

Motivating example
Design considerations
Adding flexibility to pre-planned adaptive designs

Treatment selection in multi-arm, multi-stage clinical studies

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Outline

Motivating example
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1 Motivating example

2 Design considerations

3 Adding flexibility to pre-planned adaptive designs

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MRC Hubs for Trials Methodology Research
North West Hub

Medical and Pharmaceutical Statistics Research Unit

A multi-arm phase II trial

TAILOR: Telmisartan And Insulin Resistance in HIV.

Ambition: Reduce insulin resistance in HIV patients receiving antiretroviral therapy.

Treatment: 4 different doses of a licensed drug (in a different therapeutic area). Inappropriate to assume a monotone dose-response relationship.

Endpoint: Change in insulin resistance as measured using HOMA-IR index (baseline - week 12).



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Testing multiple hypothesis

Responses: $X_{k,i} \sim N(\mu_k, \sigma^2)$, $i = 1, \dots, n$, $k = 0, 1, \dots, 4$

$$\begin{aligned} H_1 : \quad & \mu_1 \leq \mu_0 \\ \text{Individual null hypotheses:} \quad & \vdots \quad \vdots \\ H_4 : \quad & \mu_4 \leq \mu_0 \end{aligned}$$

$$\text{Test statistics: } Z_k = \frac{\bar{X}_k - \bar{X}_0}{\sigma \sqrt{\frac{2}{n}}} \text{ for } k = 1, \dots, 4$$

Familywise error rate (FWER):
 $P(\text{reject at least one true } H_k) \leq \alpha$

Analyses: J analysis planned



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Motivating example	Framework
Adding flexibility to pre-planned adaptive designs	Pre-planned adaptive design
	Fully flexible designs

Design options

- Pre-planned adaptive designs (e.g. Stallard & Todd, 2003; Magirr et al, 2012)

- + Sufficient statistics
- Unexpected modifications difficult
- + Analytic sample sizes

- Flexible adaptive designs (e.g. Bretz et al, 2006)

- + Very flexible
- Often not based on sufficient statistics

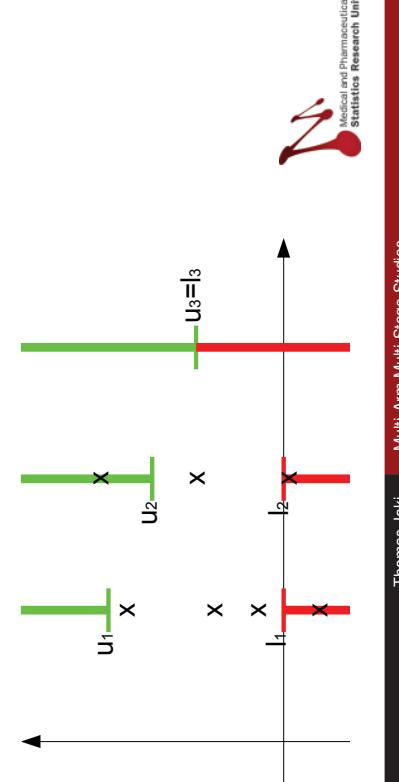


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At interim analysis j

- if $Z_{k,j} < l_j$: treatment k is dropped from trial.
- if $Z_{k,j} > u_j$: can reject H_k and stop trial.



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More Multiple Testing

- J -stage trial \Rightarrow up to $4J$ hypothesis tests.

Strong control of FWER

$$P(\text{reject at least one true } H_k) \leq \alpha$$

Weak control of FWER

$$P(\text{reject at least one true } H_k \mid H_G) \leq \alpha$$

Fact: for this design, Strong control of FWER \Leftrightarrow Weak control of FWER (Magirr et al, 2012).



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Computing $P(\text{reject at least one true } H_k \mid H_G)$

Problem: Test statistics are correlated due to the common control.

Solution: Condition on $\hat{\mu}_{0,J}$, the vector of sample means on control.

$$\alpha = 1 - \underbrace{\int_{-\infty}^{\infty} \cdots \int_{-\infty}^{\infty}}_{J \text{ times}} \left[\sum_{j=1}^J P \left\{ \left(\bigcap_{i=1}^{j-1} B_{1,i} \right) \cap A_{1,j} \mid \hat{\mu}_0, H_G \right\} \right]^K dF(\hat{\mu}_0)$$

- $2J - 1$ unknowns ($l_1, \dots, l_{J-1}, u_1, \dots, u_J$).



