

Improving interim decisions for single-arm trials by adjusting for baseline covariates and short-term endpoints

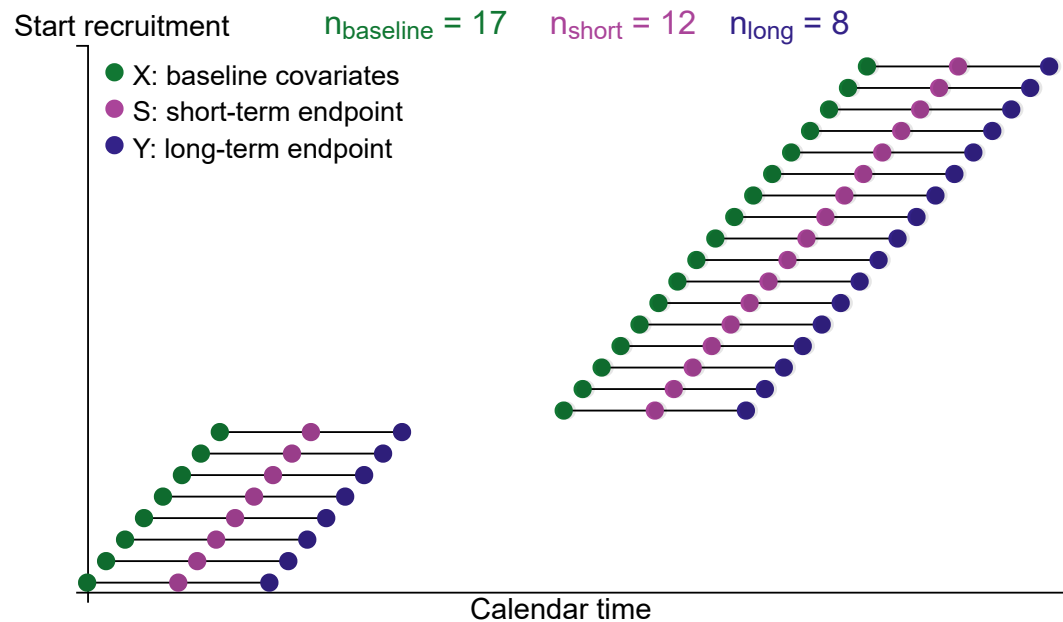
Eline Anslot

Joint work with Kelly Van Lancker

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Single-arm trials with multi-stage designs

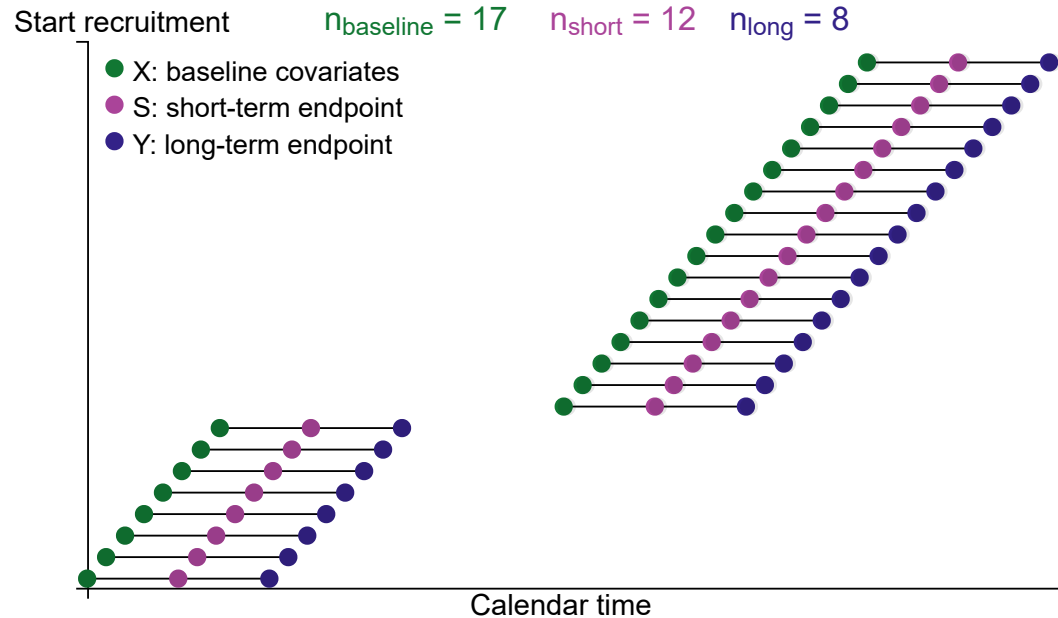


Continued/Paused

Commonly used designs:

