



MONASH University

Medicine, Nursing and Health Sciences

# ***Challenges with continuous outcomes (part A)***

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Birmingham, UK, 17-18<sup>th</sup> March 2016

# Session plan

1. Combining estimates from analyses of final values, change scores, and ANCOVA  
*Practical (group discussion)*
2. Meta-analysis of skewed data  
*Practical (computer)*
3. Ratio of arithmetic means  
*Practical (computer)*

# Effect measures for continuous outcomes

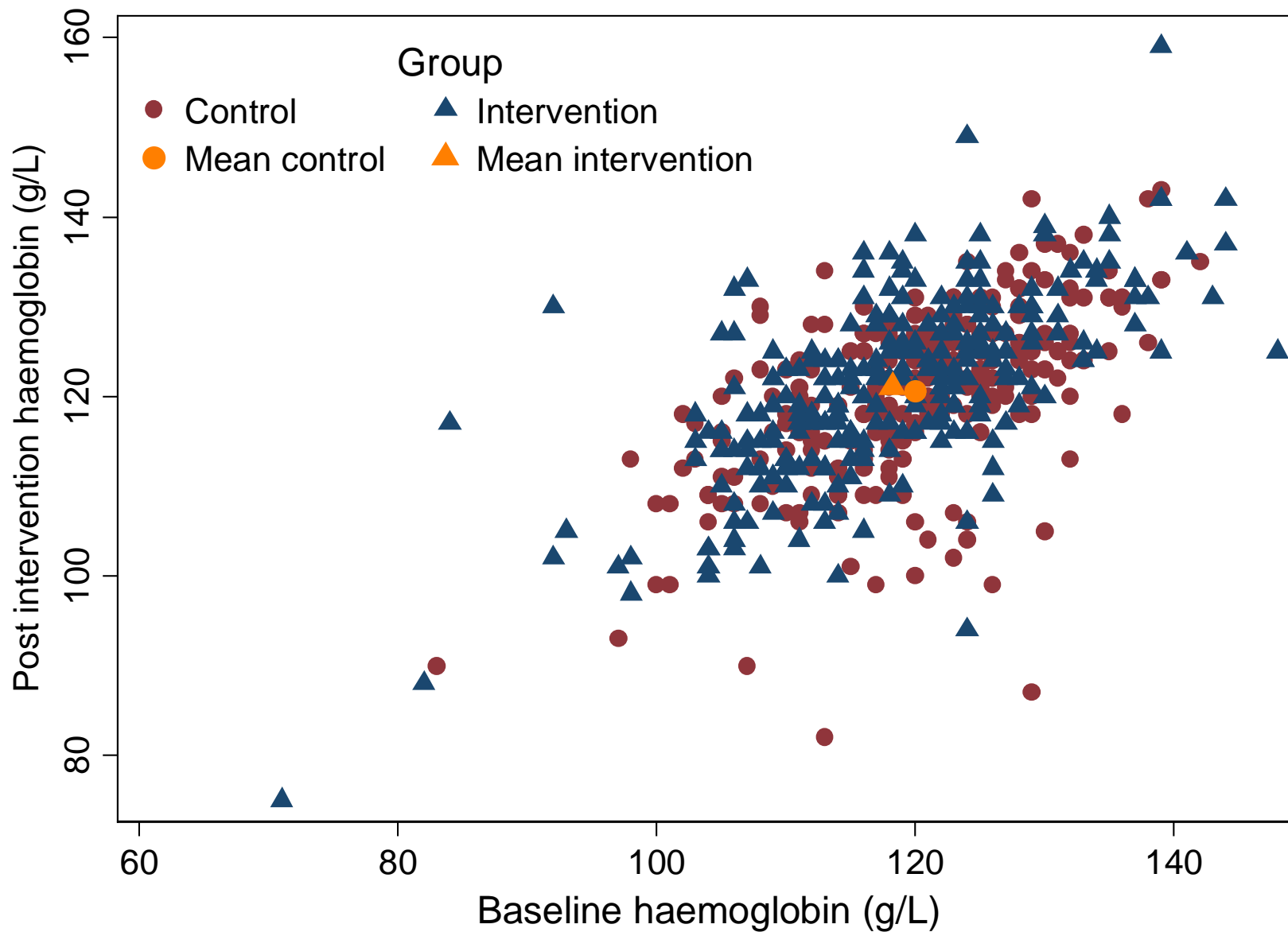
Effect measure	
<i>Difference effect measures</i>	
Mean difference	$MD = \bar{Y}_{int} - \bar{Y}_{ctrl}$
Standardised mean difference	$SMD = \left( \frac{\bar{Y}_{int} - \bar{Y}_{ctrl}}{S_{pooled}} \right)$
<i>Ratio effect measures</i>	
Ratio of means	$RoM = \frac{\bar{Y}_{int}}{\bar{Y}_{ctrl}}$
Ratio of geometric means	$RoGM = \frac{G_{int}}{G_{ctrl}}$

# Combining estimates from analyses of final values, change scores, and ANCOVA

## Analysis of a randomised trial with measurements before and after intervention

- Randomised trial carried out in the Ubon Ratchathani province NE Thailand
- Aimed to test the efficacy of a seasoning powder fortified with micronutrients
- Groups:
  - Intervention: fortified seasoning powder added to instant wheat noodles or rice
  - Control: unfortified seasoning powder added to instant wheat noodles or rice
- Data collected at baseline and follow-up (31 weeks)
- Primary outcome was anaemia (defined from the continuous variable haemoglobin)

## Post intervention haemoglobin vs baseline haemoglobin



# Analysis options

- Ignore the baseline values, and calculate the difference in means at follow-up between groups (simple analysis of final values SAFV)

$$\hat{\theta}_{SAFV} = \bar{Y}_{int} - \bar{Y}_{ctrl}$$

- Adjust for baseline by calculating the difference in mean change ( $Y - X$ ) between groups (simple analysis of change scores SACS)

$$\hat{\theta}_{SACS} = (\bar{Y}_{int} - \bar{Y}_{ctrl}) - (\bar{X}_{int} - \bar{X}_{ctrl})$$

- Adjust for baseline using regression modelling (ANCOVA)

$$\hat{\theta}_{ANCOVA} = (\bar{Y}_{int} - \bar{Y}_{ctrl}) - \hat{\beta}(\bar{X}_{int} - \bar{X}_{ctrl})$$

where  $\beta = \rho \frac{\sigma_Y}{\sigma_X}$  and is estimated from the regression model

# Data sets used to illustrate the impact of correlation on intervention effect estimates

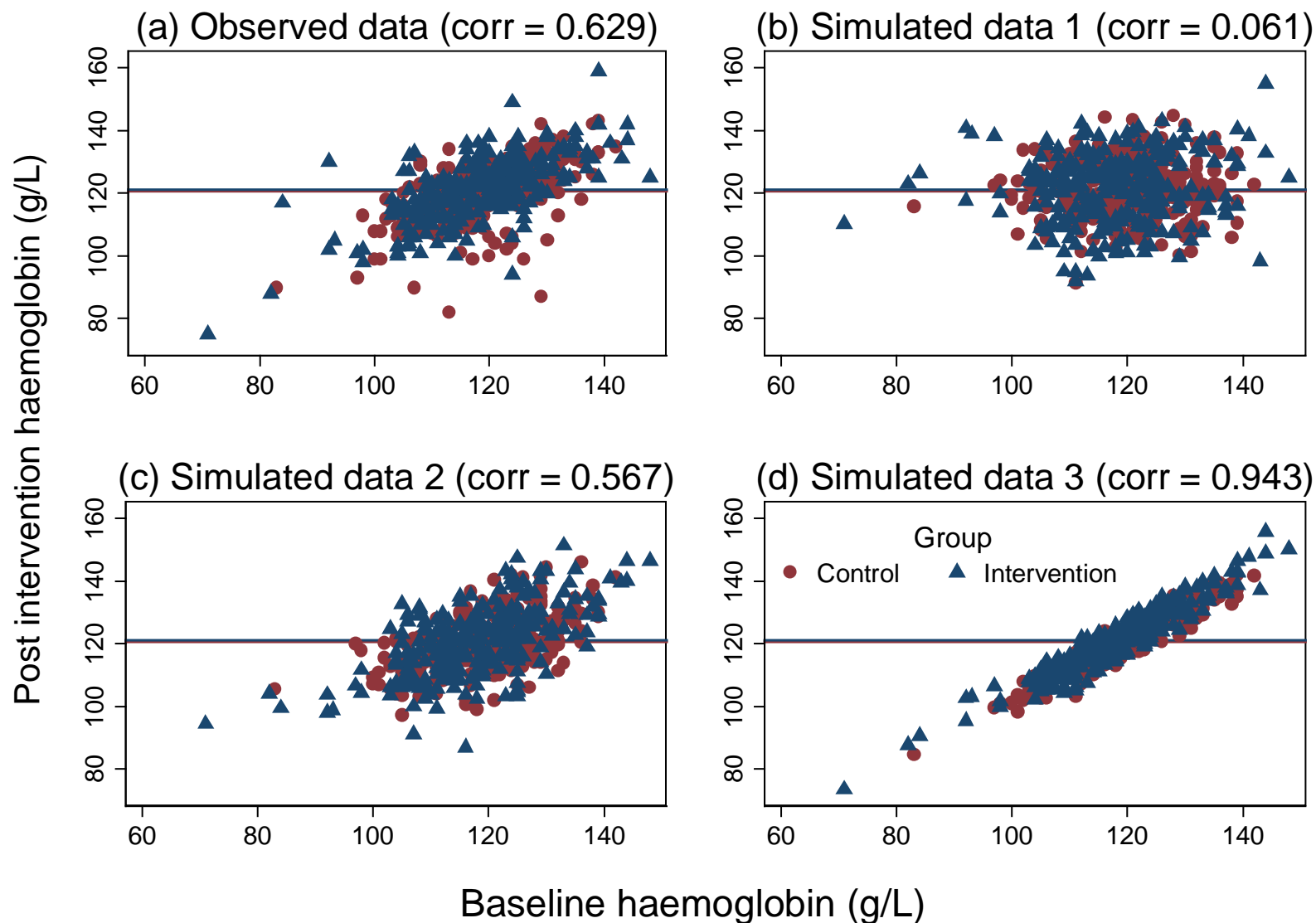
Dataset	Observed correlation	Follow-up haemoglobin (g/L)			
		Intervention group		Control group	
		Mean	SD	Mean	SD
Observed data	0.629	121.0	10.1	120.5	9.5



