

HIERARCHICAL MODELS

A framework for evidence synthesis

Innovative Methodology for Small Populations Research
(InSPiRe), WP4 “Evidence synthesis”

Tim Friede

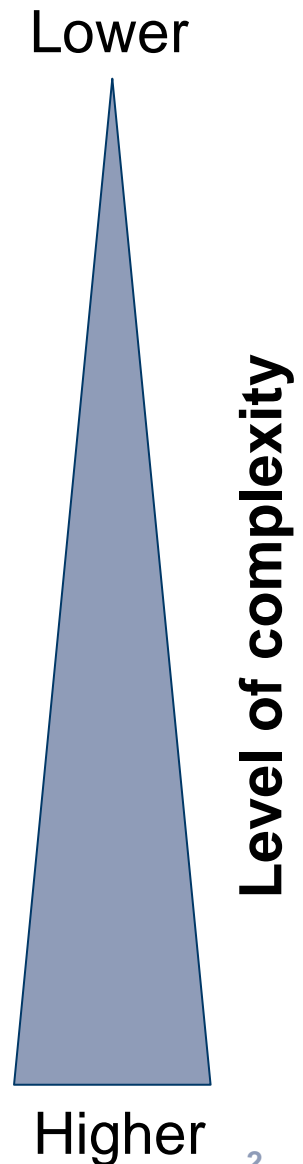
Dept. Medical Statistics

University Medical Center Göttingen

Göttingen, Germany

EVIDENCE SYNTHESIS

- ▷ **Pairwise meta-analysis**
 - ▷ comparing two treatments
- ▷ **Meta-regression**
 - ▷ including study-level covariates
- ▷ **Network meta-analysis**
 - ▷ comparing multiple treatments indirectly
- ▷ **RCT with historical controls**
 - ▷ integrating control group data from previous trials
- ▷ **Generalized (or cross design) synthesis**
 - ▷ combining data from different types of studies

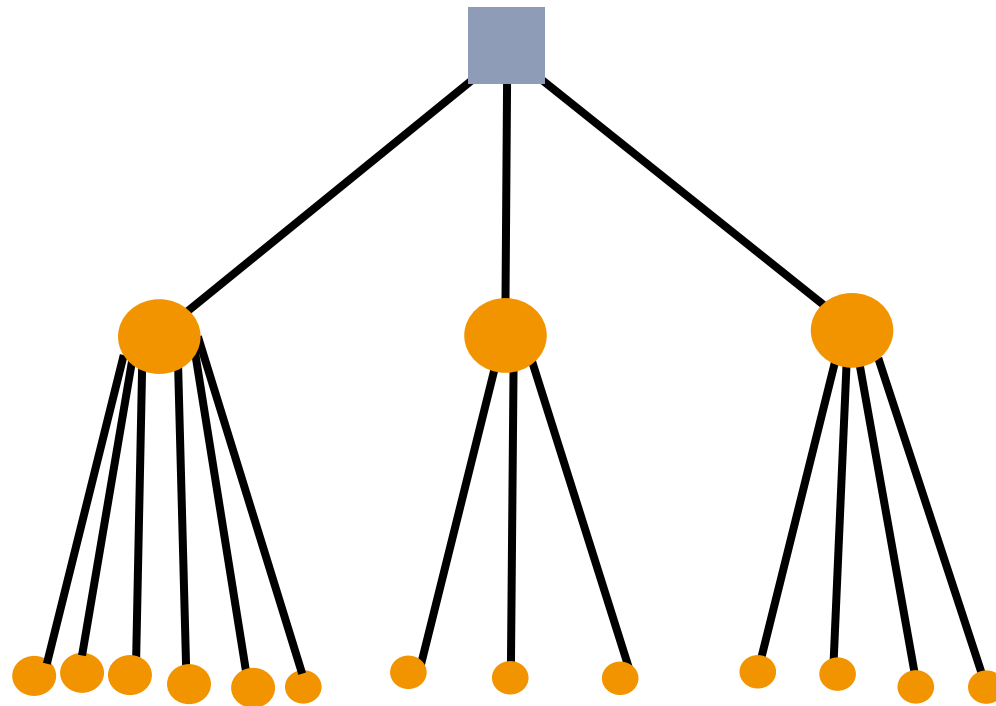


HIERARCHICAL MODELS

Meta-analysis

Studies

Patients



Example: **Normal-normal hierarchical model (NNHM)** for random-effects meta-analysis

$$y_i | \theta_i \sim \text{Normal}(\theta_i, s_i^2) \quad \theta_i | \Theta, \tau \sim \text{Normal}(\Theta, \tau^2)$$

