

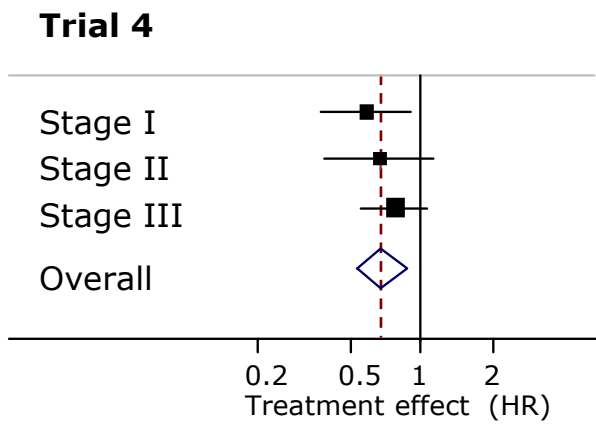
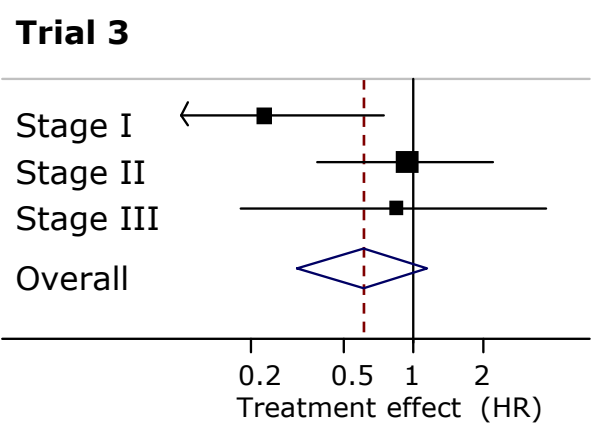
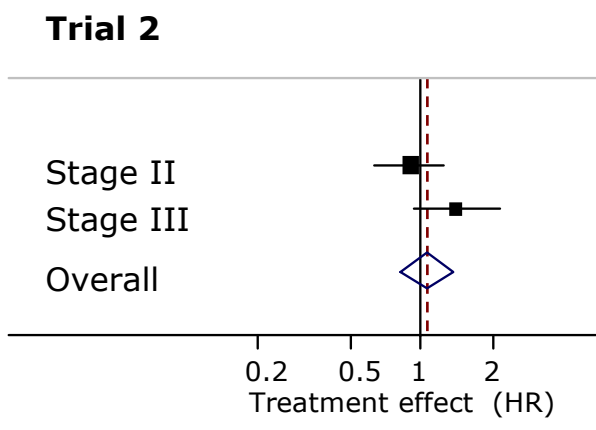
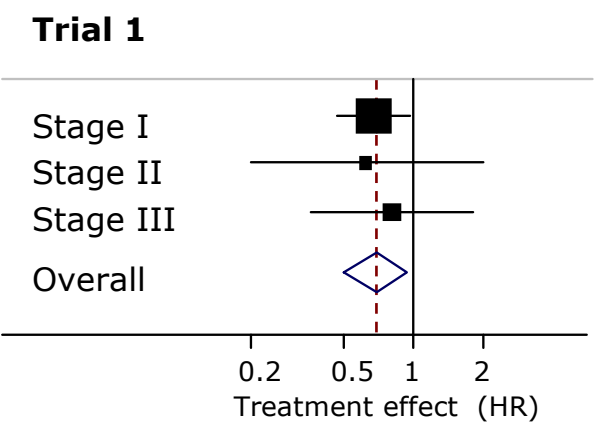
# A critical review of methods for the assessment of interactions in individual patient data meta-analysis