# Data sets used to illustrate the impact of correlation on intervention effect estimates

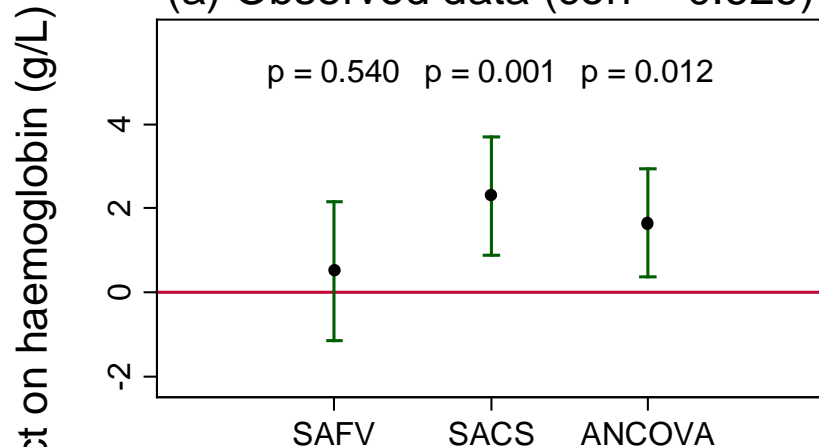
Dataset	Observed correlation	Follow-up haemoglobin (g/L)			
		Intervention group		Control group	
		Mean	SD	Mean	SD
Observed data	0.629	121.0	10.1	120.5	9.5
Simulated data 1	0.061	121.2	10.8	120.6	8.8
Simulated data 2	0.567	121.2	10.8	120.6	8.8
Simulated data 3	0.943	121.1	10.5	120.5	9.0

# Scatter plots of post intervention haemoglobin vs baseline haemoglobin for observed and simulated data sets

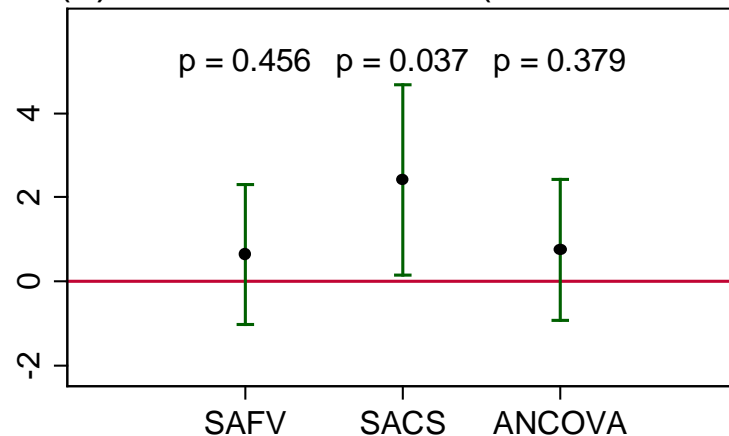


# Estimated intervention effect estimates (95% CIs) calculated using different analytical methods for the four data sets

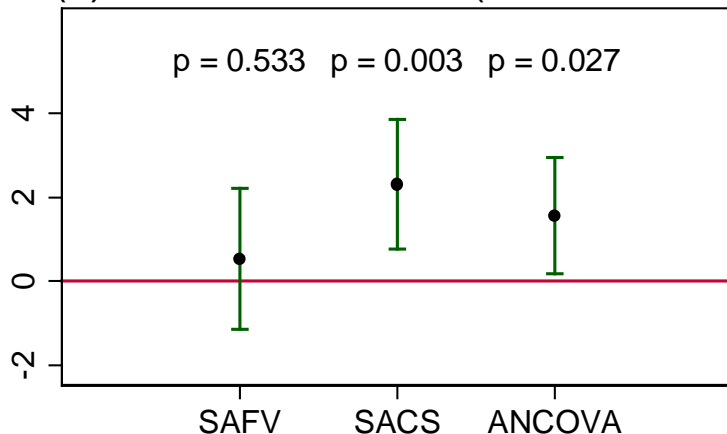
(a) Observed data (corr = 0.629)



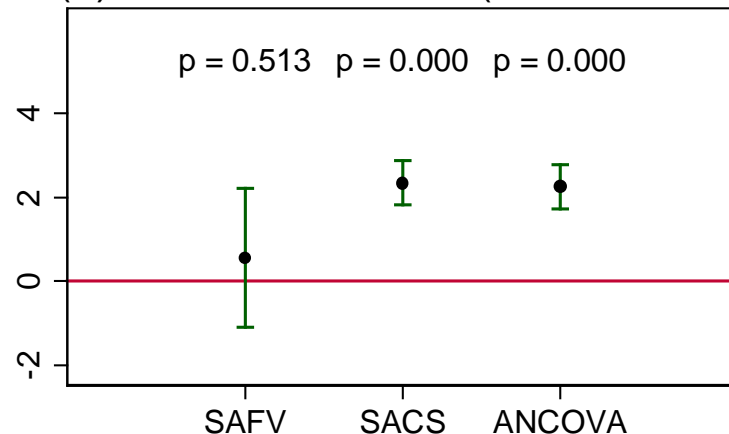
(b) Simulated data 1 (corr = 0.061)



(c) Simulated data 2 (corr = 0.567)

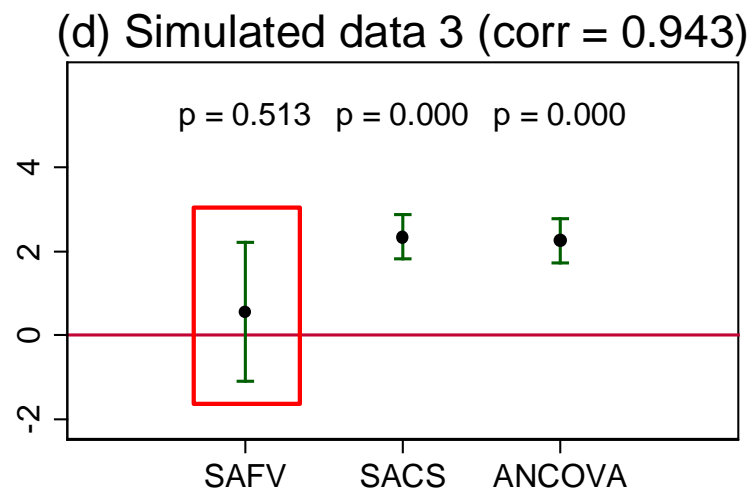
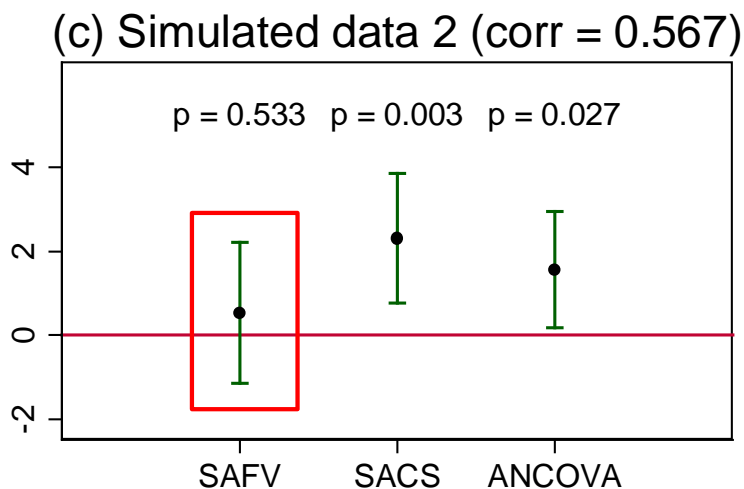
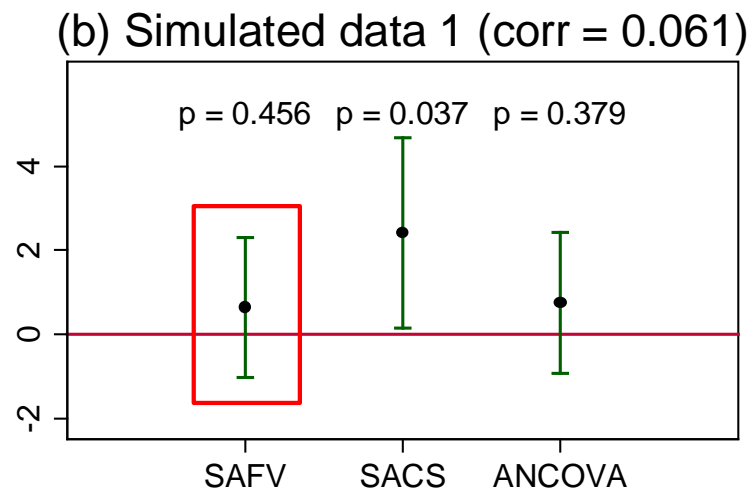
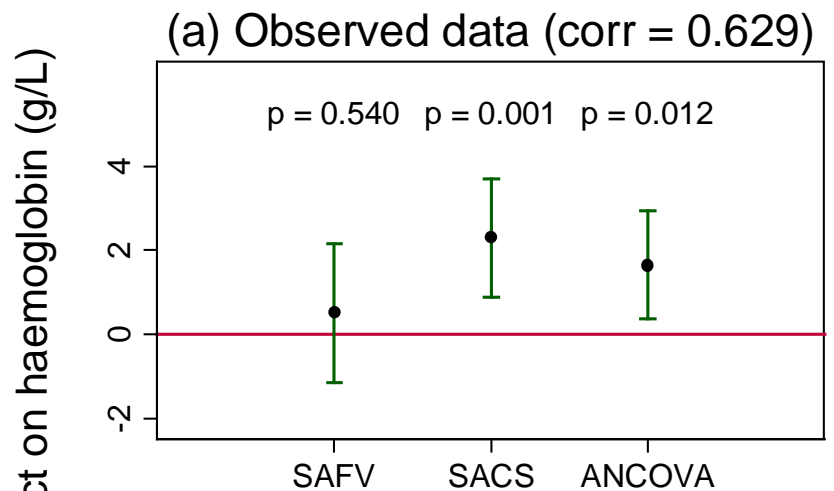


(d) Simulated data 3 (corr = 0.943)