Empirical studies scraping large databases of meta-analyses (e.g. Cochrane Library) show

- ▶ Meta-analyses of (very) **few studies** common
- ▶ **Extent of between-trial heterogeneity**

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METHODOLOGY

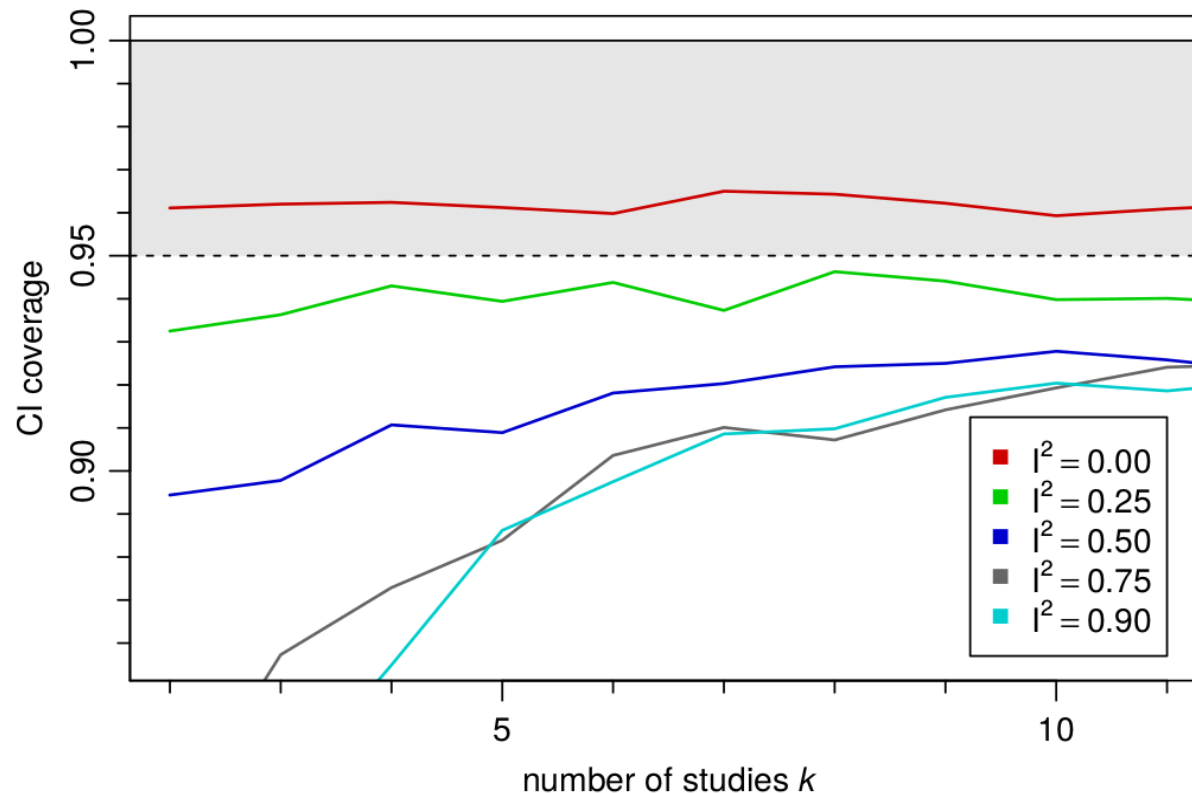
Predicting the extent of heterogeneity in meta-analysis, using empirical data from the *Cochrane Database of Systematic Reviews*

Rebecca M Turner,^{1*} Jonathan Davey,¹ Mike J Clarke,² Simon G Thompson³ and Julian PT Higgins¹