- Group sequential designs
- Simon's two-stage designs

Interim analysis of two-stage designs



Continued/Paused

Interim analysis: based on the long-term endpoint \Rightarrow Unadjusted analysis

- **Simon's two-stage design:**

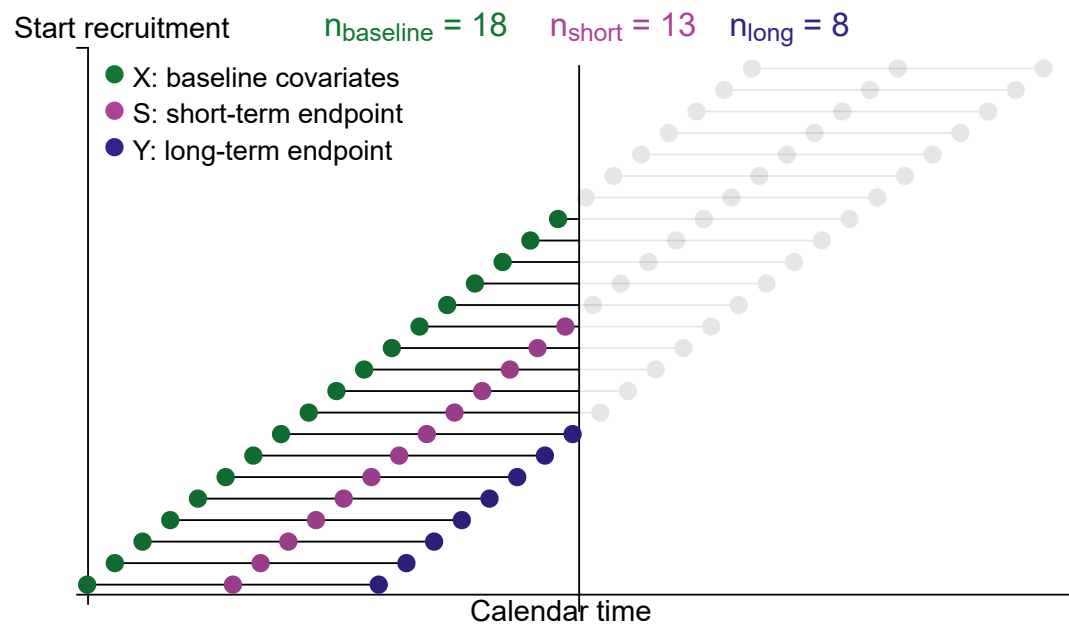
$$\sum_{i=1}^{n_{\text{long}}} y_i \leq r_1$$

- **Group sequential designs:**

$$Z = \frac{\frac{1}{n_{\text{long}}} \sum_{i=1}^{n_{\text{long}}} y_i - p_0}{\text{SE}\left(\frac{1}{n_{\text{long}}} \sum_{i=1}^{n_{\text{long}}} y_i\right)}$$

with z compared to cut-off to stop a trial for futility or efficacy based on e.g., Pocock (1977), O'Brien and Fleming (1979) or Lan and DeMets (1983) α - or β - error spending functions

Can we use more information?



