

Nuisance-parameter based sample size reestimation in adaptive enrichment designs with an application in major depression

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BMBF project (BundesMinisterium für Bildung und Forschung) „**BIOSTATISTISCHE METHODEN ZUR EFFIZIENTEN EVALUATION VON INDIVIDUALISIERTEN THERAPIEN (BIMIT)**“.

Motivating examples in major depression

Combination of baseline variables to predict treatment response

- Frank et. al (2011) compared a pharmacotherapy and a psychotherapy in a randomized controlled study (n=318)
- 17-item Hamilton Depression Rating Scale (HDRS-17) over 12 weeks as efficacy endpoint
- A number of baseline variables which are predictive of treatment outcome were identified and combined to an *optimal moderator* of treatment effect (Kraemer HC, 2013)
- For patients above (below) a certain threshold of the optimal moderator psychotherapy was superior (inferior) to pharmacotherapy

Serum BDNF levels as predictor of treatment response

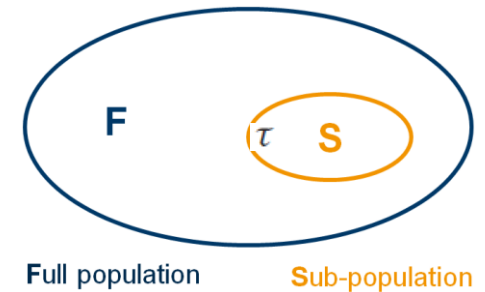
- A number of small uncontrolled studies identified baseline serum levels of brain derived neurotrophic factors (sBDNF) to predict treatment response to various pharmacological treatments (including duloxetine)
- HDRS-17 over 6 weeks (Mikoteit et al, 2014) or 8 weeks (Wolkowitz et. al, 2011)

In both examples the identified subgroups need to be confirmed in a RCT!

Adaptive Enrichment Design

Basic concept:

- ▷ **Stage 1:** Recruit patients from full population (F)
- ▷ **Interim analysis:** make the decisions on ...
 - ▷ whether trial is stopped for futility
 - ▷ if trial is continued, decide whether recruitment in **Stage 2** is from full population (F) or subpopulation (S) (enrichment)
 - ▷ e.g. epsilon-decision rule (Kelly et al 2005)
 - ▷ testing strategy in final analysis
- ▷ **Final analysis:** test for an effect in F and/ or S



Hypotheses and Test Statistics

- ▷ Normal distributed endpoints
- ▷ individual hypotheses $H_0^{\{F\}}$ (no effect in full population)
 $H_0^{\{S\}}$ (no effect in subpopulation)
- ▷ intersection hypothesis $H_0^{\{F,S\}}$ (no effect in full and subpopulation)
- ▷ standardized test statistics $Z^{\{F\}} = \sqrt{\frac{n}{2}} \frac{\bar{X}_F^T - \bar{X}_F^P}{\hat{\sigma}_F}$, $Z^{\{S\}} = \sqrt{\frac{n\hat{\tau}}{2}} \frac{\bar{X}_S^T - \bar{X}_S^P}{\hat{\sigma}_S}$
 - ▷ depend on estimates of nuisance parameters σ_F^2 , σ_S^2 , τ
- ▷ under $H_0^{\{F,S\}}$

$$\begin{pmatrix} Z^{\{F\}} \\ Z^{\{S\}} \end{pmatrix} \dot{\sim} MN \left(\begin{pmatrix} 0 \\ 0 \end{pmatrix}, \begin{pmatrix} 1 & \sqrt{\tau} \\ \sqrt{\tau} & 1 \end{pmatrix} \right)$$