**David Fisher**, Andrew Copas, Jayne Tierney, Mahesh Parmar  
Meta-analysis Group  
MRC Clinical Trials Unit

# Objectives & methods

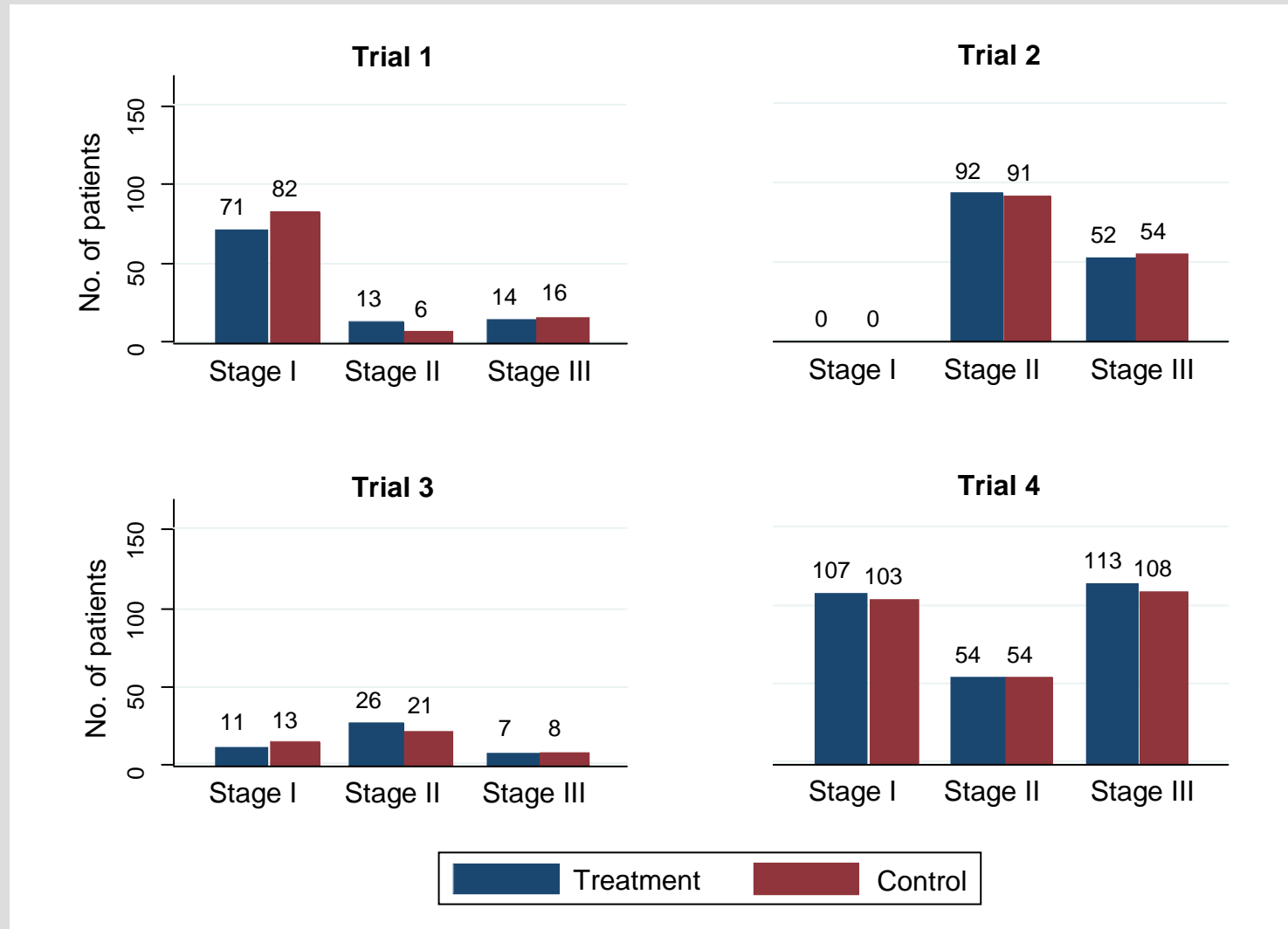
- Individual patient data (IPD) meta-analysis (MA)
  - Treatment, outcome and covariate known for each patient
    - ⇒ Treatment-covariate interactions
    - ⇒ Targeted treatment
- Research questions
  - How should treatment-covariate effects be analysed?
  - How should such analyses be presented?
- Methods & objectives
  - Literature search for existing approaches
  - Critical appraisal of identified approaches
  - Apply these approaches to existing IPD datasets
  - Develop guidance

# Treatment effect by trial and disease stage



- Post-op RT vs surgery alone in non-small-cell lung cancer
- Outcome is overall survival (time-to-event)
- Question: does trt effect differ by disease stage group?
- For simplicity, interaction measured using a linear slope fitted across subgroups