Unadjusted analysis

Reset

Adjusted analysis

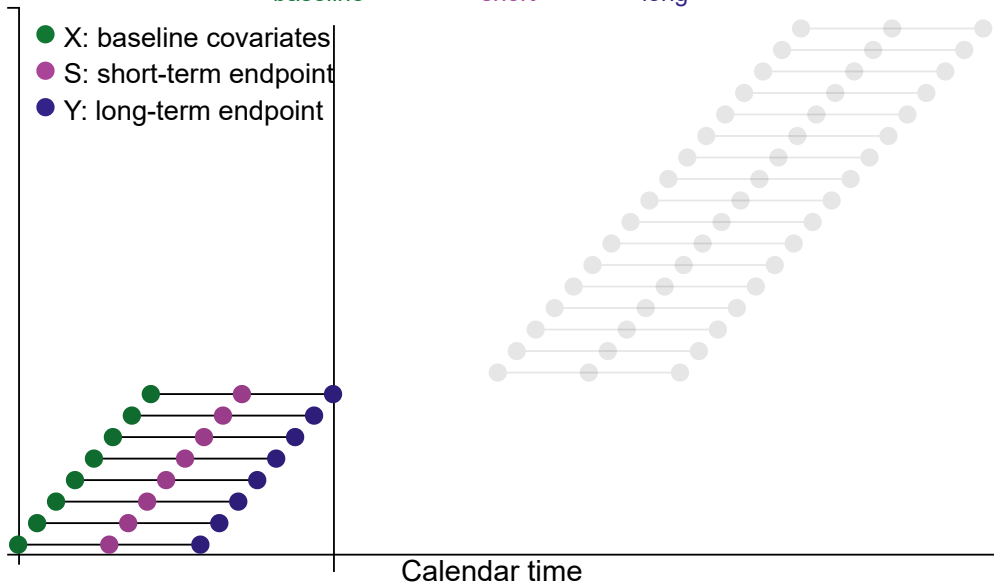
Improve decision by short-term endpoint

Start recruitment

$n_{\text{baseline}} = 8$

$n_{\text{short}} = 8$

$n_{\text{long}} = 8$



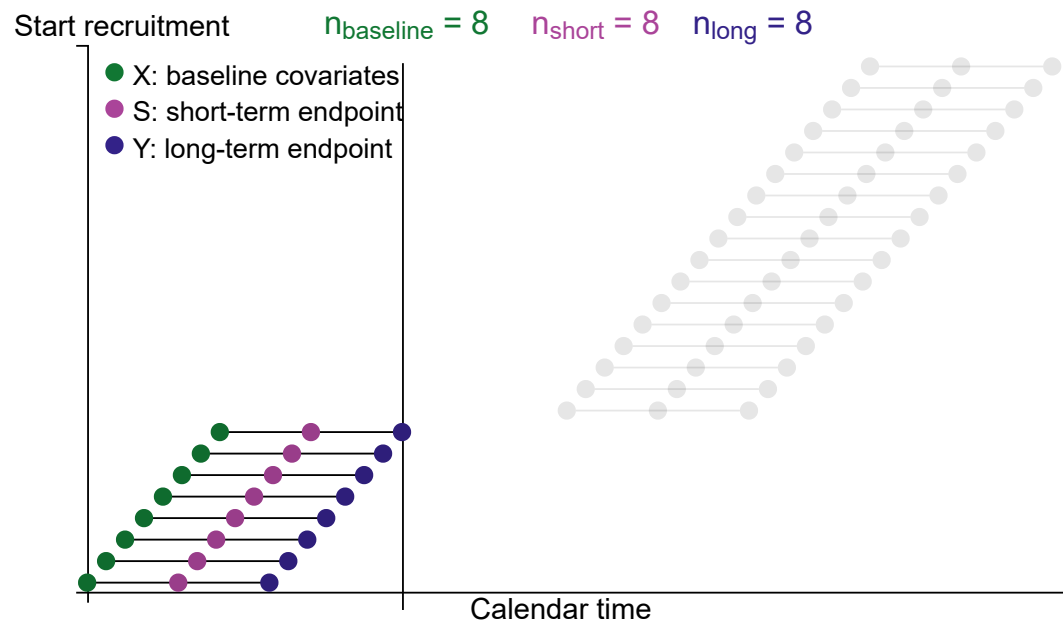
Unadjusted analysis

Kunz et al. (2017)

Zocholl et al. (2023)