Combination Test and Closure Principle

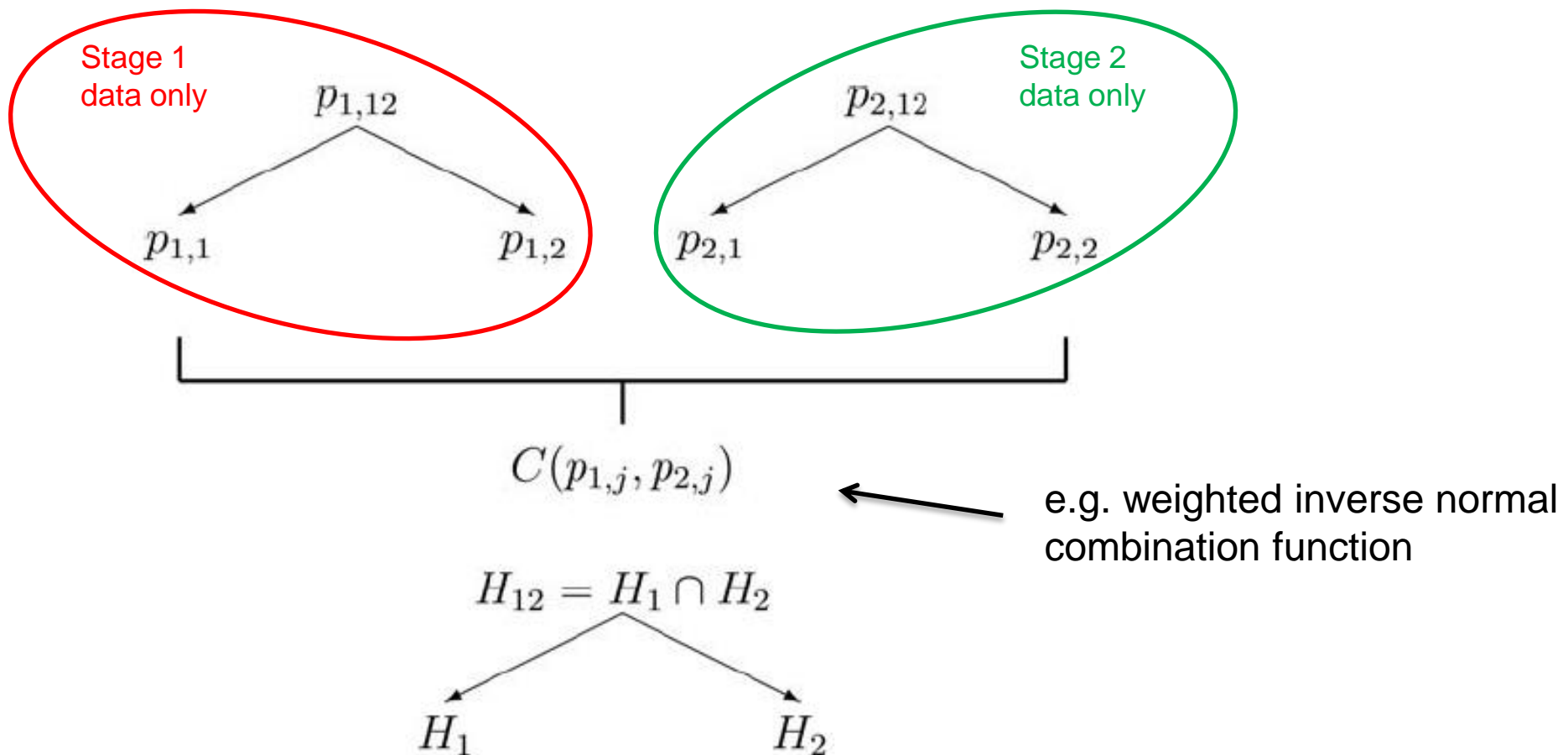
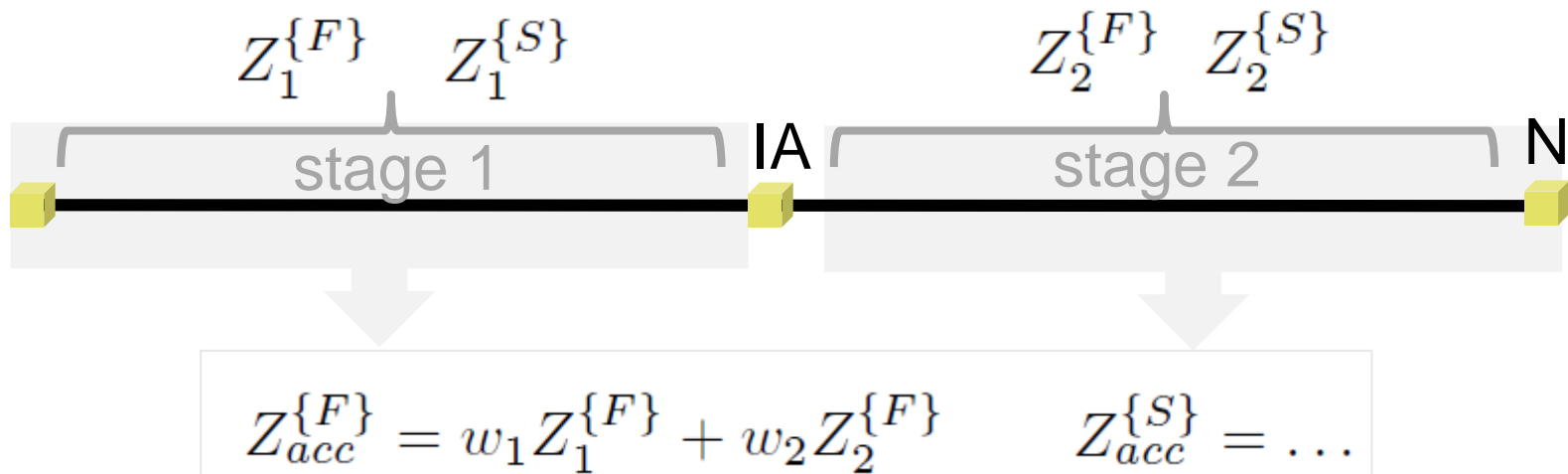