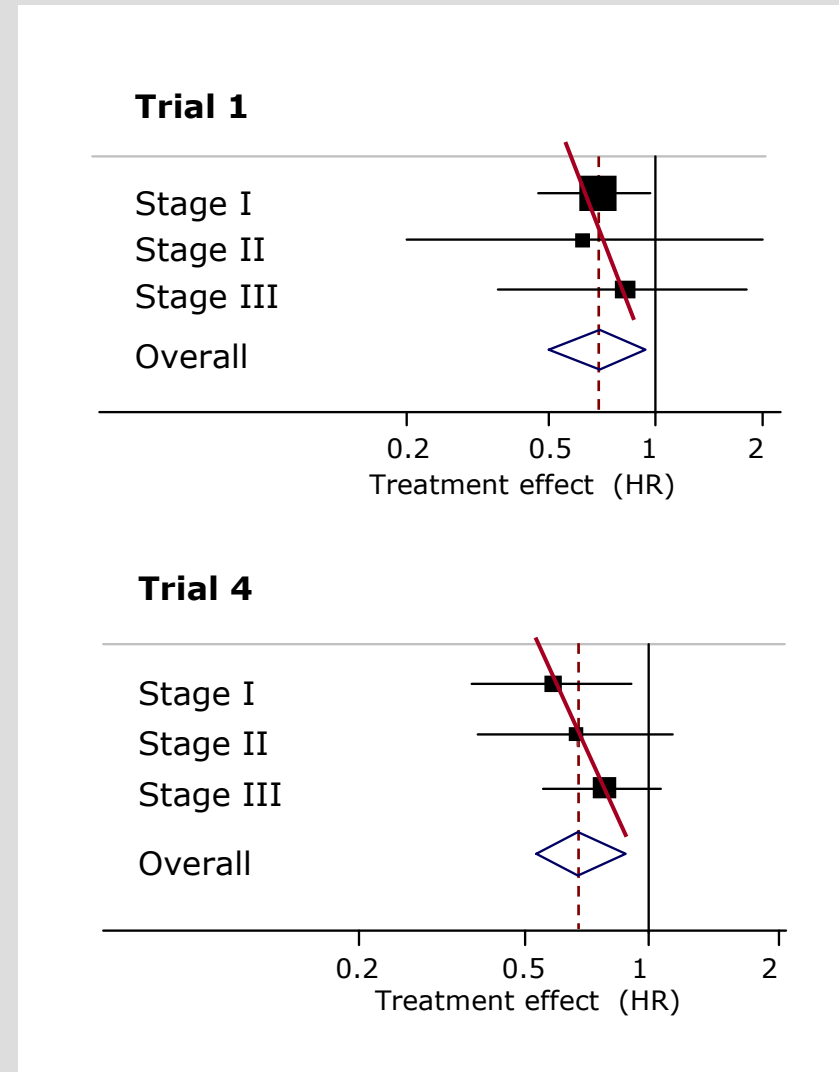
# Distribution of disease stage



# Within-trial effect

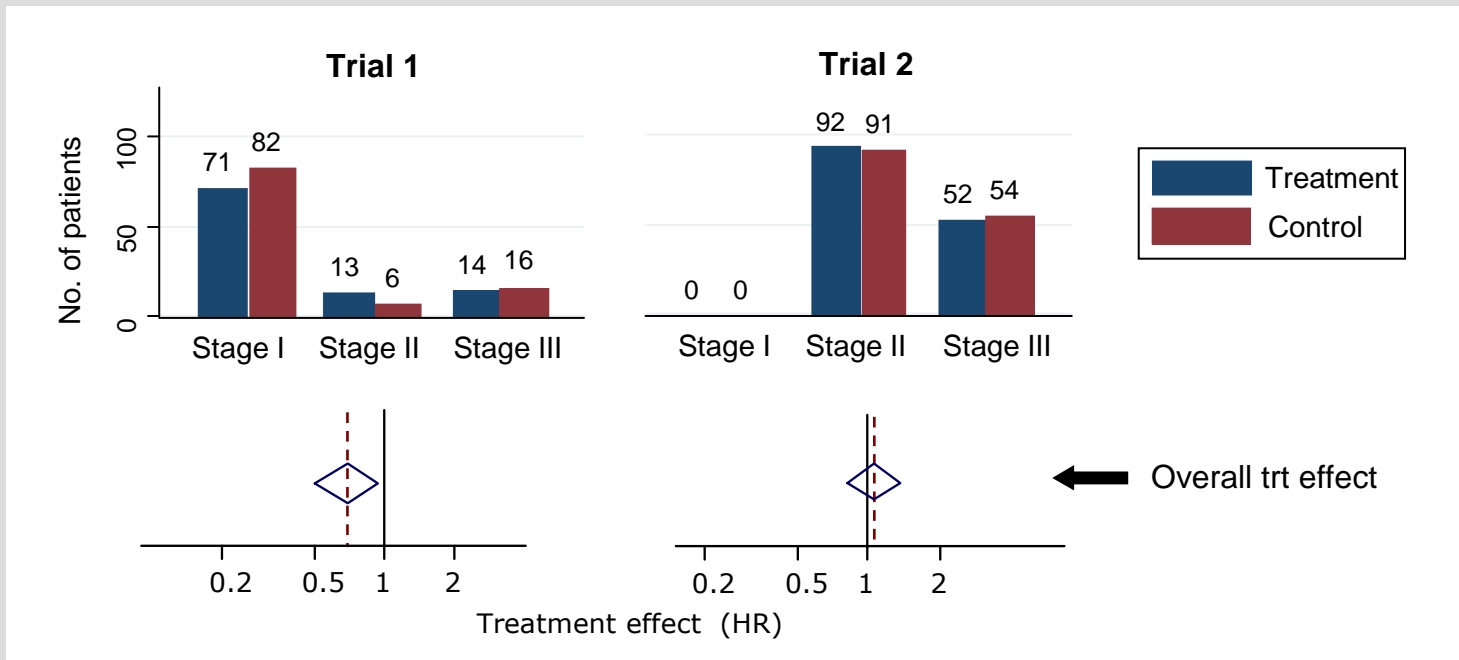
## Within-trial (WT) effect

- Measures how disease stage affects treatment outcome for individual pts within an “average” trial
- Can be visualised as “average” slope
- Patients are randomised to trt arms within disease stage subgroups, all subject to the same trial protocol



# Cross-trial effect

- **Across-trial (AT) effect**
  - Measures how the “average” disease stage in a trial affects treatment outcome
  - **Not randomised, and subject to “ecological bias”**
    - Difference in “average” between trials may not be representative of differences between individuals



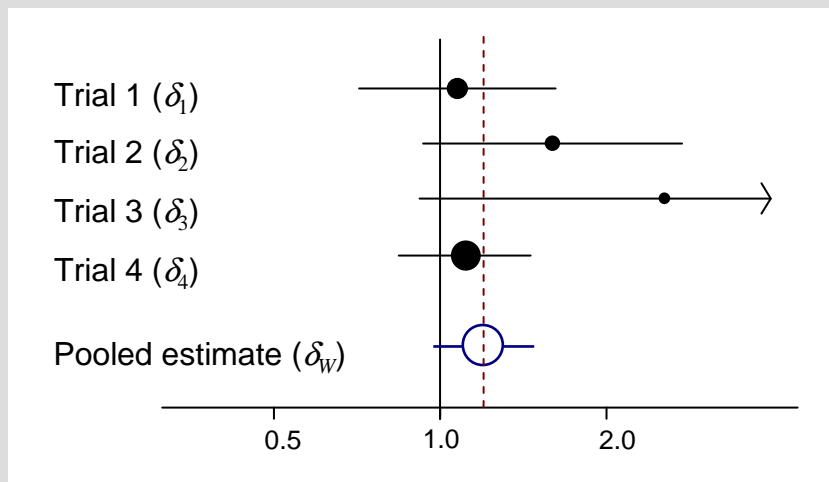
# Why use across-trial information?

- WT info tells us how treatment effect differs between patients with different values of a covariate:– true interaction
- AT info might help when combined with WT if WT has low power and AT has greater power
  - Analogous with case for non-randomised studies (cohort/case-control) – providing extra evidence where RCT evidence is low
  - If analysis is exploratory, may wish to maximise power if risk of bias is considered acceptable

# 1. Pooling of Within-Trial covariate Interactions (**WTI**)

$y$  = outcome;  $x$  = treatment;  $z$  = covariate (e.g. disease stage);  
 $\alpha$  = trial (“baseline”) coeff.;  $\beta$  = treatment coeff.;  $\gamma$  = covariate coeff.;  $\delta$  = interaction coeff.;  
 $i$  = trial;  $j$  = patient