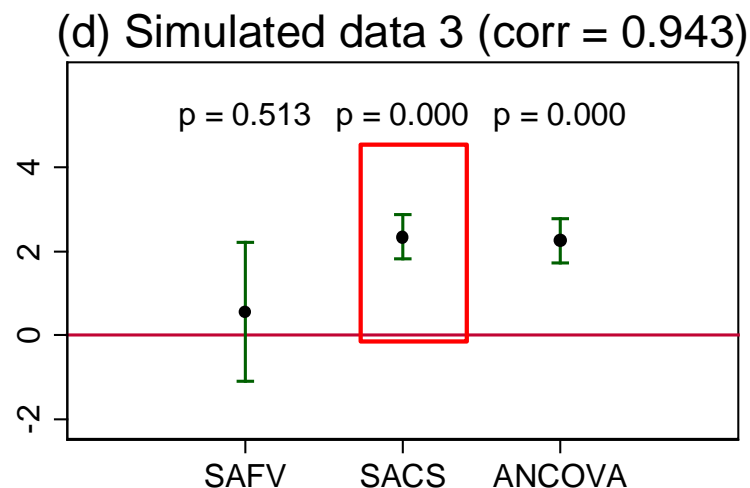
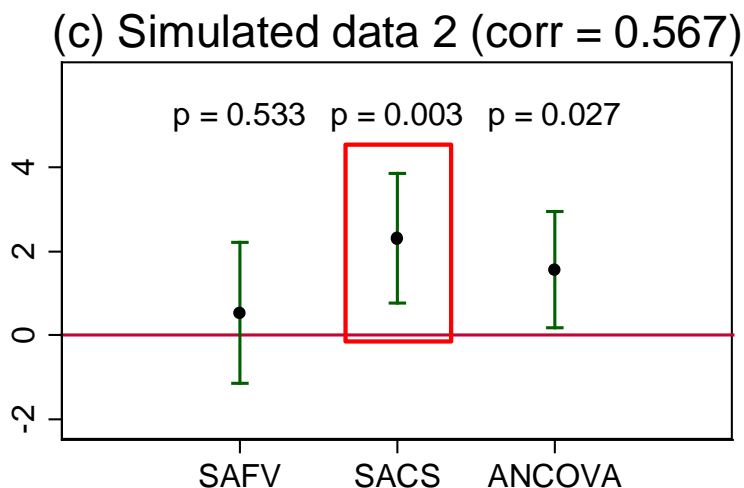
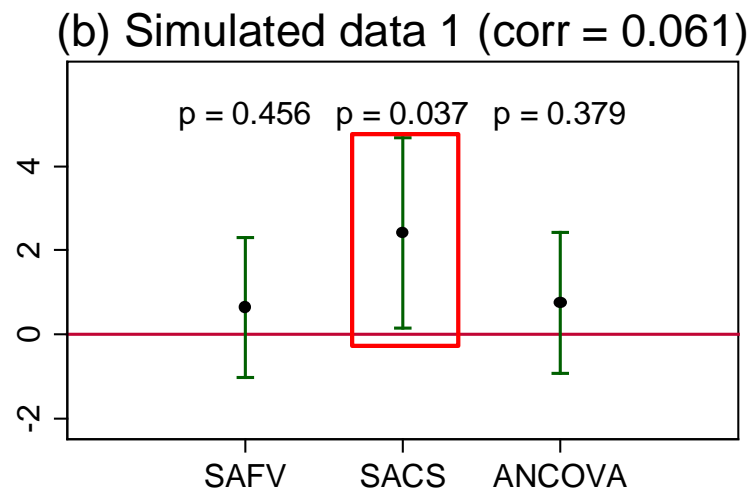
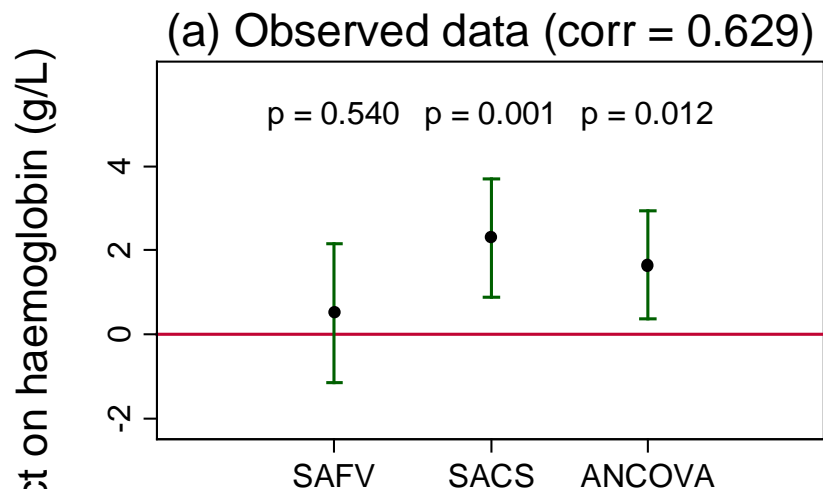
Analytical method

# Estimated intervention effect estimates (95% CIs) calculated using different analytical methods for the four data sets



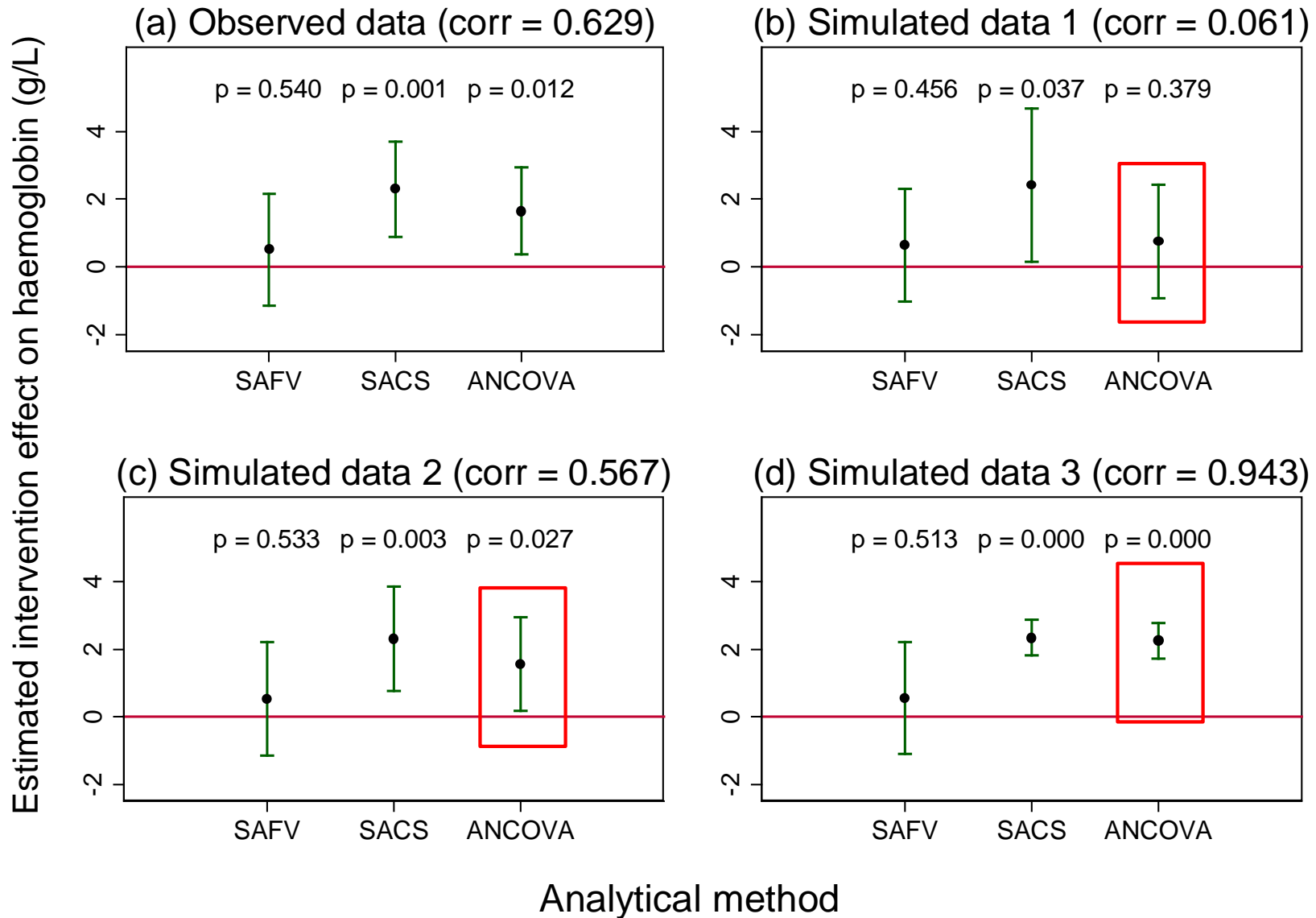
Analytical method

# Estimated intervention effect estimates (95% CIs) calculated using different analytical methods for the four data sets



Analytical method

# Estimated intervention effect estimates (95% CIs) calculated using different analytical methods for the four data sets



# Comparing the trial analysis methods

- Estimates of intervention effect:
  - For a particular data set, the three analytical methods can produce different estimates of intervention effect
  - Over the data sets (varying correlation), the ANCOVA estimate varies; SACS or SAFV estimates do not
- Standard errors:
  - The SE of the SAFV estimate is not affected by correlation
  - Increasing correlation results in a smaller SE for the SACS estimate
  - Correlation  $< 0.5$ , the SE of SACS estimate is  $>$  SE of the FV estimate. This is reversed when the correlation is  $> 0.5$
  - For a particular correlation, the SE of the ANCOVA estimate is smaller compared with SEs of FV and CS estimates

## Relationship between the three analysis methods

$$\hat{\theta}_{ANCOVA} = (\bar{Y}_{int} - \bar{Y}_{ctrl}) - \rho(\bar{X}_{int} - \bar{X}_{ctrl})$$

(assuming  $\sigma_Y^2 = \sigma_X^2$ )

- Scenario 1:  $\rho$  is close to 0

$$\hat{\theta}_{ANCOVA} \approx \hat{\theta}_{SAFV}$$

- Scenario 2:  $\rho$  is close to 1

$$\hat{\theta}_{ANCOVA} \approx \hat{\theta}_{SACS}$$

- Scenario 3: minimal baseline imbalance, i.e.  $(\bar{X}_{int} - \bar{X}_{ctrl}) \approx 0$

$$\hat{\theta}_{ANCOVA} \approx \hat{\theta}_{SACS} \approx \hat{\theta}_{SAFV}$$



# *Practical (group discussion)*

# When undertaking a systematic review ...

- Likely to encounter estimates calculated from different analysis methods
- This could include:
  - SACS
  - SAFV
  - Sometimes SACS *and* SAFV
  - Sometimes ANCOVA
- *Practical:* Discussion of a meta-analysis from a systematic review examining the effect of calcium supplementation on body weight (Trowman 2006 Br J Nut)

# Study characteristics (modified table 1) (Trowman 2006 Br J Nut)

Study	Number of participants	Age*	Sex	Intervention (Ca concentration)	Length of follow-up	Country
Chee et al. (2003)	173	58.9	Female (postmenopausal)	Ca supplement (1200 mg/d)	24 months	Malaysia
Jensen et al. (2001)	52	NA	Female (obese postmenopausal)	Ca supplement (1000 mg/d)	26 weeks	Denmark
Lau et al. (2001)	185	57.0	Female (postmenopausal)	Ca supplement (800 mg/d)	24 months	China
Reid et al. (2002)	223	72.0	Female (postmenopausal)	Ca supplement (1000 mg/d)	24 months	New Zealand
Shapses et al. (2004)	36	59.3	Female (obese postmenopausal)	Ca supplement (1000 mg/d)	25 weeks	USA
Shapses et al. (2004)	30	56.0	Female (obese postmenopausal)	Ca supplement (1000 mg/d)	25 weeks	USA
Shapses et al. (2004)	42	41.0	Female (obese postmenopausal)	Ca supplement (1000 mg/d)	25 weeks	USA
Winters-Stone & Snow (2004)	23	24.8	Female (athletes)	Ca supplement (1000 mg/d)	12 months	USA
Zemel et al. (2004)	41	46	Mixed (obese)	Calcium supplement (800 mg/d)	24 weeks	USA

NA, not available

\* Mean age. When age was reported separately by subgroups, the mean between the groups was calculated.