Figure 4 Closure principle for testing adaptively $n = 2$ null hypotheses H_1 and H_2 .

Figure taken from Bretz et al (2006) Biometrical Journal

Conditional Error Function Approach



- use **Stage 1** data to calculate the conditional error

$$CE = \mathbf{P} \left(\text{reject } H_0(Z_{acc}^{\{F\}}, Z_{acc}^{\{S\}}) | z_1^{\{F\}}, z_1^{\{S\}} \right)$$

- after **Stage 2** test with the accumulated data to the level of the conditional error
- For each individual hypothesis → apply closed testing procedure

Sample Size Calculation

- ▶ under the alternative
$$\begin{pmatrix} Z\{F\} \\ Z\{S\} \end{pmatrix} \overset{\sim}{\sim} \mathcal{N} \left(\underbrace{\begin{pmatrix} \sqrt{\frac{n}{2}} \frac{\Delta_F}{\sigma_F} \\ \sqrt{\frac{n\tau}{2}} \frac{\Delta_S}{\sigma_S} \end{pmatrix}}_{\boldsymbol{\delta}}, \underbrace{\begin{pmatrix} 1 & \sqrt{\tau} \\ \sqrt{\tau} & 1 \end{pmatrix}}_{\mathbf{V}} \right)$$
- ▶ let $G_{\mathcal{N}(\boldsymbol{\delta}, \mathbf{V})}$ denote the distribution function of $\mathcal{N}(\boldsymbol{\delta}, \mathbf{V})$ and $z_{\mathcal{N}(\mathbf{0}, \mathbf{V}), 1-\alpha}$ the $(1 - \alpha)$ - equicoordinate quantile of $\mathcal{N}(\mathbf{0}, \mathbf{V})$
- ▶ use estimates of nuisance parameters and effect sizes, e.g. based on previous studies, to calculate the initial sample size via