- “Two-step” pooling of within-trial covariate interactions
  - Step 1: Fit regression model with interaction term separately within each trial  $i$ :  $g(y_{ij}) = \alpha_i + \beta_i x_{ij} + \gamma_i z_{ij} + \delta_i x_{ij} z_{ij}$
  - Step 2: Combine interaction coeffs  $\delta_i$  using inverse-variance MA



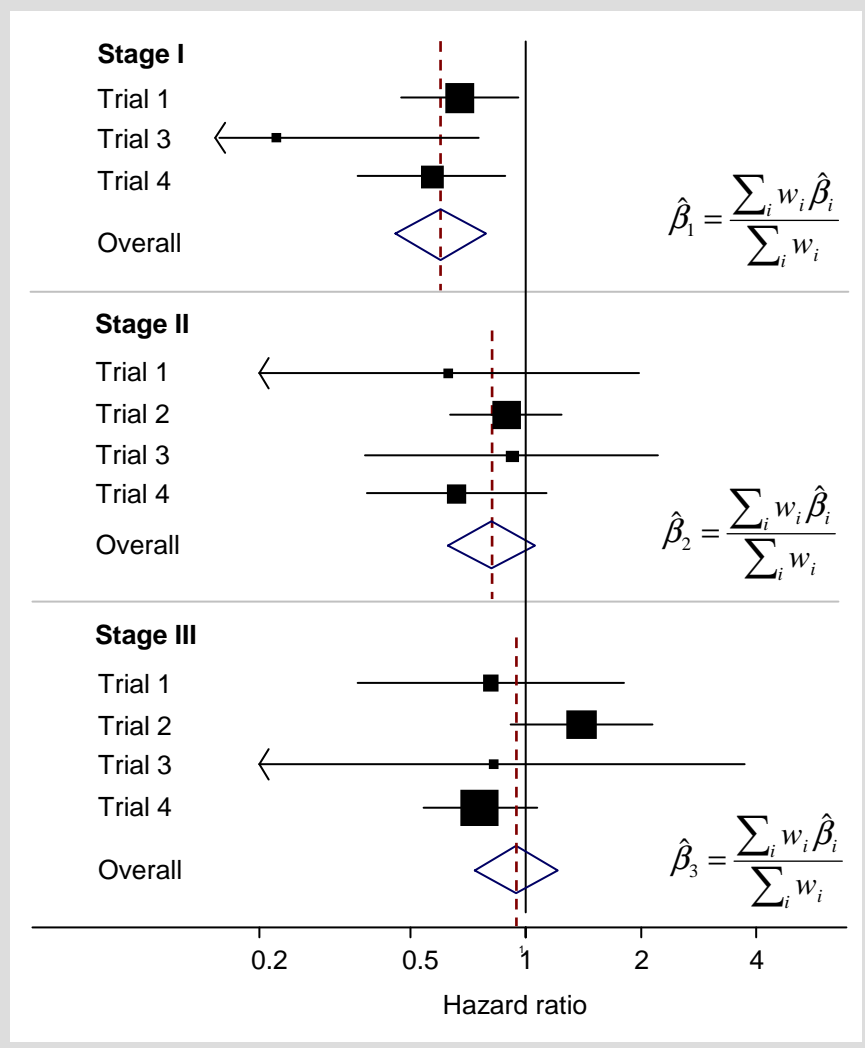
$$\hat{\delta}_w = \frac{\sum_i w_i \hat{\delta}_i}{\sum_i w_i} \quad se(\hat{\delta}_w) = 1/\sqrt{\sum_i w_i}$$

$W$  = “within-trials”

# Critique of WTI

- Within-trials effect only, so randomisation is retained
- Solid, simple approach for continuous and ordered categorical covariates (e.g. disease stage)
- BUT:
  - Trials with data in only one subgroup cannot be included
  - Unordered covariates cannot be analysed