# Calcium supplementation on body weight (Trowman 2006 Br J Nut)

Trial	Year	Baseline (weight kg)				Follow-up (weight kg)		Change (weight kg)	
		Intervention		Control		Intervention	Control	Intervention	Control
		N	Mean (SD)	n	Mean (SD)	Mean (SD)	Mean (SD)	Mean (SD)	Mean (SD)
Chee	2003	91	56.1 (8.9)	82	57.2 (9.4)			0.0 (2.6) <sup>a</sup>	0.2 (2.6) <sup>a</sup>
Jensen	2001	25	94.6 (14.0) <sup>a</sup>	27	93.8 (14.0) <sup>a</sup>	89.0 (12.7) <sup>a</sup>	89.1 (14.7) <sup>a</sup>		
Lau	2001	95	56.9 (7.1)	90	58.9 (7.5)			0.5 (2.6) <sup>a</sup>	-0.3 (2.7) <sup>a</sup>
Reid	2002	111	66.0 (10.0)	112	68.0 (11.0)			-0.3 (1.8)	-0.1 (2.4)
Shapses 1 <sup>c</sup>	2004	17	84.1 (9.4)	19	89.4 (10.3)			-7.0 (4.6)	-7.3 (5.3)
Shapses 2 <sup>c</sup>	2004	11	85.9 (9.2)	11	94.2 (15.7)			-6.7 (2.6)	-7.6 (5.7)
Shapses 3 <sup>c</sup>	2004	18	93.7 (13.6)	24	93.5 (14.3)			-6.7 (5.5)	-4.3 (3.5)
Winters-Stone	2004	13	57.2 (4.9)	10	54.1 (7.2)	56.3 (4.3)	54.8 (7.2)		
Zemel	2004	11	99.8 (14.9)	10	103.1 (19.3)			-8.6 (5.3) <sup>a</sup>	-6.6 (8.2) <sup>a</sup>

a Calculated from the standard error

b Follow-up sample size ntrt = 24 and nctrl = 24

c Shapses *et al* (Shapses *et al*, 2004) report on three randomised controlled trials.

Trials 1, 2, and 3 include postmenopausal women, postmenopausal women special diet, and premenopausal women respectively.

# Practical

- How would you undertake a meta-analysis of this data?
  - What data would you choose, and why?
  - Would you impute any data, and how?

# Meta-analysis options: a proposed hierarchy

Option	Advantages	Disadvantages
<b>1) Individual patient data:</b> Obtain IPD for each trial. Reanalyse using conventional two-step approach, or a more complex approach such as multilevel modelling. ANCOVA would be the method of choice within each trial	<ul style="list-style-type: none"><li>• Avoids the issue of trialists selectively reporting results</li><li>• Able to re-analyse data in a consistent way</li><li>• Potentially adjust for other prognostic factors</li><li>• Can use the most powerful analytical method</li><li>• Do not have to rely on summary data provided in publications</li></ul>	<ul style="list-style-type: none"><li>• Generally not possible to obtain IPD</li></ul>

# Meta-analysis options: a proposed hierarchy

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<b>2) Meta-analysis using only ANCOVA results:</b> Use available ANCOVA estimates. When not available, recreate the estimates from available summary statistics, or imputing missing statistics (e.g. correlations)	<ul style="list-style-type: none"> <li>• Reduce bias from random baseline imbalance across the included randomised trials or from selective reporting of results</li> <li>• May provide greater precision compared with pooling results from SAFV or SACS</li> </ul>	<ul style="list-style-type: none"> <li>• Will generally require assumptions to be made regarding the correlation</li> <li>• Will require data manipulation</li> </ul>

# Meta-analysis options: a proposed hierarchy

Option	Advantages	Disadvantages
<b>3) Meta-analysis using results from only one analysis method (SAFV or SACS):</b> Meta-analyse estimates using the same analysis method. This approach may involve imputing missing statistics (e.g. correlations)	<ul style="list-style-type: none"><li>• Removes bias from trialists selectively reporting analyses</li></ul>	<ul style="list-style-type: none"><li>• Can provide a biased pooled estimate when there is baseline imbalance across randomised trials. Generally only a problem when there are a small number of trials with few participants</li><li>• May provide less precision compared with meta-analysing ANCOVA results</li><li>• May require assumptions to be made about missing data</li><li>• Will require data manipulation (generally less than option 2)</li></ul>



# Meta-analysis options: a proposed hierarchy

Option	Advantages	Disadvantages
<b>3) Meta-analysis using results from only one analysis method (SAFV or SACS):</b> Meta-analyse estimates using the same analysis method. This approach may involve imputing missing statistics (e.g. correlations)	<ul style="list-style-type: none"><li>• Removes bias from trialists selectively reporting analyses</li></ul>	<ul style="list-style-type: none"><li>• Can provide a biased pooled estimate when there is baseline imbalance across randomised trials. Generally only a problem when there are a small number of trials with few participants</li><li>• May provide less precision compared with meta-analysing ANCOVA results</li><li>• May require assumptions to be made about missing data</li><li>• Will require data manipulation (generally less than option 2)</li></ul>
<b>4) Meta-analysis using a mix of results from different analysis methods:</b> The meta-analysis may include estimates from SAFV, SACS, and ANCOVA	<ul style="list-style-type: none"><li>• Generally less imputation and data manipulation required</li></ul>	<ul style="list-style-type: none"><li>• Prone to selective reporting of results</li><li>• May provide less precision compared with meta-analysing ANCOVA results</li></ul>

# Calcium supplementation on body weight (Trowman 2006 Br J Nut)

Trial	Year	Baseline (weight kg)				Follow-up (weight kg)		Change (weight kg)	
		Intervention		Control		Intervention	Control	Intervention	Control
		n	Mean (SD)	n	Mean (SD)	Mean (SD)	Mean (SD)	Mean (SD)	Mean (SD)
Chee	2003	91	56.1 (8.9)	82	57.2 (9.4)	56.1 (?)	57.4 (?)	0.0 (2.6) <sup>a</sup>	0.2 (2.6) <sup>a</sup>
Jensen	2001	25	94.6 (14.0) <sup>a</sup>	27	93.8 (14.0) <sup>a</sup>	89.0 (12.7) <sup>a</sup>	89.1 (14.7) <sup>a</sup>	-5.6 (?)	-4.7 (?)
Lau	2001	95	56.9 (7.1)	90	58.9 (7.5)	57.4 (?)	58.6 (?)	0.5 (2.6) <sup>a</sup>	-0.3 (2.7) <sup>a</sup>
Reid	2002	111	66.0 (10.0)	112	68.0 (11.0)	65.7 (?)	67.9 (?)	-0.3 (1.8)	-0.1 (2.4)
Shapses 1 <sup>c</sup>	2004	17	84.1 (9.4)	19	89.4 (10.3)	77.1 (?)	82.1 (?)	-7.0 (4.6)	-7.3 (5.3)
Shapses 2 <sup>c</sup>	2004	11	85.9 (9.2)	11	94.2 (15.7)	79.2 (?)	86.6 (?)	-6.7 (2.6)	-7.6 (5.7)
Shapses 3 <sup>c</sup>	2004	18	93.7 (13.6)	24	93.5 (14.3)	87.0 (?)	89.2 (?)	-6.7 (5.5)	-4.3 (3.5)
Winters-Stone	2004	13	57.2 (4.9)	10	54.1 (7.2)	56.3 (4.3)	54.8 (7.2)	-0.9 (?)	0.7 (?)
Zemel	2004	11	99.8 (14.9)	10	103.1 (19.3)	91.2 (?)	96.5 (?)	-8.6 (5.3) <sup>a</sup>	-6.6 (8.2) <sup>a</sup>

a Calculated from the standard error

b Follow-up sample size n<sub>trt</sub> = 24 and n<sub>ctrl</sub> = 24

c Shapses *et al* (Shapses *et al*, 2004) report on three randomised controlled trials.

Trials 1, 2, and 3 include postmenopausal women, postmenopausal women special diet, and premenopausal women respectively.

### **Option 3 (v1): Meta-analysis of results from only one analysis method**

- Trowman (2006) used this option
- For each trial, estimated treatment effect by calculating the difference in mean follow-up measurements
- For missing SDs at follow-up, they assumed the baseline SD

# Calcium supplementation on body weight (Trowman 2006 Br J Nut)

Trial	Year	Baseline (weight kg)				Follow-up (weight kg)		Change (weight kg)	
		Intervention		Control		Intervention	Control	Intervention	Control
		n	Mean (SD)	n	Mean (SD)	Mean (SD)	Mean (SD)	Mean (SD)	Mean (SD)
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Reid	2002	111	66.0 (10.0)	112	68.0 (11.0)	65.7 (10.0)	67.9 (11.0)	-0.3 (1.8)	-0.1 (2.4)
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Zemel	2004	11	99.8 (14.9)	10	103.1 (19.3)	91.2 (14.9)	96.5 (19.3)	-8.6 (5.3) <sup>a</sup>	-6.6 (8.2) <sup>a</sup>