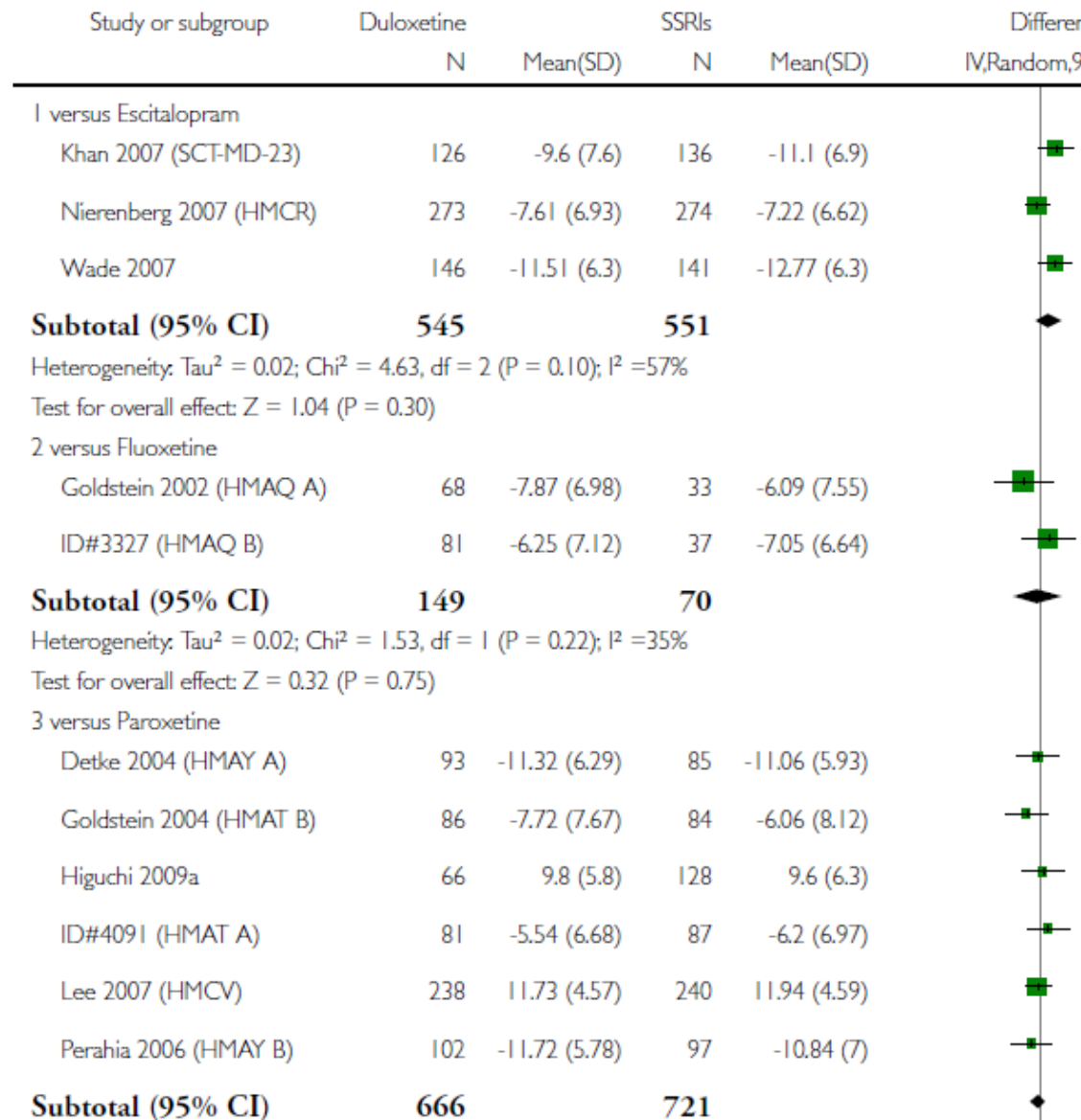
$$N_{init} = \min n, \text{ s.t. } 1 - G_{\mathcal{N}(\boldsymbol{\delta}, \mathbf{V})}(z_{\mathcal{N}(\mathbf{0}, \mathbf{V}), 1-\alpha}) \geq 1 - \beta$$

Problems?

- ▶ misspecifications of nuisance parameters
- ▶ example: variance of 17-HDRS outcome
- ▶ Cipriani et al. (2012)

Duloxetine versus other anti-depressive agents for depression (Review)

The Cochrane Library



Baseline HDRS score	22.3±5.1
(Mikoteit 2014)	22.0±4.1

	Escitalopram
(Wolkowitz 2011)	Baseline' (Week 8)
HDRS-17	26.1±8.3 13.2±8.9'

Problems?