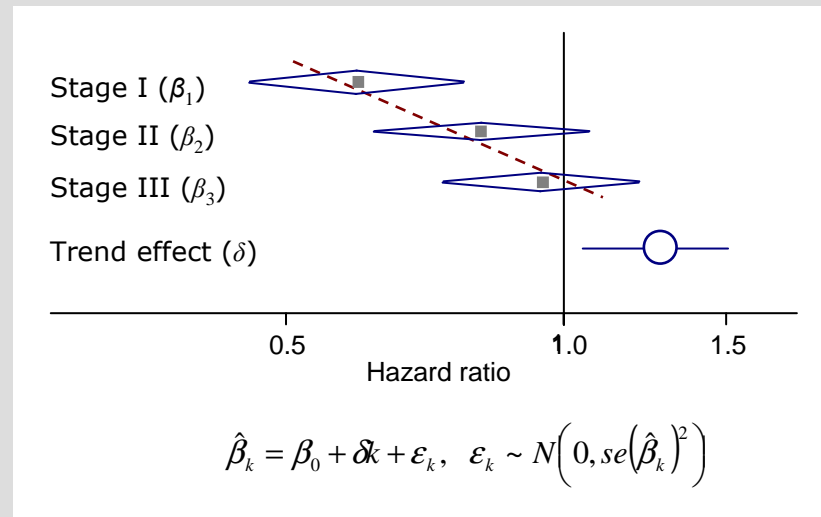
# 2. Across Covariate Subgroup effect (ACS)



Step 1: Split data into disease stage subgroups

Step 2: Carry out standard IPD MA for treatment effect  $\beta_k$  within each subgroup  $k$

Step 3: Fit a trend line to the treatment estimates  $\beta_k$  to obtain estimate of interaction  $\delta$ :

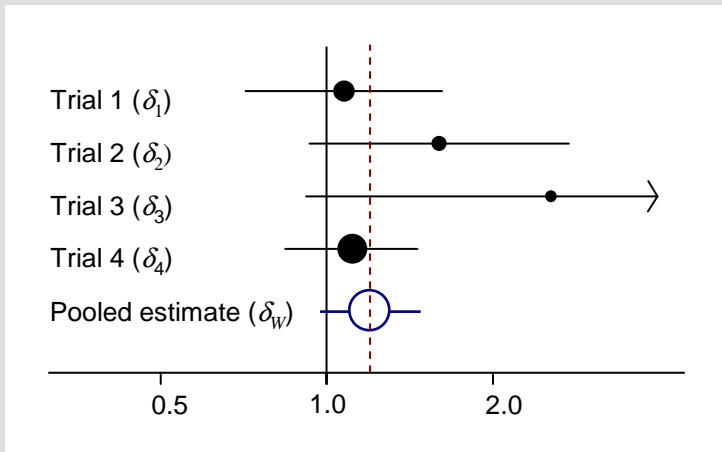


# Critique of ACS

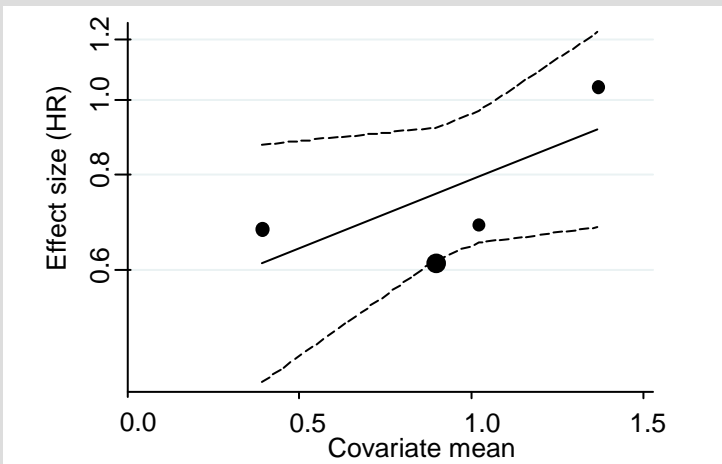
- Relatively simple
- Very commonly used (e.g. cancer)
- BUT:
  - Randomisation structure is broken
  - Within-trial correlations across covariate subgroups are unaccounted for
  - Mixture of within- and across-trials info, which cannot be separated: “black box”
    - ⇒ no way to assess ecological bias

# 3. Combining Within- & Across-trials effects (CWA)

Pooled within-trials effect (WTI)