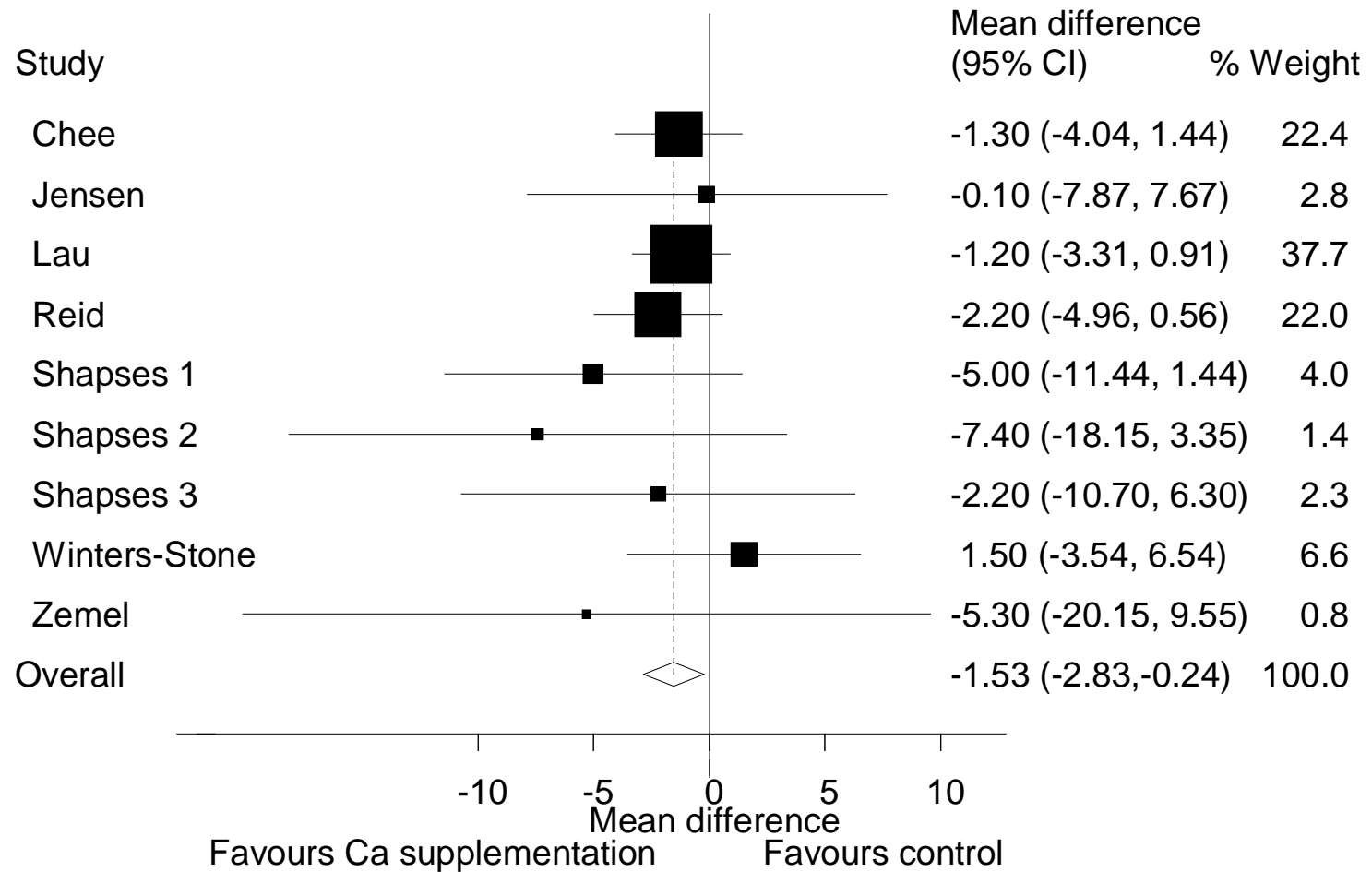
a Calculated from the standard error

b Follow-up sample size n<sub>trt</sub> = 24 and n<sub>ctrl</sub> = 24

c Shapses *et al* (Shapses *et al*, 2004) report on three randomised controlled trials.

Trials 1, 2, and 3 include postmenopausal women, postmenopausal women special diet, and premenopausal women respectively.

# Combining intervention estimates from SAFV only



## Option 3 (v2): Meta-analysis of results from only one analysis method

- Use change scores
  - Impute missing change score SDs, in each intervention group, by calculating the median of the other SDs
- Many other options for imputing missing SDs
  - e.g. in trials with baseline SDs and change SDs, assume follow-up SDs are the same as baseline SDs (seems reasonable assumption based on Jensen 2001 and Winters-Stone 2004) then calculate correlations and SDs at follow-up using

$$Corr_{int} = \frac{SD_{int,X}^2 + SD_{int,Y}^2 - SD_{int,C}^2}{2 \times SD_{int,X} \times SD_{int,Y}}$$

$$SD_{int,C} = \sqrt{SD_{int,X}^2 + SD_{int,Y}^2 - (2corr_{int} \times SD_{int,X} \times SD_{int,Y})}$$

**Wiebe 2006 J Clin Epi; Balk 2012 AHRQ**

# Calcium supplementation on body weight (Trowman 2006 Br J Nut)

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Jensen	2001	25	94.6 (14.0) <sup>a</sup>	27	93.8 (14.0) <sup>a</sup>	89.0 (12.7) <sup>a</sup>	89.1 (14.7) <sup>a</sup>	-5.6 (2.6)	-4.7 (3.5)
Lau	2001	95	56.9 (7.1)	90	58.9 (7.5)	57.4	58.6	0.5 (2.6) <sup>a</sup>	-0.3 (2.7) <sup>a</sup>
Reid	2002	111	66.0 (10.0)	112	68.0 (11.0)	65.7	67.9	-0.3 (1.8)	-0.1 (2.4)
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Winters-Stone	2004	13	57.2 (4.9)	10	54.1 (7.2)	56.3 (4.3)	54.8 (7.2)	-0.9 (2.6)	0.7 (3.5)
Zemel	2004	11	99.8 (14.9)	10	103.1 (19.3)	91.2	96.5	-8.6 (5.3) <sup>a</sup>	-6.6 (8.2) <sup>a</sup>

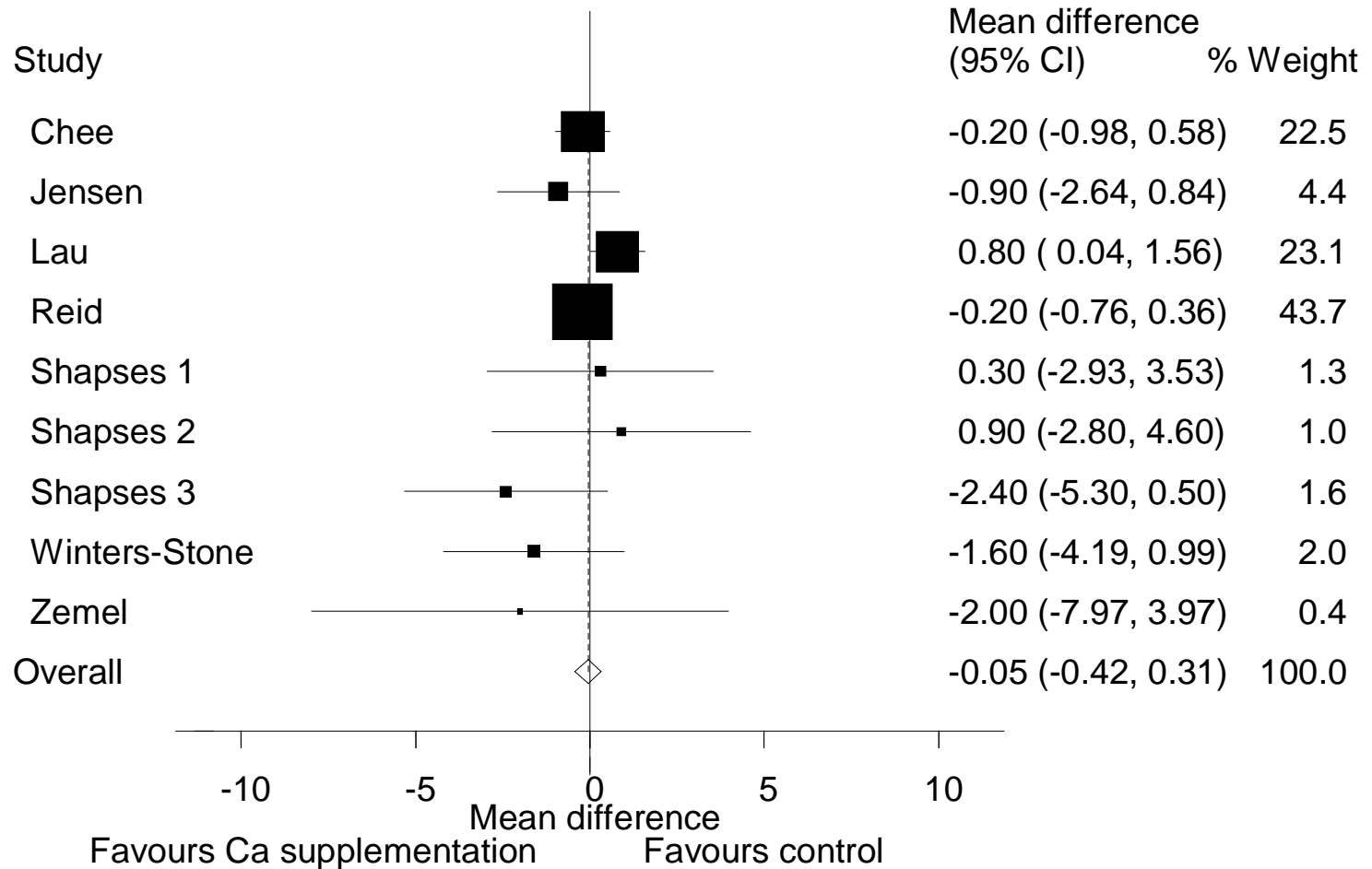
<sup>a</sup> Calculated from the standard error

<sup>b</sup> Follow-up sample size n<sub>trt</sub> = 24 and n<sub>ctrl</sub> = 24


<sup>c</sup> Shapses *et al* (Shapses *et al*, 2004) report on three randomised controlled trials.

Trials 1, 2, and 3 include postmenopausal women, postmenopausal women special diet, and premenopausal women respectively.

# Combining intervention estimates from SACS only







## **Option 4: Meta-analysis using a mix of results from different analysis methods**

# Calcium supplementation on body weight (Trowman 2006 Br J Nut)

Trial	Year	Baseline (weight kg)				Follow-up (weight kg)		Change (weight kg)	
		Intervention		Control		Intervention	Control	Intervention	Control
		N	Mean (SD)	n	Mean (SD)	Mean (SD)	Mean (SD)	Mean (SD)	Mean (SD)
Chee	2003	91	56.1 (8.9)	82	57.2 (9.4)	<b>89.0 (12.7)<sup>a</sup></b>	<b>89.1 (14.7)<sup>a</sup></b>	<b>0.0 (2.6)<sup>a</sup></b>	<b>0.2 (2.6)<sup>a</sup></b>
Jensen	2001	25	94.6 (14.0) <sup>a</sup>	27	93.8 (14.0) <sup>a</sup>				
Lau	2001	95	56.9 (7.1)	90	58.9 (7.5)			<b>0.5 (2.6)<sup>a</sup></b>	<b>-0.3 (2.7)<sup>a</sup></b>
Reid	2002	111	66.0 (10.0)	112	68.0 (11.0)			<b>-0.3 (1.8)</b>	<b>-0.1 (2.4)</b>
Shapses 1 <sup>c</sup>	2004	17	84.1 (9.4)	19	89.4 (10.3)			<b>-7.0 (4.6)</b>	<b>-7.3 (5.3)</b>
Shapses 2 <sup>c</sup>	2004	11	85.9 (9.2)	11	94.2 (15.7)	<b>56.3 (4.3)</b>	<b>54.8 (7.2)</b>	<b>-6.7 (2.6)</b>	<b>-7.6 (5.7)</b>
Shapses 3 <sup>c</sup>	2004	18	93.7 (13.6)	24	93.5 (14.3)			<b>-6.7 (5.5)</b>	<b>-4.3 (3.5)</b>
Winters-Stone	2004	13	57.2 (4.9)	10	54.1 (7.2)				
Zemel	2004	11	99.8 (14.9)	10	103.1 (19.3)			<b>-8.6 (5.3)<sup>a</sup></b>	<b>-6.6 (8.2)<sup>a</sup></b>