- ▶ For example here: misspecification of σ_S^2

- ▶ Adaptive design: CEF approach

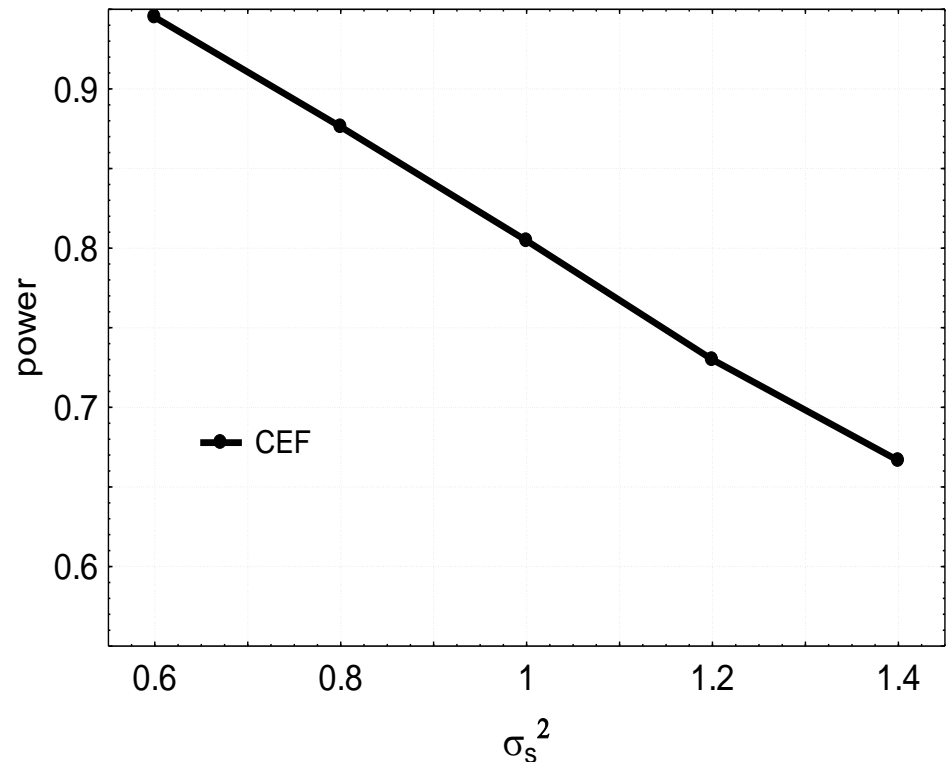
- ▶ nsim=10,000, N=128

- ▶ prevalence $\tau = 0.5$

- ▶ $\alpha = 0.025$, $1 - \beta = 0.8$

- ▶ $\Delta_{F \setminus S} = 0$, $\Delta_S = 0.5$

- ▶ $\sigma_F^{2*} = \sigma_F^2 = 1$, $\sigma_S^{2*} = 1$

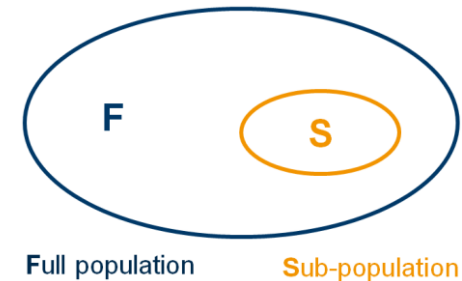


- ▶ motivation for sample size

recalculation procedure

- ▶ Internal Pilot Study Design (Wittes & Brittain, 1990)