Meta-regression of disease stage means

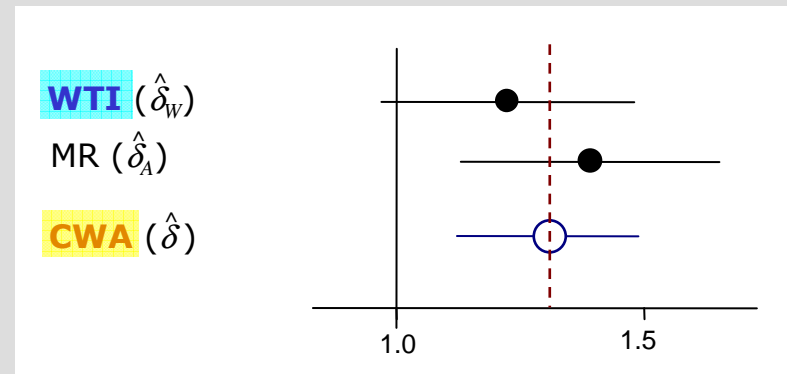


Step 1: Separately estimate within-trials effect  $\delta_w$  using WTI, and across-trials effect  $\delta_A$  using meta-regression (MR) on trial-level disease stage means

Step 2: Combine the two estimates in a further meta-analysis:

$$\hat{\delta} = \frac{w_W \hat{\delta}_W + w_A \hat{\delta}_A}{w_W + w_A}, \quad se(\hat{\delta}) = 1/\sqrt{w_W + w_A}$$

where  $w_W = 1/se(\hat{\delta}_W)^2$  and  $w_A = 1/se(\hat{\delta}_A)^2$



# Critique of CWA

- Relatively simple
- Can increase power over **WTI** in certain circumstances
- Allows WT & AT to be combined: pros & cons of this discussed later
- Transparency and user control
  - Each effect derived independently and combined manually by the analyst
- More open to bias and/or bad judgment?
  - Analyst could downweight unfavourable results

## 4. Mixed-Effects Models (MEM)

- “One-step” mixed-effects model
  - Uses all available data in a single interaction model:

$$g(y_{ij}) = \alpha_i + \beta x_{ij} + \gamma z_{ij} + \delta x_{ij} z_{ij}$$

- Interaction term  $\delta x_{ij} z_{ij}$  can also be parameterised so as to separate within- and across-trial effects:

$$g(y_{ij}) = \alpha_i + \beta x_{ij} + \gamma z_{ij} + \delta_W x_{ij} (z_{ij} - \bar{z}_i) + \delta_A x_{ij} \bar{z}_i$$

where  $\bar{z}_i$  is the covariate mean value in trial  $i$ ;

$\delta_W$  = within-trial effect;  $\delta_A$  = across-trial effect.

Difference  $\delta_A - \delta_W$  is a quantification of ecological bias.

Reference: Simmonds (2005)

# Critique of MEM

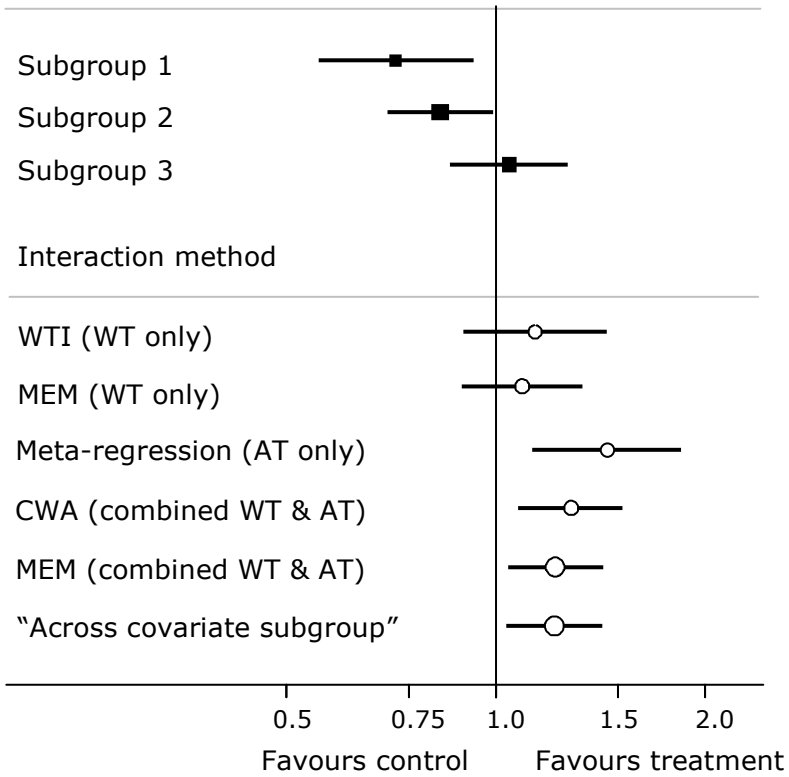
- Arguably 'gold standard' since other models are special cases
- Well-established in other study areas e.g. multi-centre trials
- Flexible & maximises power
  - Allows simultaneous estimation of multiple parameters, allowing non-linear interactions etc.
  - Can be parameterised to isolate WT & AT
- BUT:
  - Currently issues with random-effects with time-to-event data
  - Complex and computationally intensive; greater expertise & care required
- Allows WT & AT to be combined: pros & cons of this discussed later

