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# Optimal designs for group-sequential biomarker enrichment oncology trials

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# Oncology Trials (State-of-the-art)

- High attrition rate (90%)
- Paradigm shift in cancer research: classify disease on underlying molecular biology
- Measures
  - Integrated translational research: biological question integrated into study design.
  - Biomarkers: pair biological measurements with clinical outcomes.
  - Stratified / Personalised Medicine: *enhance efficacy, reduce futility, cut costs.*

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## **Biomarkers**

- prognostic: predicts how a disease may develop in an individual regardless of the type of treatment.
- predictive: provides an indication of the probable effect of treatment on patient.

Focus on predictive biomarkers since the goal of the biomarker-based designs is to establish whether the biomarker predicts treatment response.

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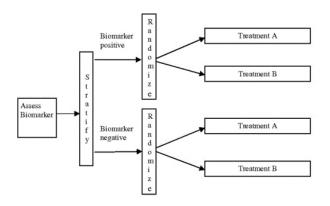
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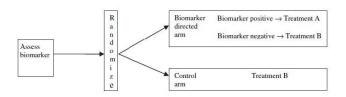
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#### INTERACTION (BIOMARKER-STRATIFIED) DESIGN



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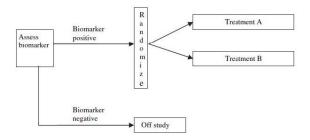
## BIOMARKER-STRATEGY DESIGN WITH STANDARD CONTROL



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## TARGETED (SELECTION) / ENRICHMENT DESIGN



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# Issues with enrichment design

- No info about biomarker —ve patients.
- ② Design may not be efficient if drug has at least some activity in biomarker —ve patients.
- 3 Effect in biomarker —ve patients may never be known.
- Study would provide no new clinical evidence w.r.t. biomarker -ve patients.

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#### **Hypotheses**

$$H_0: p \le p_0$$
  $H_1: p \ge p_1(>p_0)$   
 $p \sim \text{true response rate}$ 

#### Fix

- $p_0 = \text{maximum unacceptable response rate}$
- $p_1$  = minimum acceptable response rate
- $\alpha = \text{desired type I error}$
- $\beta = \text{desired type II error}$

 $X_1, X_2$  are the number of responders in stage 1, 2 resp., binomially distributed.

R. Simon, Cont. Clin. Trials, 1989

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#### Strategy

- recruit  $n_1$  patients at start
- if  $\leq r_1$  responders at stage 1, stop for futility
- if  $> r_1$  responders at stage 1, continue to recruit up to n patients
- reject  $H_0$  if  $\geq r$  responders

Probability of rejecting  $H_0$ 

= 1 - Probability of NOT rejecting  $H_0$ 

$$=1-\left[P(X_1\leq r_1)+\sum_{i=r_1+1}^{min(r,n_1)}P(X_1=i)P(X_2\leq r-i)
ight]$$

where  $P(X_1 \le r_1)$  is the probability of early termination, PET(p)

Note: stop for efficacy if  $\geq r$  responders at first stage

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Expected sample size

$$E(n|p) = n_1 + [1 - PET(p)](n - n_1)$$

OPTIMALITY  $\sim$  acceptable design (satisfying the desired  $\alpha$  and  $\beta$  error rates) with smallest expected sample size under  $H_0$ 

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## **Example: Adaptive Biomarker Enrichment**

#### Phase II targeted cancer therapy

- Determine whether drug has activity only in target population or the general population
- Outcome is (RECIST) tumour response
- Single-arm trial
- Enrichment adaptation based on Simon two-stage design

Jones & Holmgren, Cont. Clin. Trials 2007

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# Hypotheses (Group-sequential)

$$H_0^-: p^- = p_0^-, \qquad H_0^+: p^+ = p_0^+, \ H_1^-: p^- = p_1^-, \qquad H_1^+: p^+ = p_1^+$$
Assume  $p^- < p^+$ 