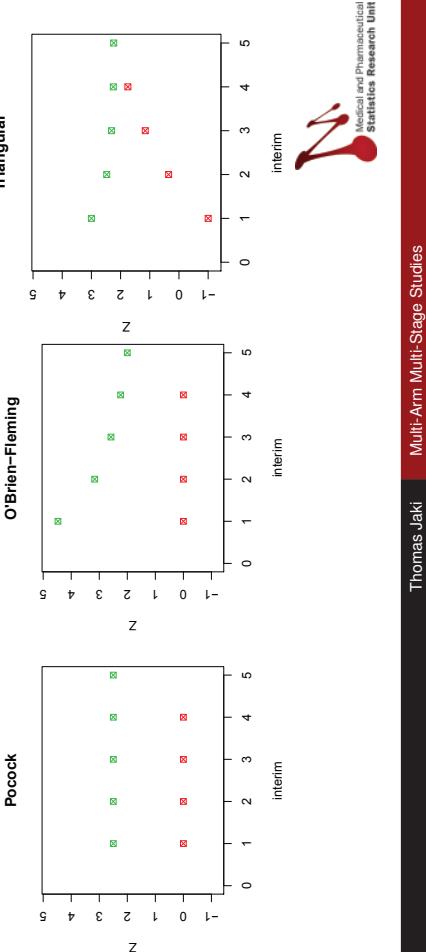
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Motivating example	Framework
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Boundary Constraints

For $J > 1$ set $l_h = g(u_J)$ and $u_h = f(u_J)$, $h = 1, \dots, J - 1$.



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Example

Table: TALLoR trial with $1 - 3$ stages. Boundaries and sample size for $\alpha = 0.05$ (one-sided) and $\beta = 0.1$ and equal sample size per arm and stage (n) is used. Total sample size is denoted \mathcal{N} .

J	Design	I	u	n	$E(\mathcal{N} H_G)$	$E(\mathcal{N} IFC)$
1	Fixed	$-\infty$	2.16	84	420	420
	OBF	(0, 2.17)	(3.07, 2.17)	44	342.3	346.9
	P	(0, 2.38)	(2.38, 2.38)	50	382.9	318.9
2	T	(0.81, 2.29)	(2.43, 2.29)	50	309.4	302.0
	OBF	(0, 0, 2.18)	(3.78, 2.67, 2.18)	31	307.6	317.3
	P	(0, 0, 2.48)	(2.48, 2.48, 2.48)	37	359.2	286.3
3	T	(0, 1.44, 2.34)	(2.71, 2.39, 2.34)	36	292.5	285.2
	OBF	(0, 0, 2.18)	(3.78, 2.67, 2.18)	31	307.6	317.3
	P	(0, 0, 2.48)	(2.48, 2.48, 2.48)	37	359.2	286.3



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Fully flexible designs

Use two fundamental concepts:

- p-value combination
- closed testing

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Choices, choices, choices

Advantages of pre-planned adaptive tests

- ① Sufficient statistics
- ② Sample size calculations
- ③ Confidence intervals

BUT treatment selection is typically more complex than a pre-planned rule so do we need to use flexible adaptive designs?

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Conditional error, König et al, 2008

- The conditional error, $A(X)$, is the maximal probability under H of rejecting H with the original test, conditional on the interim data X
- $B(X)$ is the conditional error for a new test following an unplanned adaption.
- If $B(X) \leq A(X)$ the new test controls the FWER.
- Can be used to update boundaries to reflect a reduced number of treatments being continued

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Simulation

- Compare pre-planned adaptive (without and with conditional error adjustment) and fully flexible design
- No early stopping for futility (no selection) in design
- Selection rule: All treatment that are within ϵ of best performing
- $K = 3$ or 4 and $J = 3$
- 100,000 simulation runs, $\alpha = 0.025$
- 1 treatment with an effect of 1 standard deviation

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Motivating example	Design considerations
Adding flexibility to pre-planned adaptive designs	

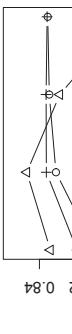
Simulation

Figure: Simulated type I error and power for the pre-planned adaptive method without (\circ) and with conditional error method (+) and the fully flexible design (Δ).

$K = 3$



$K = 4$



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Discussion

- Different ideas to design multi-arm multi-stage trials discussed
- Introduced a strategy to add flexibility to pre-planned adaptive designs
- Conditional error used in unconventional way - traditionally used for data-dependent adaptations (e.g. sample-size re-estimation)
- Should be used as a back-up plan rather than the planned test
- Design of pre-planned adaptive designs available in MAMS package

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Motivating example	Motivation
Design considerations	References
Adding flexibility to pre-planned adaptive designs	
References	
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