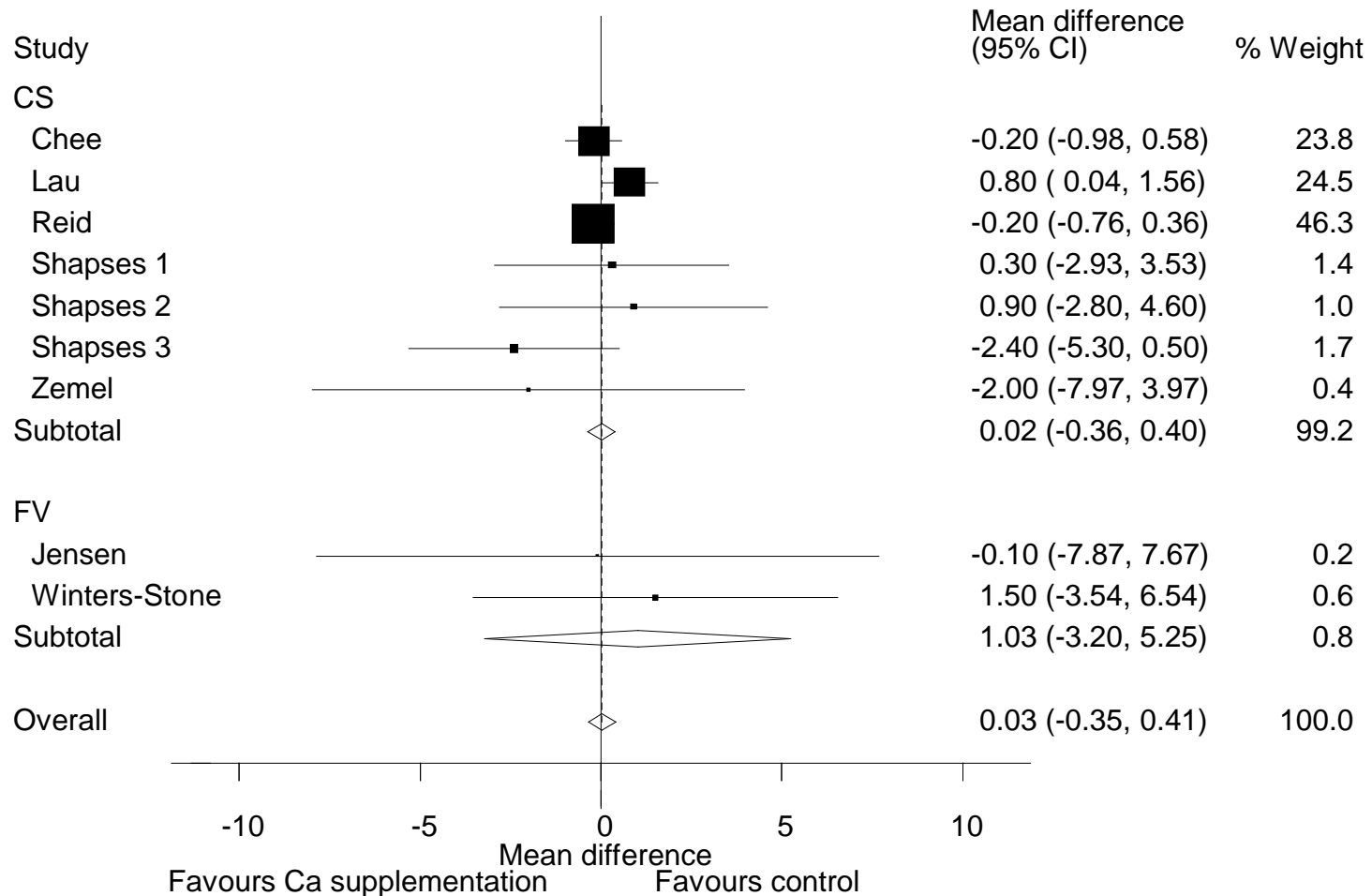
a Calculated from the standard error

b Follow-up sample size ntrt = 24 and nctrl = 24

c Shapses *et al* (Shapses *et al*, 2004) report on three randomised controlled trials.

Trials 1, 2, and 3 include postmenopausal women, postmenopausal women special diet, and premenopausal women respectively.

# Combining intervention estimates from SAFV and SACS



# Which option?

		Meta-analysis options		
Domain		2) Only ANCOVA	3) Only one (SAFV or SACS)	4) Mix (SAFV, SACS, and ANCOVA)
Bias (few small trials)	Chance baseline imbalance across trials	✓		
	Selective reporting	✓	✓	
Precision	No heterogeneity	✓		
	Heterogeneity	=	=	=
Practical issues		✗	✗	✓

# Which option?

- In many circumstances combining estimates calculated from a mix of analysis methods is reasonable (option 4)
  - Include available ANCOVA estimates where possible
- If combining estimates based on only one analytical method (SAFV or SACS), options for choice of method include
  - the analysis method most frequently reported
  - the analysis method that is likely to yield the greatest precision (correlations likely to be large → use SACS; correlations likely to be small → use SAFV)

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# Meta-analysis of skewed data

# Meta-analysis of skewed data

- Standard meta-analytic methods assume normality in the distribution of the means (not raw data)
- Many outcomes are not normally distributed. Examples include:
  - Concentrations
    - e.g. urinary iodine
  - Ratio or reciprocal measures
    - e.g. ratio of partial pressure of arterial oxygen to fraction of inspired oxygen
  - Resource use
    - e.g. length of stay
  - Assessment scales
    - e.g. large proportion of 'normal' participants fall towards one extreme of the scale

*Higgins 2008 Stats in Med*



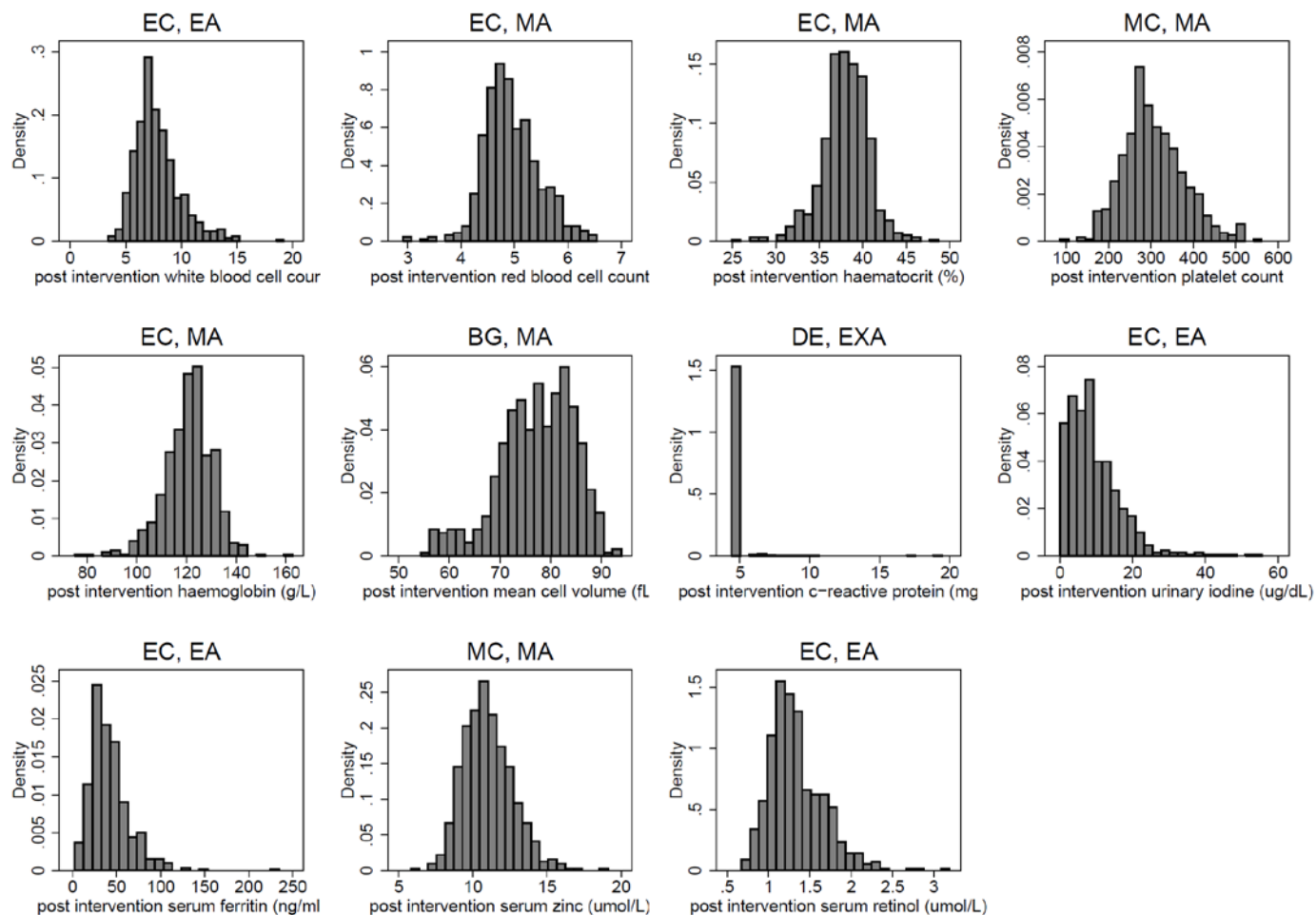
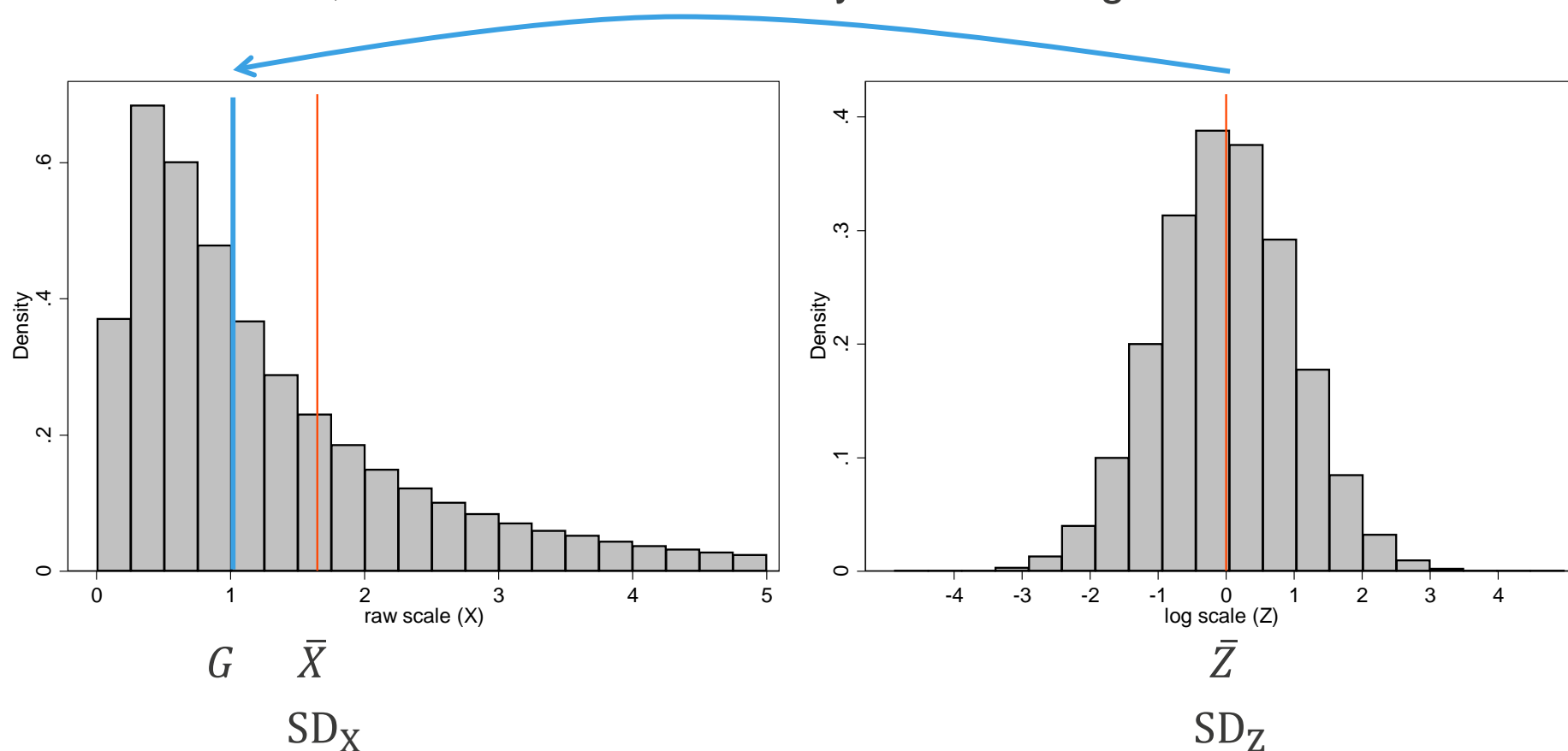


