups		P value†	Intraclass coefficient
	No of patients	Percentage (crude 95% CI*)	No of patients	Percentage (crude 95% CI*)		
C reactive protein test:	n=227		n=204			
Antibiotics at index consultation	70	30.8 (21.8 to 39.8)	108	52.9 (43.0 to 62.8)	0.02	0.12
Antibiotics at days 1 to 28	102	44.9 (35.2 to 54.6)	119	58.3 (48.5 to 68.1)	<0.01	0.12
Reconsultation within 28 days	79	34.8 (28.3 to 41.3)	62	30.4 (23.8 to 37.0)	0.50	0.01
Communication skills training:	n=201		n=230			
Antibiotics at index consultation	55	27.4 (25.6 to 36.6)	123	53.5 (43.8 to 63.2)	<0.01	0.12
Antibiotics at days 1 to 28	76	37.8 (28.1 to 47.5)	145	63 (53.6 to 72.4)	<0.001	0.12
Reconsultation within 28 days	56	27.9 (21.4 to 34.4)	85	37.0 (30.4 to 43.6)	0.14	0.01

*Calculated and inflated for clustering by using standard deviation inflated by variance inflation factor.⁵³

†Calculated from second order penalised quasi-likelihood multilevel logistic regression model adjusted for variance at general practitioner and practice level (random intercept at practice and general practitioner level). Models included both interventions and interaction term of interventions. See web extra for corresponding β coefficients.

Including Cluster RCTs

- design effect:

$$1 + (M - 1)ICC$$

M=average cluster size, ICC=intraclass correlation coefficient

- can “borrow” ICC from similar study
- effective sample size:

$$\frac{\textit{sample size}}{\textit{design effect}}$$

- dichotomous outcome: divide events and totals by design effect (may need to round to whole no.)
- continuous outcome: divide totals by design effect

What can we do here?

Table 3| Re-consultation rates for respiratory tract infections within seven, 14, and 31 days in practices undergoing educational programme aimed at reducing antibiotic prescribing in primary care

Re-consultation rate*	Median (IQR)		Median difference (95% CI‡)	P value§
	Intervention (n=20)†	Control (n=17)†		
Within 7 days	2.66 (1.88-4.25)	3.35 (2.16-4.31)	-0.65 (-1.69 to 0.55)	0.446
Within 14 days	5.10 (4.70-7.92)	6.43 (4.04-7.84)	-1.33 (-2.12 to 0.74)	0.411
Within 31 days	9.06 (7.53-12.62)	11.38 (7.39-14.05)	-2.32 (-4.76 to 1.95)	0.503

IQR=interquartile range.

*Median No of individuals who reconsulted for respiratory tract infection per 1000 registered patients.

†Values in each group refer to subset of intervention practices for which data on reconsultation were available.

‡Computed with bootstrapping methods.

§From Mann-Whitney U test.

What can't we meta-analyse?

- Medians
 - IQR is not a 'statistical' measure of variability
 - i.e. can't be used for weighting studies in MA

Summary

- often possible to use measures other than proportion and mean
- need to understand 'generic inverse variance'
- generally $\ln(\text{measure})$ and $\ln(\text{SE})$
 - (natural logarithm)
- consult a statistician
- don't discard 'weird' data immediately!