IPS design with Blinded Review



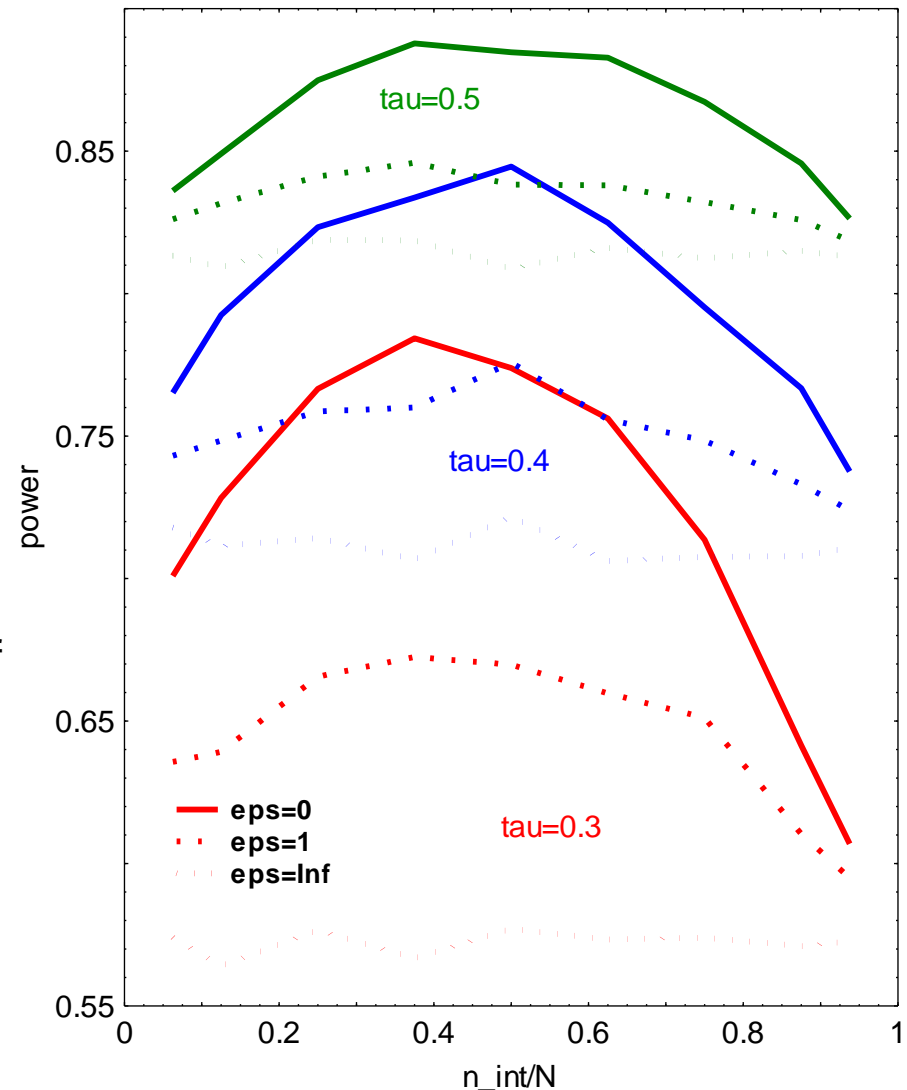
- ▶ here: nuisance parameters σ_F^2 , σ_S^2 and τ
- ▶ after $n_1 = p \cdot N_0$ subjects per group (treatment/control):
 - ▶ blinded reestimation via „lumped variance“

$$\hat{\sigma}_{F,OS}^2, \quad \hat{\sigma}_{S,OS}^2 = \frac{1}{2n_{1S}-1} \sum_{i \in \{T,C\}} \sum_{j=1}^{n_{1S}} (X_{ij} - \bar{X}_S)^2, \quad \hat{\tau} = n_{1S}/n_1$$