Figure 3: Histograms of post biochemical variables from Thailand RCT

First abbreviation = classification based on tail weight (U = Uniform, BG = Below Gaussian, G = Gaussian, MC = Moderate Contamination, EC = Extreme Contamination, DE = Double Exponential), second abbreviation = classification based on asymmetry (S = Near Symmetry, MA = Moderate Asymmetry, EA = Extreme Asymmetry, EXA = Exponential Asymmetry).

# Dealing with skewed data in studies

Common approach to dealing with skewed data is to log transform the observations, then undertake the analysis on the log scale



# Meta-analysis of skewed data

- Meta-analytic methods are likely to be valid in large trials even when the outcome is skewed, but practical issues arise
- Same outcome: different scales (raw, log), different statistics



- Study 1: Mean, SD, raw scale ( $\bar{X}, SD_X$ )
- Study 2: Mean, SE, raw scale ( $\bar{X}, SD_X/n$ )
- Study 3: Mean, SD, log scale ( $\bar{Z}, SD_Z$ )
- Study 4: Geometric mean, CI ( $G, G_l, G_U$ )
- ...
- Study X

# Meta-analysis of skewed data

- Aim to include as many trials in the one meta-analysis as possible
- To achieve this, we need to transform the summary statistics from one scale to another

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## Meta-analysis of skewed data: Combining results reported on log-transformed or raw scales

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**Method 1:** Transform  $\bar{X}_i$  and  $SD_{X_i}$  within each group

**Method 2:** Transform  $\bar{X}_i$  and  $SD_{X_i}$  assuming a common underlying SD on the log scale

Methods assume  $X$  follows a log normal distribution (i.e.  $Z = \ln(X) \sim N(\mu, \sigma_Z^2)$ ) and utilise standard transformation

Two methods (*ad hoc*, *Taylor series*) for calculating  $SE(D_Z)$

## Scenario 1: Convert raw scale to log scale

Trials requiring  
conversion

$\bar{X}_i, SD_{X_i}$



$\bar{Z}'_i, SD'_{Z_i}$

### Standard result:

If variable  $X$  follows a log normal distribution, then:

$$E(X) = \exp\left(\mu + \frac{\sigma_Z^2}{2}\right)$$

$$\text{var}(X) = (\exp(\sigma_Z^2) - 1)\exp(2\mu + \sigma_Z^2)$$

**Method 1:** Transform  $\bar{X}_i$  and  $SD_{X_i}$  within each group

**Method 2:** Transform  $\bar{X}_i$  and  $SD_{X_i}$  assuming a common underlying SD on the log scale

Methods assume X follows a log normal distribution (i.e.  $Z = \ln(X) \sim N(\mu, \sigma_Z^2)$ ) and utilise standard transformation

Two methods (*ad hoc*, *Taylor series*) for calculating  $SE(D_Z)$

**Method 3:** Targets difference between the groups ( $D_X$ ), rather than group means.

Method does *not* assume X follows a log normal distribution

## Scenario 1: Convert raw scale to log scale

Trials requiring  
conversion

Trials not requiring  
conversion

$\bar{X}_i, SD_{X_i}$

$\bar{Z}'_i, SD'_{Z_i}$

$\bar{Z}_i, SD_{Z_i}$

$D'_Z = \bar{Z}'_2 - \bar{Z}'_1$   
 $SE(D'_Z)$

$D_Z = \bar{Z}_2 - \bar{Z}_1$   
 $SE(D_Z)$

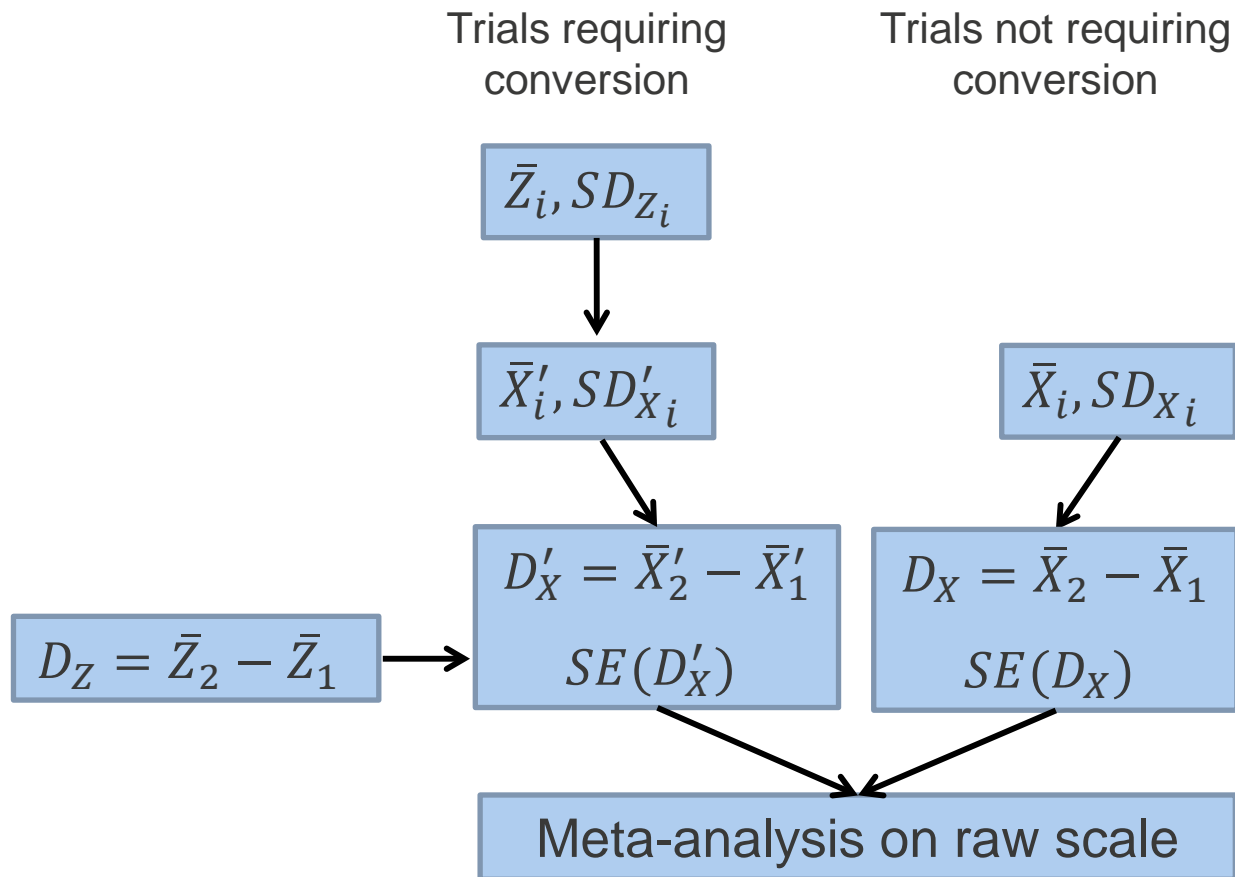
$D_X = \bar{X}_2 - \bar{X}_1$

Meta-analysis on log scale

Exponentiate

Meta-analytic  
ratio of geometric means

## Scenario 2: Convert log scale to raw scale



## Meta-analysis from a systematic review (Sagoo et al)

Examines association between triglyceride level and being a carrier or non-carrier of the D9N polymorphism in the LPL gene.