¹MRC Biostatistics Unit, Institute of Public Health, Cambridge, UK, ²All-Ireland Hub for Trials Methodology Research, Centre for Public Health, Queen's University Belfast, Northern Ireland and ³Department of Public Health and Primary Care, University of Cambridge, Cambridge, UK

*Corresponding author. MRC Biostatistics Unit, Institute of Public Health, Robinson Way, Cambridge CB2 0SR, UK.
E-mail: rebecca.turner@mrc-bsu.cam.ac.uk

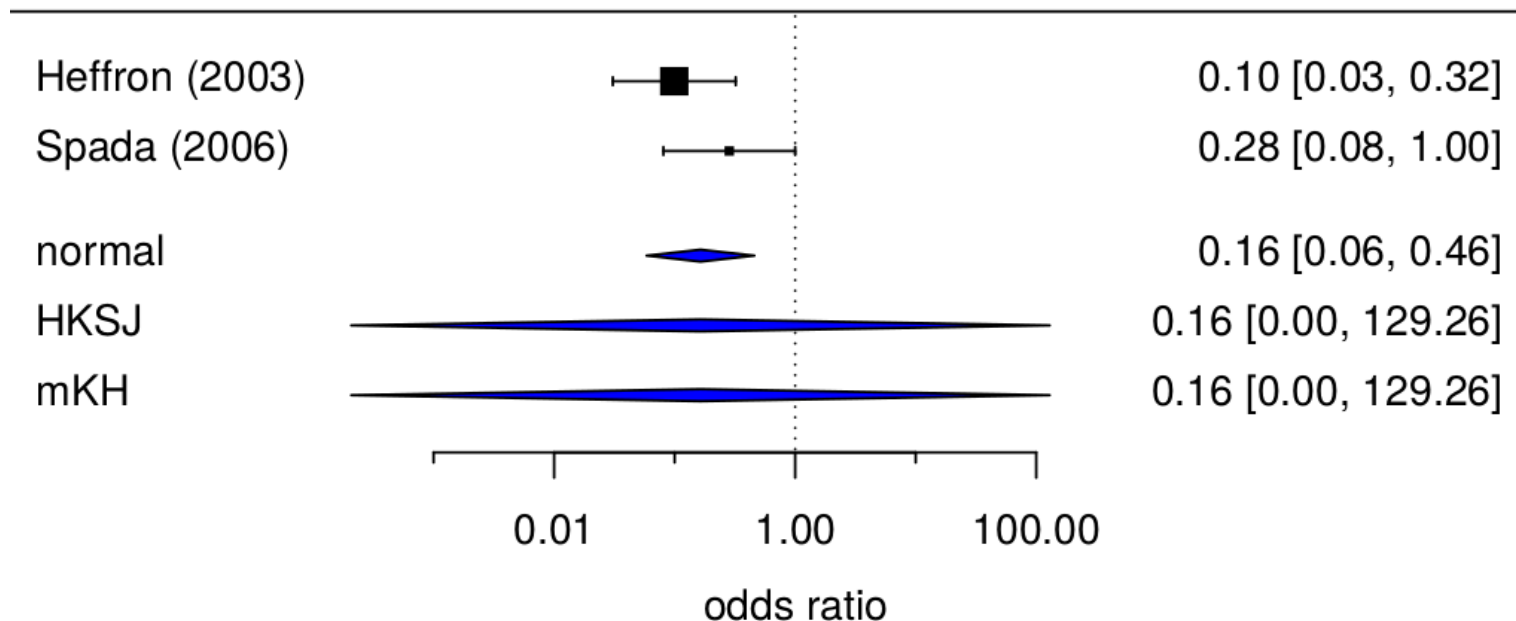
- ▷ Standard method (DerSimonian-Laird, DL)
 - ▷ Underestimates between-study heterogeneity
 - ▷ Fails to account for uncertainty in estimation of heterogeneity



WITH VERY FEW STUDIES: KNAPP-HARTUNG METHOD DOES NOT SOLVE THE PROBLEM

- ▶ 97.5% quantile of t-distribution with 1 df = 12.7 !!!
- ▶ Example from Friede et al (2017b)