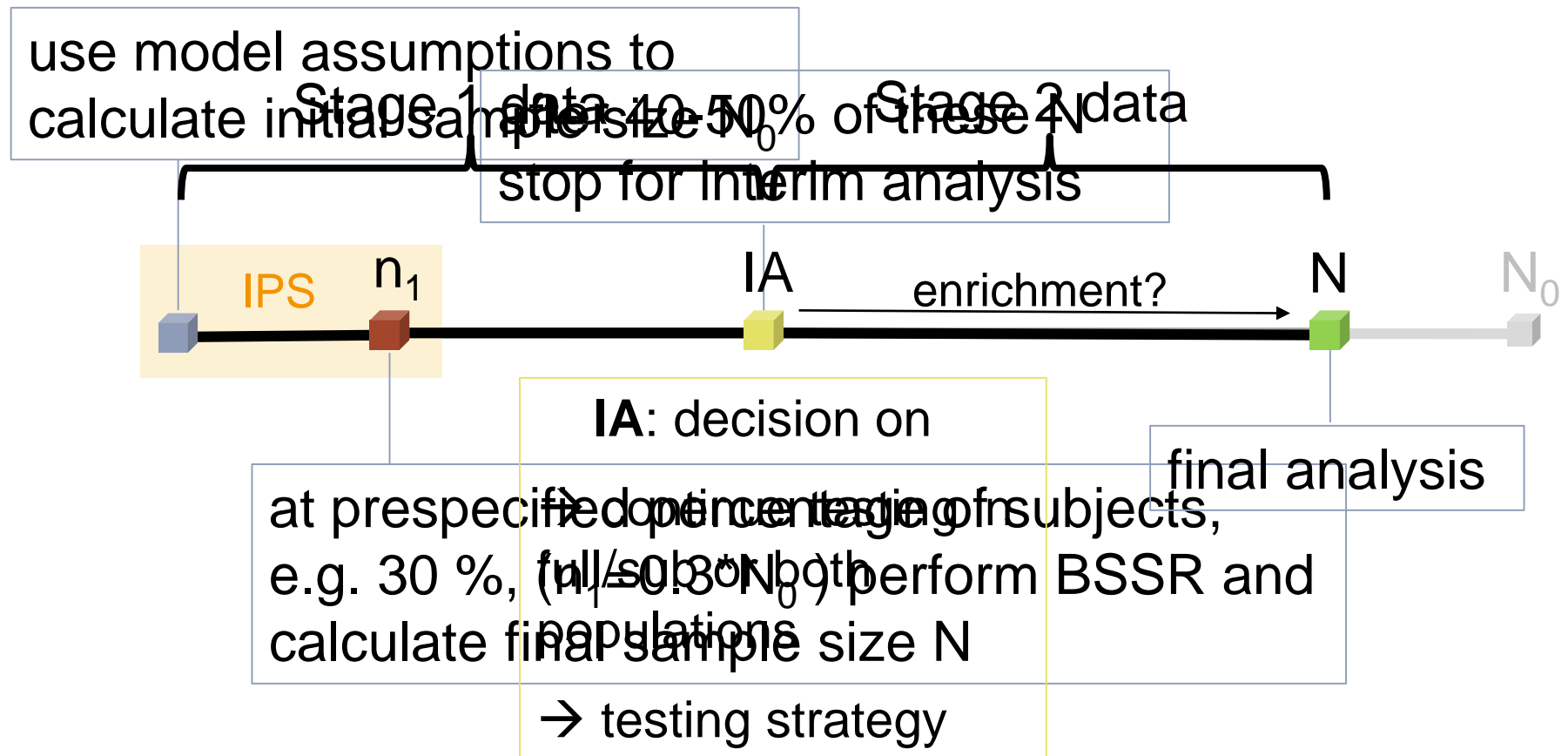
- ▶ here OS=OneSample means no unblinding of treatment/control group
- ▶ plug in new estimates and recalculate sample size $N = n_1 + n_2$ for final analysis

Optimal Timepoint for Interim Analysis?

- ▶ Adaptive design: CEF approach
- ▶ Simulation results for $n_{\text{sim}}=10,000$
- ▶ $N=400$ subjects per group (treatment/placebo)
- ▶ under the alternative
 $\Delta_{F \setminus S} = 0, \Delta_S = 0.3$
- ▶ maximum in power after 40-50% of the subjects