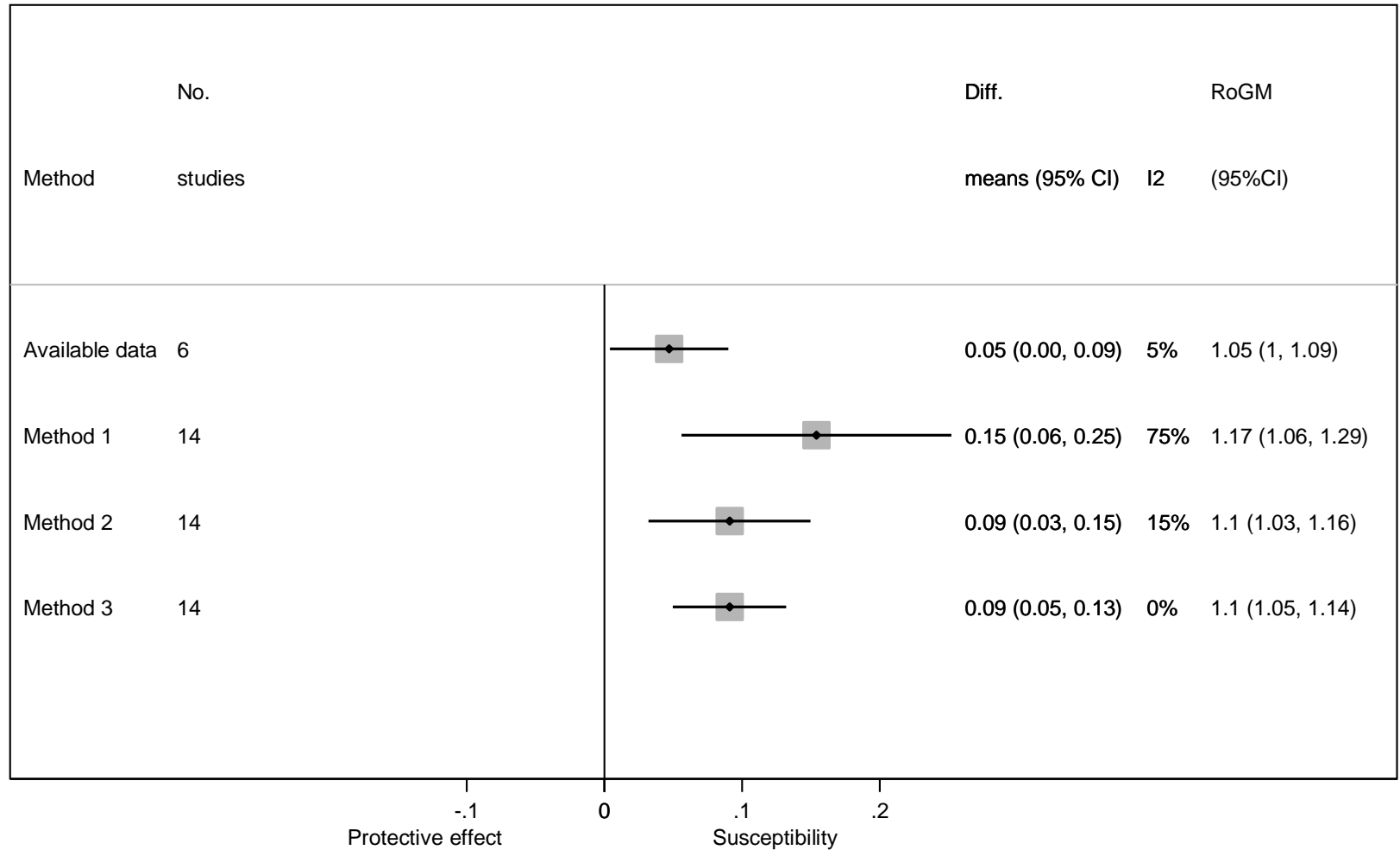
Table I. Data available for D9N polymorphism in the lipoprotein lipase gene and triglyceride levels.

	Carriers					Non-carriers				
	Raw		Log			Raw		Log		
	<i>n</i>	Mean	SD	Mean	SD	<i>n</i>	Mean	SD	Mean	SD
Boer 2003b	34	—	—	0.31	0.58	1002	—	—	0.33	0.53
Copenhagen	241	2.10	1.46	1.05	0.37	8429	1.85	1.54	0.98	0.34
CDRFMP	14	2.05	1.21	—	—	364	1.57	1.11	—	—
EARS I & II	71	1.12	0.34	—	—	1608	0.99	0.80	—	—
ECTIM	22	1.82	1.46	—	—	784	1.84	1.47	—	—
Ehrenborg 1997	15	1.01	0.36	—	—	77	0.99	0.53	—	—
Ferencak 2003	5	2.04	0.92	—	—	195	1.81	0.84	—	—
FOS	58	1.61	0.72	—	—	2200	1.38	1.16	—	—
Glisic 2003b	4	2.42	1.53	0.74	0.60	129	1.64	0.94	0.37	0.49
Reykjavik	10	1.64	1.64	0.20	0.74	274	1.04	0.49	−0.05	0.42
Rios 2003	10	1.60	0.70	0.39	0.41	187	1.75	0.92	0.43	0.50
Schulte 1996	17	1.96	0.82	—	—	644	1.56	0.82	—	—
Talmud 1998	12	1.35	0.52	—	—	96	1.27	0.52	—	—
Yang 2004	235	2.39	1.46	0.74	0.50	1275	2.34	1.26	0.73	0.49



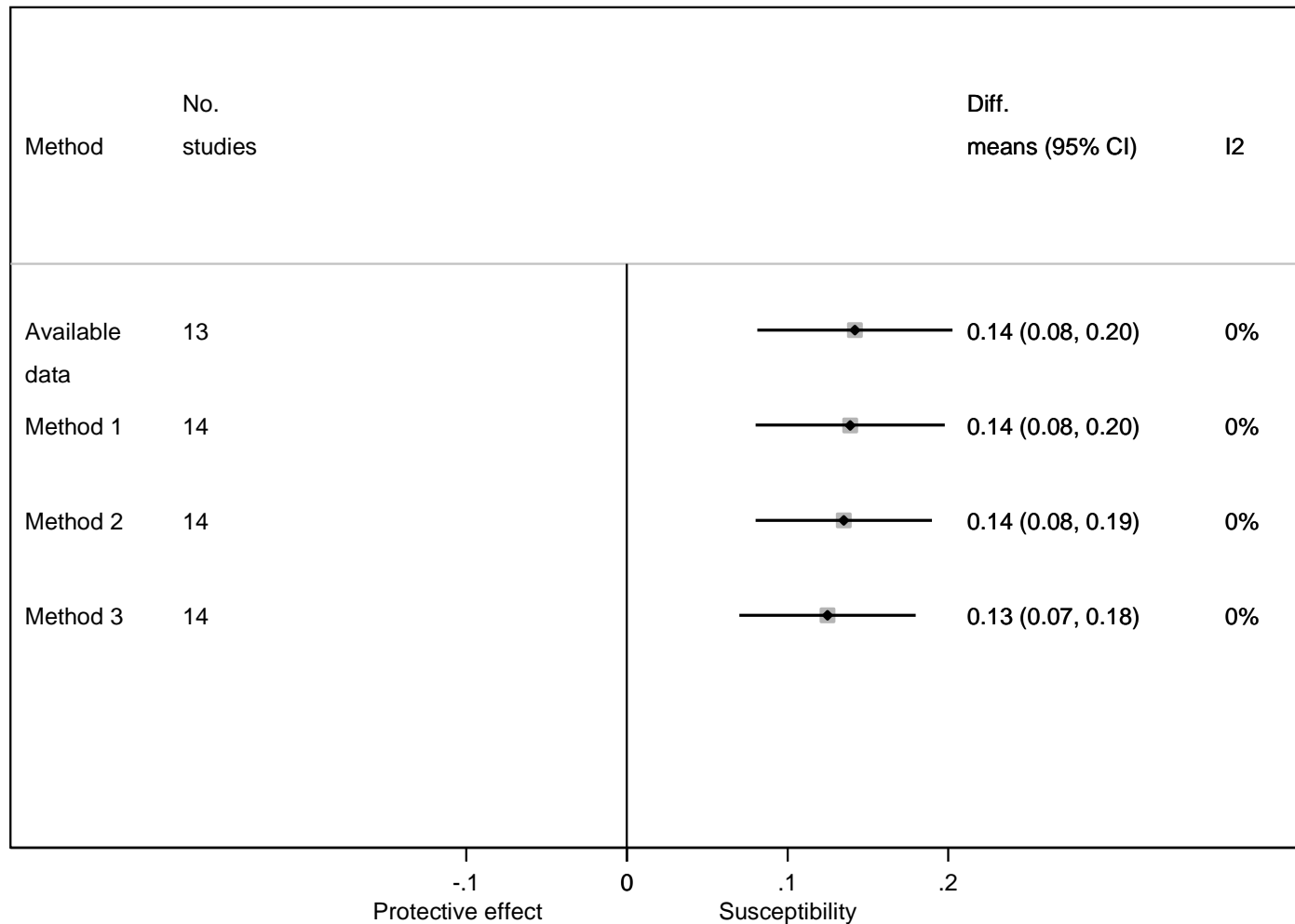
# Raw to log scale

Difference in mean log triglyceride level (Carriers – Non-carriers) using alternative conversions



# Log to raw scale

Difference in mean raw triglyceride level (Carriers – Non-carriers) using alternative conversions



# Which conversion method? Which direction?

## *Which conversion method?*

- Not a uniformly preferable method
- All methods reasonably robust to data having distributions other than log normal
- Method 1 preferable when SDs differ across groups; when SDs are similar, greater precision obtained using Method 2

## *Which direction? Raw to log scale OR log to raw scale?*

- Use the scale most frequently used
- Scale best meeting meta-analytic assumptions (scale believed to be less skewed)
- Meta-analysis on the log-scale may reduce heterogeneity

# Length of intubation (hours)

Study	High dose opioid					Low dose opioid					Significance as reported by authors
	n	Mean	SD	Median	Range	n	Mean	SD	Median	Range	
Slogoff 1989 Enflurane Halothane Isoflurane	254	22.8	12.3			257 253 248	14.5 16.8 14.7	6.3 7.2 5.4			ANOVA p = 0.001
Bell 1994	19			12.96		20			4.42		p = 0.0005
Cheng 1996	51	18.9	1.4			51	4.1	1.1			p < 0.02
Myles 1997	66	21.5	5.1	12.3	3.5-31.5	58	11.4	9.9			p = 0.006
Silbert 1998	42			7.0	2.1-19	38			4.0	0.5-15.5	p < 0.01
Michalopoulos 1998	72	11.6	1.3			72	7.3	0.7			p = 0.0001
Sakaida 1998	20	14.5	4.5	15	6-25.3	20	5.6	1.6	5.2	3.5-9	p < 0.05
Berry 1998	42			12.62	8.32-20.67	43			1.83	0.1-4.25	significant
Myles 2001	24			9.7	1.1-25	24			6.5	0.4-150	?significant

Anesthesiology 2003; 99:982-7

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## *A Systematic Review of the Safety and Effectiveness of Fast-track Cardiac Anesthesia*

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# Dealing with non-parametric statistics

- Assume that the data are (close to) lognormal
  - median  $\sim$  geometric mean
  - Convert centiles to mean and SD on the log scale
    - Converting ranges see: Walter 2007 J Clin Epi; Hozo 2005 BMC Med Res Methodol
    - Converting IQR: Section 7.7.3.5 Cochrane Handbook; Wan 2014 BMC Med Res Methodol
- Combine using conversion approaches in Higgins 2008 Stats Med

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