Crins et al. (2014) example: acute graft rejection



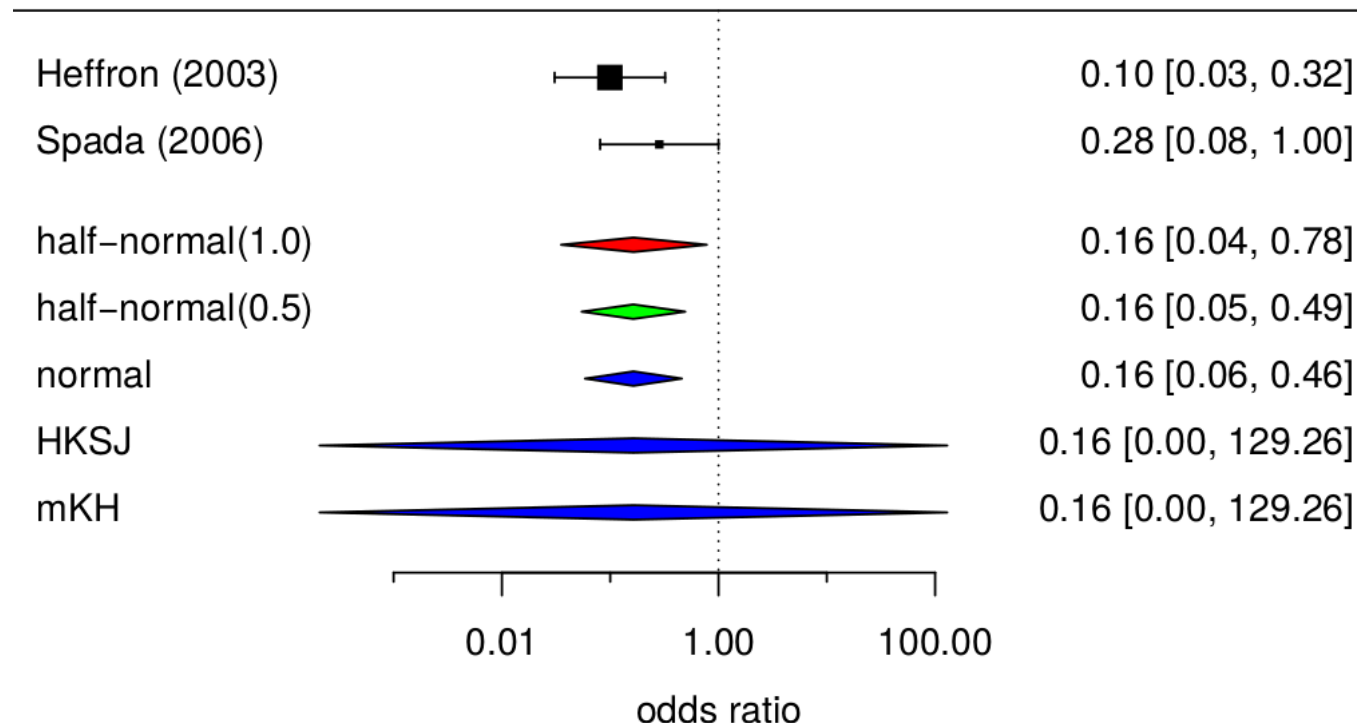
BAYESIAN META-ANALYSIS

- ▶ **Idea:** Weakly informative prior on between-trial heterogeneity for meta-analysis with few studies (Spiegelhalter et al, 2004), with uninformative prior on treatment effect
 - ▶ Avoids zero estimates of between-trial heterogeneity
 - ▶ Accounts for uncertainty in the estimation
- ▶ **Easy to compute**
 - ▶ Application of DIRECT algorithm (Röver & Friede, 2017a) (which is faster than MCMC sampling and does not require inspection of convergence diagnostics)
 - ▶ R package bayesmeta (available from CRAN)

EXAMPLE REVISITED

- **Bayesian intervals** appear to be a reasonable compromise (supported by simulation studies in e.g. Friede et al, 2017a,b)