Reset

More precise interim estimator



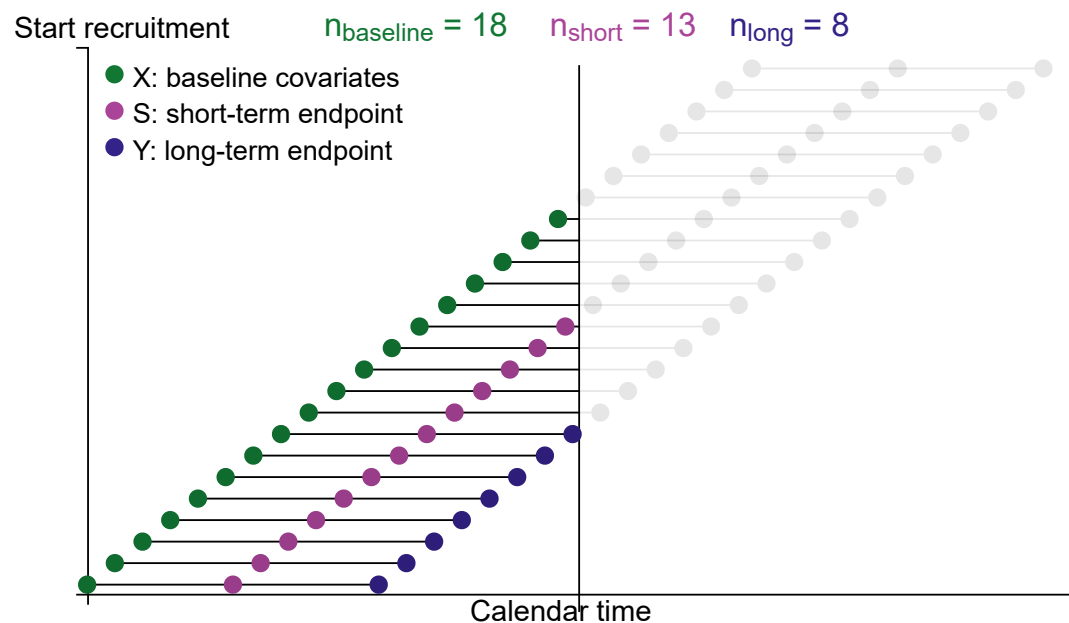
Continued/Paused

- **Recruitment with a pause:**
Possible as in Kunz et al. (2017) and Zocholl et al. (2023)
- **Continuous recruitment:**
⇒ Focus of the talk

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Proposed method



Step 1: Model fitting in cohort 1:

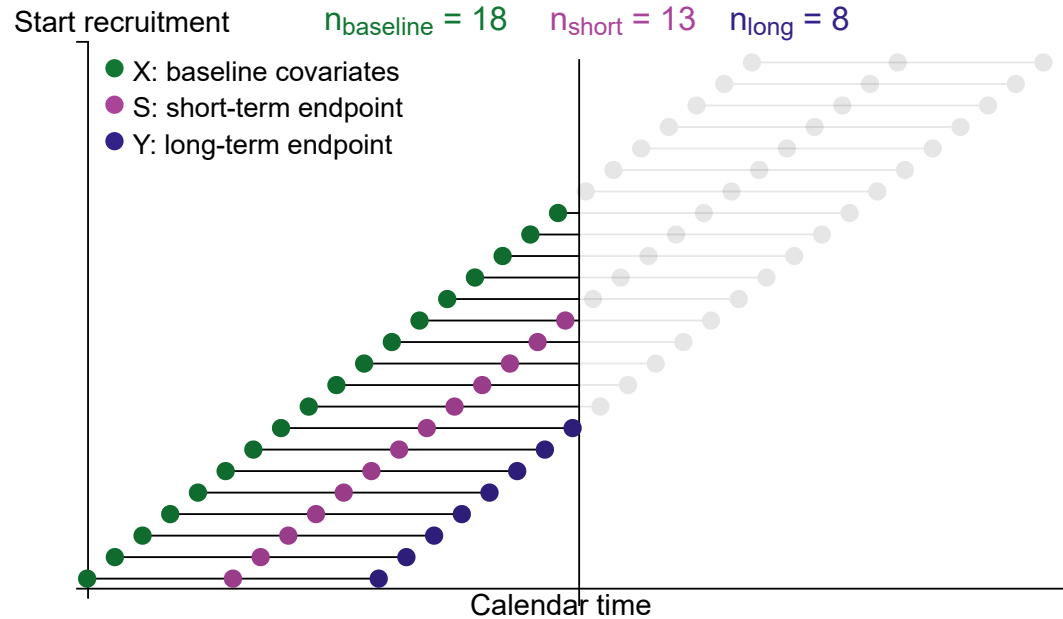
$$h[E(Y | X, S)] = \beta_0 + \beta_1 X + \beta_2 S$$