# Using across-trials information

- Use of AT is controversial
  - Many analysts will never want to use it
  - Including it in the wrong circumstances risks ecological bias
  - Might be used if analysis is hypothesis generating, & more power is desired at the risk of increased bias
- In the event that AT info is under consideration, it is important to decide a priori:
  - Whether to consider it, with reasons
  - How to include it, with specific criteria
- Examples of criteria:
  - Estimates could differ by less than a pre-specified amount
  - Variance of the across-trial estimate could be increased by a factor before combining (CWA)

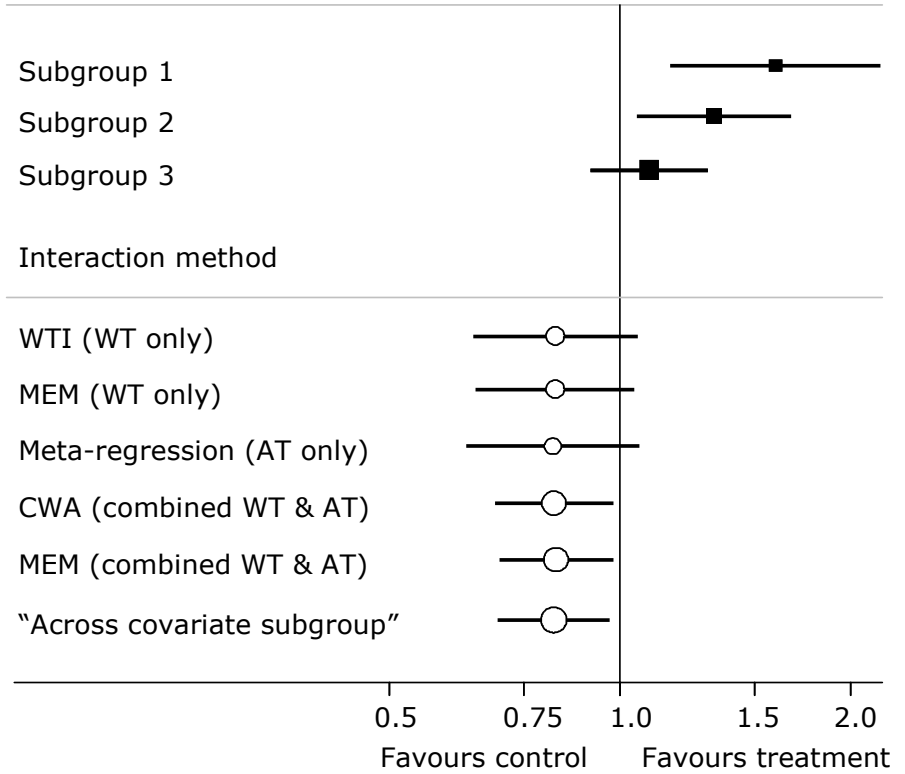
# Examples

### Treatment effect by covariate subgroup (1)



Using AT will introduce bias

### Treatment effect by covariate subgroup (2)



AT might be included to increase power

## Proposed strategy:

1. Make an *a priori* decision whether or not to consider using across-trials info, together with criteria for inclusion
  - Most analysts will decide not to
2. **WTI** is recommended in many situations, especially if complex stats are unnecessary
3. If across-trials info is to be considered, fit a meta-regression and test results against the pre-specified criteria; if criteria are met, use **CWA**
4. For those with suitable expertise and resources, **MEM** is a powerful and versatile alternative, and can include or exclude across-trials info as desired
5. **ACS** is not recommended for patient-level covariates and should be avoided.

Děkuji!

Thankyou!