Crins et al. (2014) example: acute graft rejection



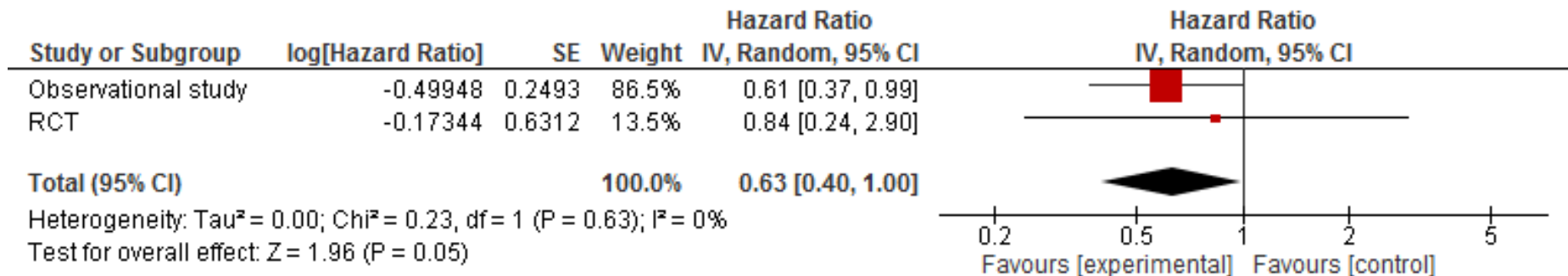
AN EXAMPLE OF CROSS-DESIGN SYNTHESIS

▶ Creutzfeldt-Jakob disease (CJD)

- ▶ prevalence of 1–9 cases per 1,000,000 people

▶ Varges et al (2017) investigated doxycycline in early CJD:

- ▶ double-blinded randomized phase II trial (n=12)
- ▶ observational study (n=88) (Cox regression stratified by terciles of the propensity scores)
- ▶ survival time as primary outcome