$h(\cdot)$: canonical link function

Step 2: Predicting in cohort 1 and 2:

$$\hat{Y} = h^{-1} [\hat{\beta}_0 + \hat{\beta}_1 X + \hat{\beta}_2 S]$$

Proposed method



Step 3: Model fitting in cohort 1 and 2:

$$h[E(\hat{Y} | X)] = \gamma_0 + \gamma_1 X$$

Step 4: Predicting in cohort 1, 2 and 3:

$$\hat{Y}' = h^{-1}[\hat{\gamma}_0 + \hat{\gamma}_1 X]$$

Step 5: Averaging

$$\hat{p}_{\text{int}} = \frac{1}{n_{\text{baseline}}} \sum_{i=1}^{n_{\text{baseline}}} \hat{Y}'_i$$

Proposed method - Decision at interim

Decision at interim:

$$Z_{\text{int}} = \frac{\hat{p}_{\text{int}} - p_0}{\text{SE}(\hat{p}_{\text{int}})}$$

In Group Sequential Design:

Z_{int} compared to cut-off to stop a trial for futility or efficacy based on e.g.,

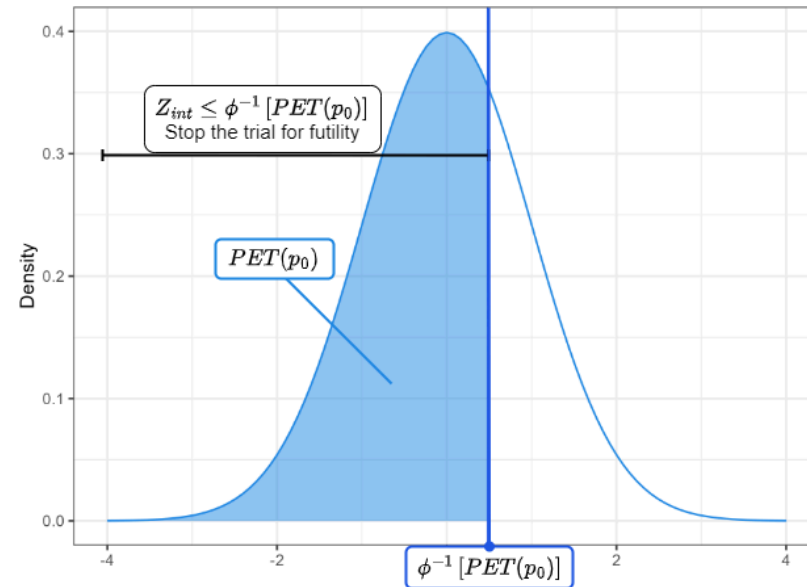
Pocock (1977), O'Brien and Fleming (1979) or

Lan and DeMets (1983) α - or β - error spending functions

In Simon's Two-Stage:

$$Z_{\text{int}} \leq \phi^{-1}[\text{PET}(p_0)]$$

with $\text{PET}(p_0) = B(r_1; n_{\text{long}}, p_0)$



Proposed method

Decision at interim:

- Adjusting for multiple short-term endpoints and baseline covariates
- Asymptotically unbiased even with misspecified models
 - Under random recruitment
- Asymptotically efficient when models are correct

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Simulation settings

$\alpha = 0.05$, power = 0.80

Design: two different optimal two-stage designs to generate e.g., n_{long} and r_1

Design	p_0	p_A	n_{final}	n_{long}	r_1	PET(p_0)
1	.25	.35	149	56	15	.6853
2	.25	.30	522	223	57	.6112

Setting 1:

Design	Adjustment		$n_{interim}$		Proportion n_{short}	Degree of Predictivity
	Baseline covariate(s)	Short-term endpoint	n_{long}	n_{short}		
1	/	1	56	15	0.20	Low to High
				25	0.30	
				58	0.50	
				86	0.60	
				58	0.20	
2	/	1	223	99	0.30	Low to High
				228	0.50	
				299	0.60	

Simulation settings

$\alpha = 0.05$, power = 0.80

Design: two different optimal two-stage designs to generate e.g., n_{long} and r_1

Design	p_0	p_A	n_{final}	n_{long}	r_1	PET(p_0)
1	.25	.35	149	56	15	.6853

Setting 2:

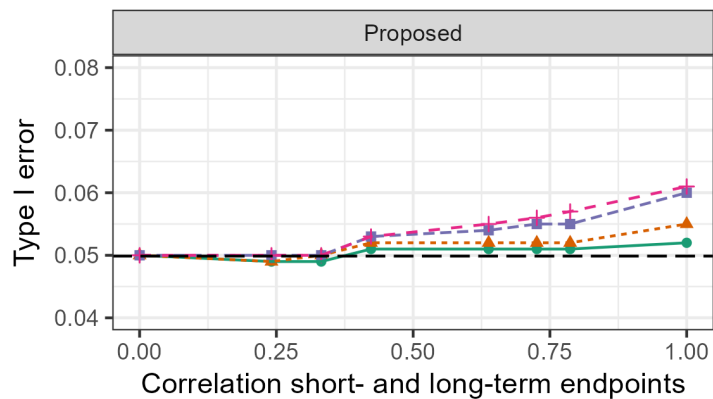
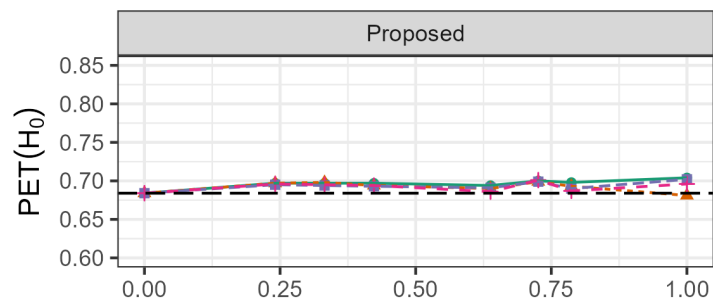
Design	Adjustment		n_{interim}			Degree of Predictivity	Models	
	Baseline covariate(s)	Short-term endpoint	n_{cohort1}	n_{cohort2}	n_{cohort3}		Correct	Misspecified
1	3	1	56	29	29	Low, Moderate, High	✓	✗

Setting 1 - Under the null hypothesis

Proportion n_{short} — 0.2 — 0.3 — 0.5 — 0.6

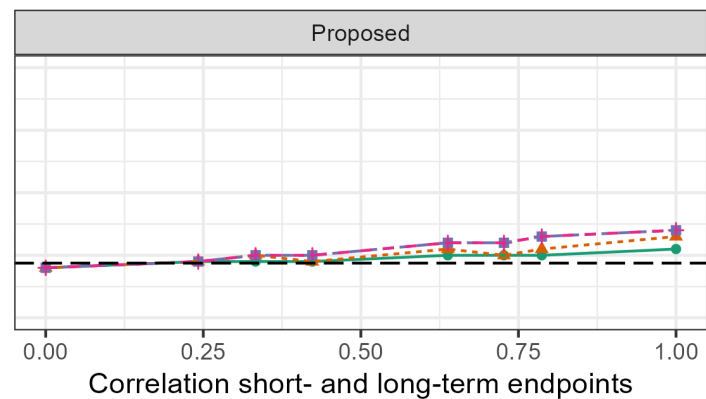
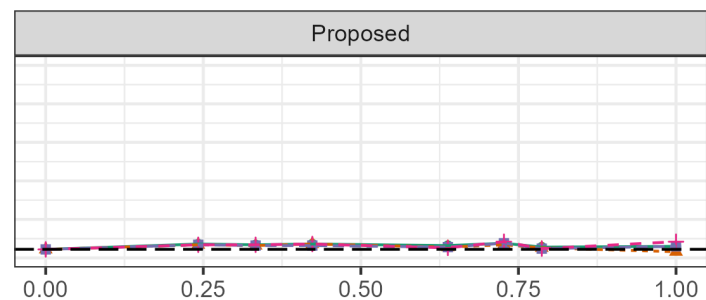
Design 1

$n_{\text{final}} = 149$, $n_{\text{long}} = 56$, $p_A = .35$ and $p_0 = .25$



Design 2

$n_{\text{final}} = 522$ and $n_{\text{long}} = 223$, $p_A = .30$ and $p_0 = .25$