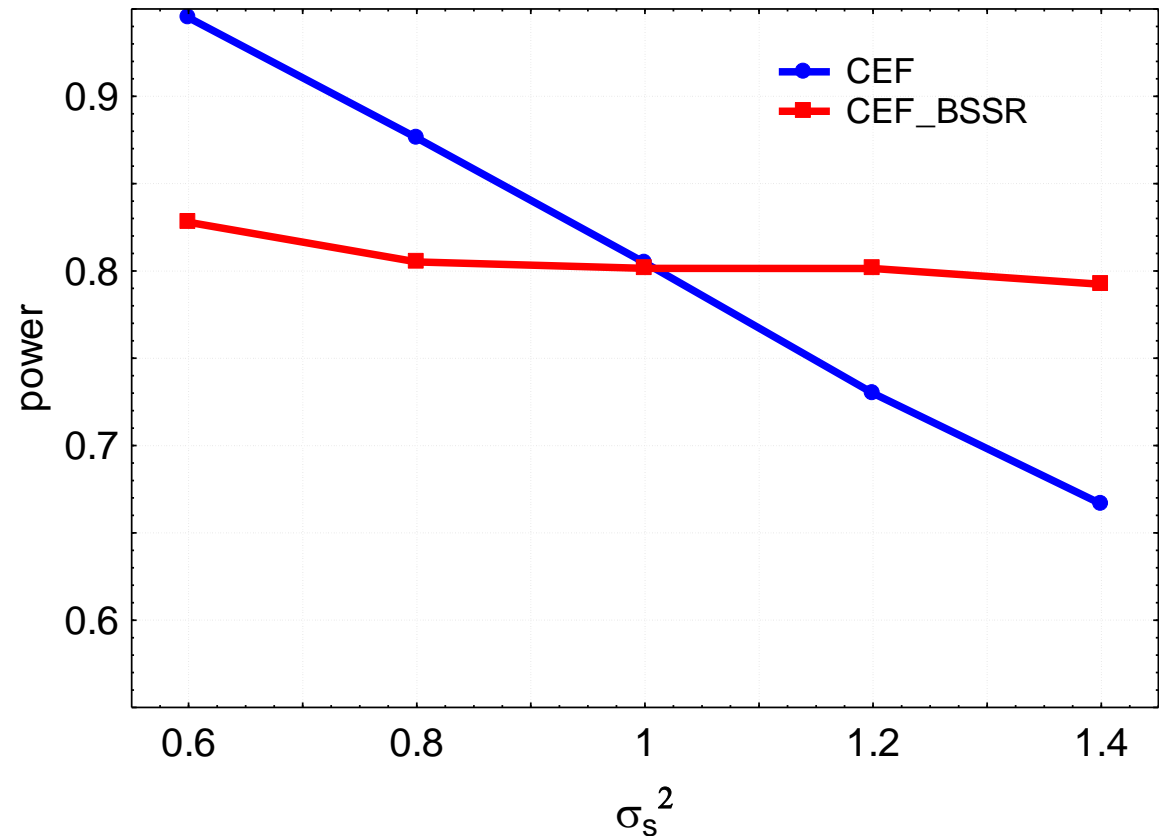


Combine BSSR and Adaptive Enrichment Methods



Combine BSSR and Adaptive Enrichment Methods

- ▷ $\text{nsim}=10,000$, $\tau = 0.5$
- ▷ $\alpha = 0.025$, $1 - \beta = 0.8$
- ▷ $\Delta_{F \setminus S} = 0$, $\Delta_S = 0.5$
- ▷ $\sigma_F^{2*} = \sigma_F^2 = 1$, $\sigma_S^{2*} = 1$
- ▷ BSSR at 30% of N_0
- ▷ Interim Analysis at 50% of N ($\varepsilon = 1$)



Combine BSSR and Adaptive Enrichment Methods

type-I-error rates

- ▷ $\text{nsim}=100,000, \tau^* = 0.4$
- ▷ $\sigma_F^2 = \sigma_S^2 = \sigma_{F^*}^2 = \sigma_{S^*}^2 = 1$
- ▷ $\alpha = 0.025$
- ▷ $\Delta_{F \setminus S} = \Delta_S = 0$
- ▷ BSSR at 30% of N_0
- ▷ Interim Analysis at 50%
of N ($\varepsilon = 1$)

tau	N_0	N	CTSD	CEF
0.3	662	843	0.02435	0.02509
0.4	662	664	0.02491	0.02581
0.5	662	537	0.02426	0.02489
0.3	371	475	0.02487	0.02528
0.4	371	375	0.02426	0.02528
0.5	371	303	0.02506	0.02516
0.3	237	305	0.02561	0.02594
0.4	237	241	0.02484	0.02558
0.5	237	194	0.02462	0.02500
0.3	169	213	0.02531	0.02564
0.4	169	168	0.02566	0.02578
0.5	169	136	0.02553	0.02583

Conclusions & Discussion

- ▶ Combination of BSSR and Adaptive Enrichment Methods leads to robust and flexible design
- ▶ increasing computational time due to computational complexity with increasing number of subgroups (simulations in planning stage)
- ▶ extension to nonnormal endpoints, e.g. count data
- ▶ include modeling of drop-outs

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