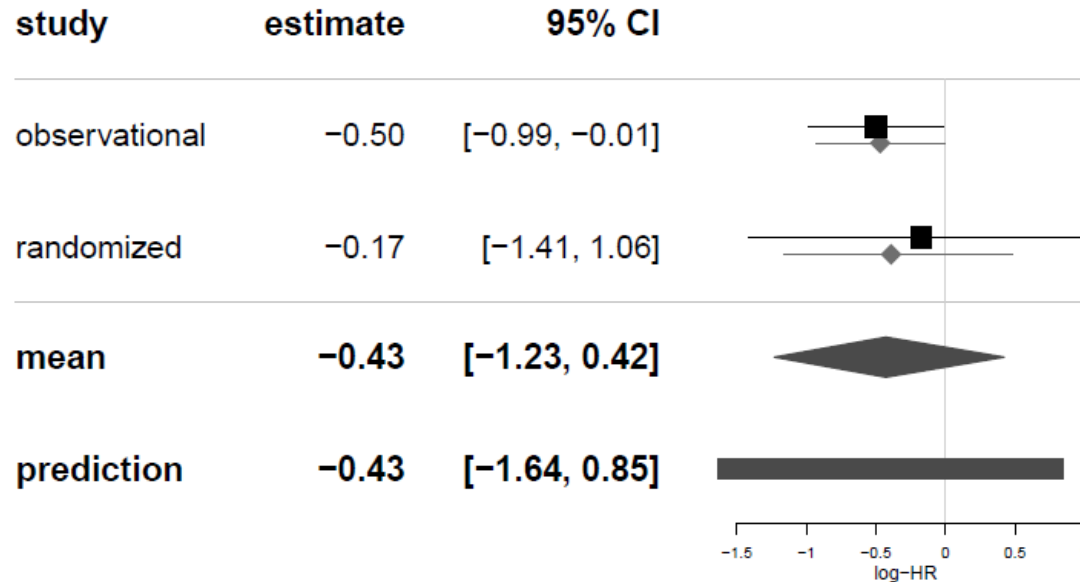


QUANTITIES OF INTEREST

Different **quantities of interest** in a random effects meta-analysis

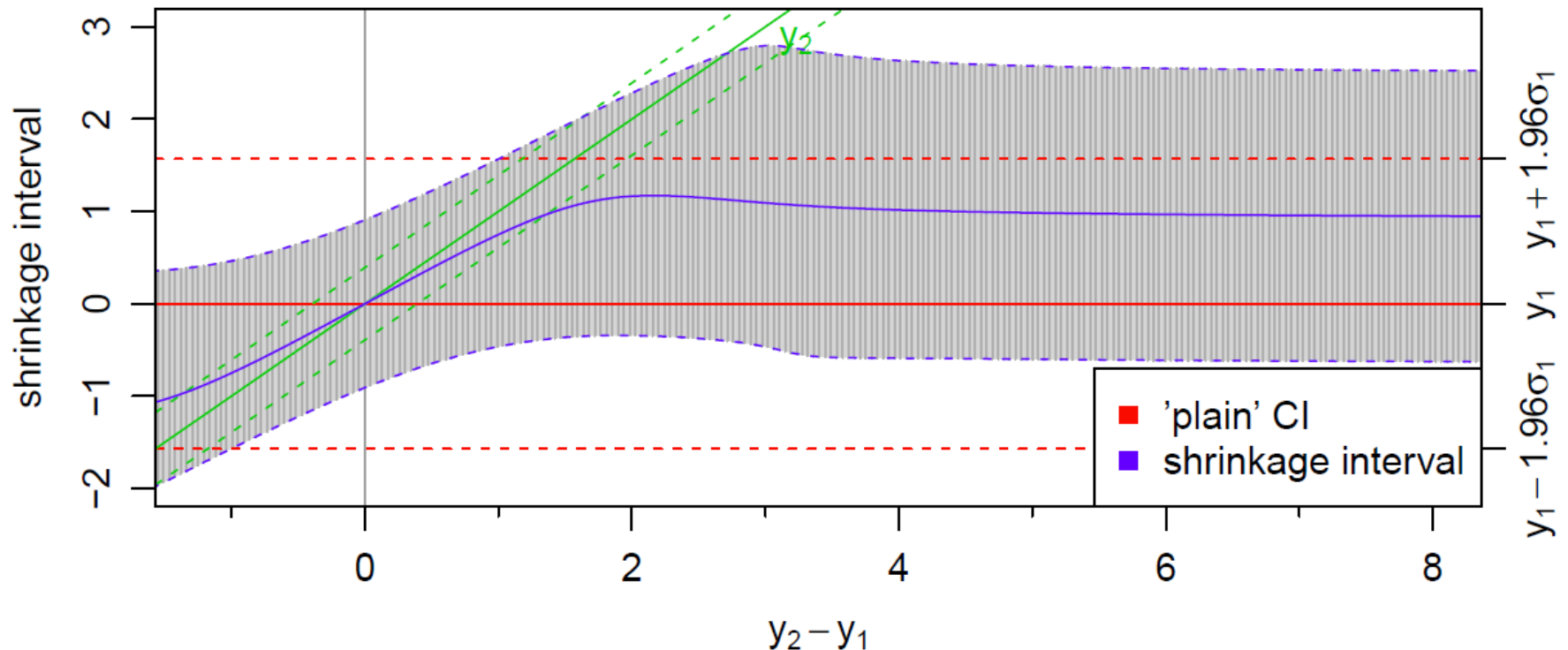
- ▶ average effect (θ) across studies
- ▶ effect (θ_{k+1}) of a future study (prediction / extrapolation)
- ▶ effect (θ_i) of an individual study in the light of the other studies (shrinkage estimator)

SHRINKAGE ESTIMATOR: EXAMPLE IN CJD



- ▶ RCT shrinkage interval width: 66% of original CI width
- ▶ Translates into 129% gain in sample size (about 27 instead of 12 patients)