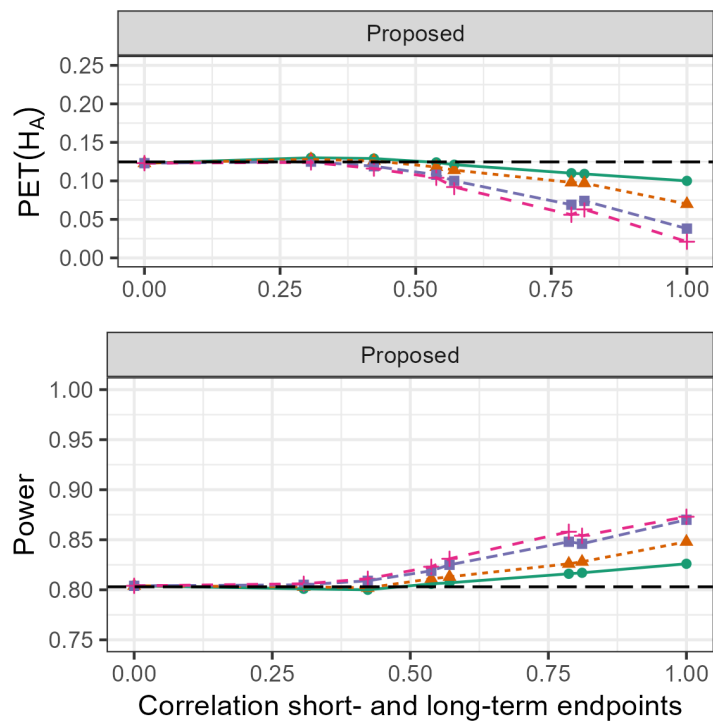


Setting 1 - Under the alternative hypothesis

Proportion n_{short} — 0.2 — 0.3 — 0.5 — 0.6

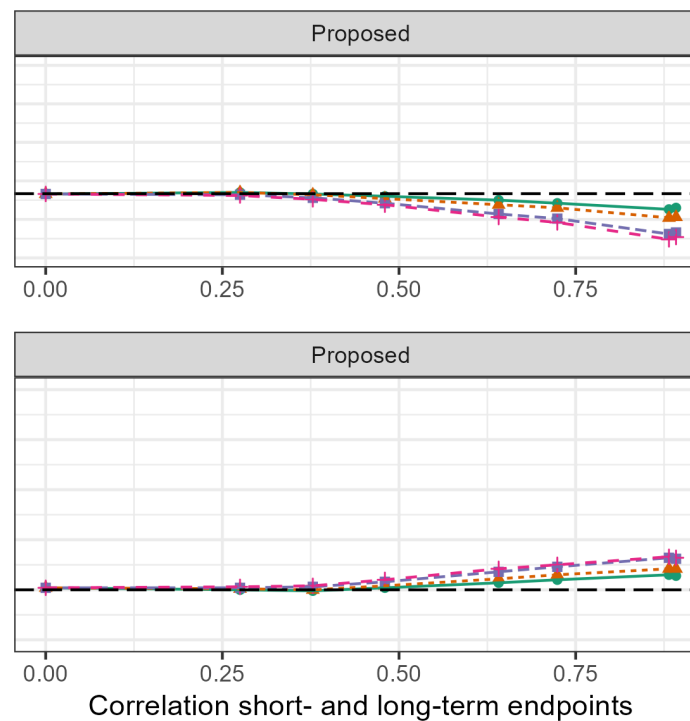
Design 1

$n_{\text{final}} = 149$, $n_{\text{long}} = 56$, $p_A = .35$ and $p_0 = .25$



Design 2

$n_{\text{final}} = 522$ and $n_{\text{long}} = 223$, $p_A = .30$ and $p_0 = .25$



Setting 2 - Model misspecification

Models	Degree of predictivity	power
Correct	Not predictive	79.7%
	Moderately predictive	82.4%
	Highly predictive	84.6%
Main	Not predictive	79.6%
	Moderately predictive	82.4%
	Highly predictive	84.7%
$ X_1 $	Not predictive	79.9%
	Moderately predictive	80.2%
	Highly predictive	80.8%

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Discussion

Additional gain of the proposed method, depends on:

- Proportion of additional participants in the pipeline
 - But ideally not everybody should be recruited at interim
- Predictivity of baseline covariates and short-term endpoint
- Model misspecification
 - Extension: data-adaptive methods to help build the models (see e.g., Van Lancker et al., 2024)

Discussion

- Calculate sample size as if no power gain occurred
- Type I error rate inflated in small sample
 - Estimator's variance leans on asymptotic theory
 - Decision at interim relies on approximation of standard normal distribution
 - Alternatives: exact logistic regression, Firth correction, and Bayesian logistic regression

Thank you for your attention

Questions?