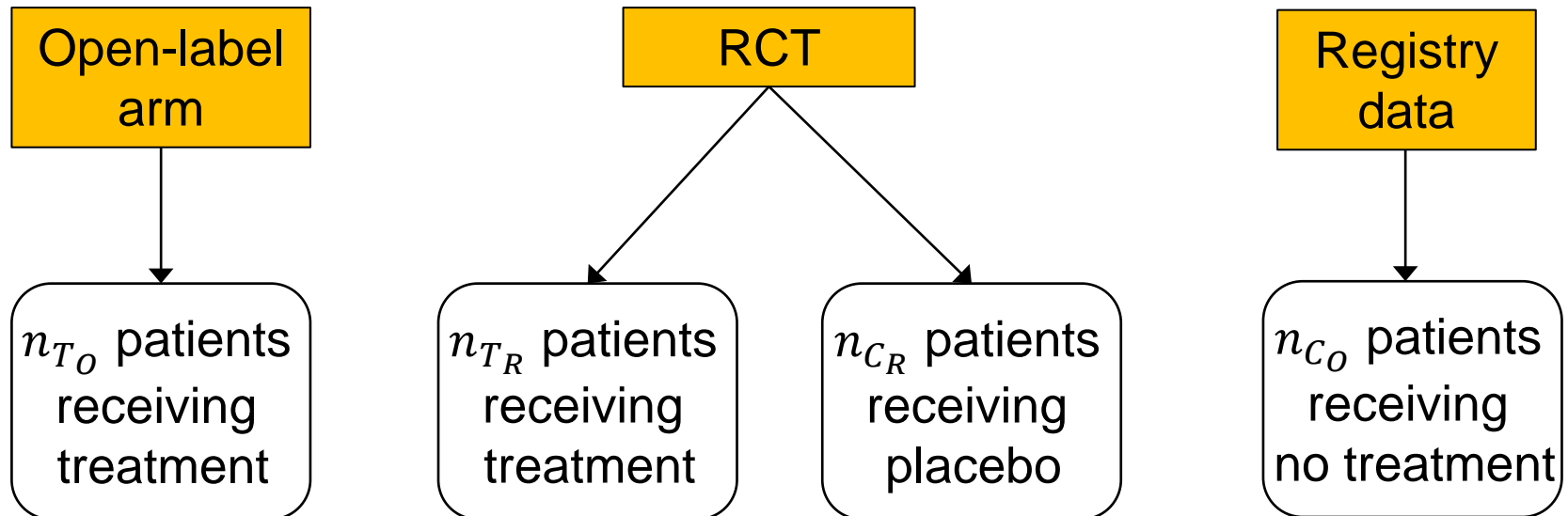
- **Idea:** Use of heavy-tailed meta-analytic predictive (MAP) prior (Schmidli et al, 2014)



- $n_1 = 25$, $n_2 = 400$, $p(\tau) = \text{HN}(0.5)$, interested in θ_1

MORE COMPLEX EXAMPLE: EARLY PRO-TECT TRIAL IN ALPORT DISEASE

- ▶ Alport syndrome is a rare genetic disorder that inevitably leads to end-stage kidney disease.
- ▶ Observational data suggest that the ACE inhibitor ramipril delays renal failure and improves life-expectancy in Alport patients.
- ▶ Our work (Unkel et al, 2017) is inspired by the ongoing EARLY PRO-TECT Alport trial in paediatric Alport patients (Gross et al. 2012).



CONCLUSIONS AND DISCUSSION

- ▶ **Hierarchical models**
 - ▶ flexible statistical framework for evidence synthesis
- ▶ **Bayesian inference:** advantages over traditional methods in the presence of heterogeneity and only (very) few studies
 - ▶ easy to apply using R package `bayesmeta`
- ▶ **Cross-design synthesis of available evidence**
 - ▶ Promising in rare diseases
 - ▶ more practical (and regulatory) experience needed

THE TEAM

The InSPiRe WP4 “Evidence synthesis” team includes

- ▶ **University Medical Center Göttingen (UMG):** Tim Friede, Steffen Unkel, Christian Röver, Burak Günhan, Katharina Kramer
- ▶ **Medical University Vienna (MUW):** Martin Posch
- ▶ **INSERM (Paris):** Sarah Zohar
- ▶ **University of Warwick:** Nigel Stallard
- ▶ **BfArM:** Norbert Benda
- ▶ **Novartis:** Beat Neuenschwander, Simon Wandel

- ▶ Friede T, Röver C, Wandel S, Neuenschwander B (2017a) Meta-analysis of few small studies in orphan diseases. *Research Synthesis Methods* 8: 79–91.
- ▶ Friede T, Röver C, Wandel S, Neuenschwander B (2017b) Meta-analysis of two studies in the presence of heterogeneity with applications in rare diseases. *Biometrical J* (in press).
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