- Conclude efficacy in unselected if we reject  $H_0^-$
- Conclude efficacy in biomarker positive if we reject  $H_0^+$

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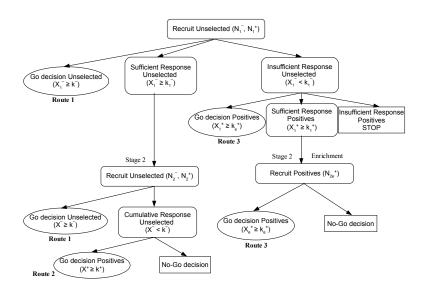
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## Design Schematic



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# Issues with J & H Design

- Hypotheses testing not properly addressed
- 2 Type I error and power calculations wrong
- Expected sample size formulae do not take into account early stopping for efficacy
- Designs not optimal

Our work rectifies these issues and provides a robust framework for adaptive enrichment

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## **Error Probabilities**

#### Type I error

- $R_1(p^-) \sim$  probability of rejecting  $H_0^-$  via Route (1)
- $R_2(p^-, p^+) \sim$  probability of rejecting  $H_0^+$  via Route (2) (non-monotonic)
- $R_3(p^-, p^+) \sim$  probability of rejecting  $H_0^+$  via Route (3) (Enrichment)

$$R_{23}() = R_2() + R_3()$$
  
 $R_{123}() = R_1() + R_2() + R_3() \le \alpha$ 

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## **Error Probabilities**

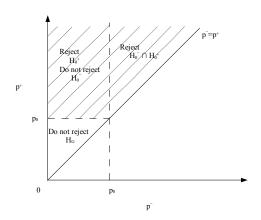
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# Parameter space of $(p^-, p^+)$



Generic null hypotheses:  $H_0^-:p^-\leq p_0,\ H_0^+:p^+\leq p_0$  Global null  $H_G:H_0^-\cap H_0^+;\ p^-\leq p^+$ 

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## The formulae

$$R_1(p^-) = \left(\sum_{i=k_1^-}^{min(N_1^-,k^--1)} P(X_2^- \ge k^- - i) P(X_1^- = i)\right) + P(X_1^- \ge k^-)$$

$$R_2(p^-,p^+) = P(X^+ \ge k^+) \left( \sum_{i=k_1^-}^{min(N_1^-,k^--1)} P(X_2^- < k^- - i) P(X_1^- = i) \right)$$

$$\begin{array}{lcl} R_{3}(p^{-},p^{+}) & = & P(X_{1}^{-} < k_{1}^{-}) \times \\ & & \left( \left\{ \sum_{i=k_{1}^{+}}^{min(N_{1}^{+},k_{e}^{+}-1)} P(X_{2e}^{+} \geq k_{2e}^{+}-i) P(X_{1}^{+}=i) \right\} + P(X_{1}^{+} \geq k_{e}^{+}) \right) \end{array}$$

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## Errors all the way

	Outcomes		
Real World	R <sub>0</sub> No Efficacy	$R_1$ Unselected	R <sub>23</sub> Positive only
$W_0$ : No Efficacy $(p_0^-, p_0^+)$	$R_0( ho_0^-, ho_0^+)$ True negative	$R_1(p_0^-)$ False positive	$R_{23}(p_0^-,p_0^+)$ False positive
$W_1$ : Unselected $(p_1^-, p_1^-)$	$R_0(p_1^-,p_1^-)$ False negative	$R_1( ho_1^-)$ True positive	$R_{23}(p_1^-,p_1^-)$ Wrong positive
$W_{23}$ : Positive only $(p_0^-, p_1^+)$	$R_0( ho_0^-, ho_1^+)$ False negative	$R_1(p_0^-)$ Wrong positive	$R_{23}( ho_0^-, ho_1^+)$ True positive

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#### Power

$$min(R_1(p_1^-), R_{23}(p_0^-, p_1^+)) \ge 1 - \beta$$

Family Wise Error Rates (Type I)

- Family:= set of null hypotheses  $\{H_{01}, H_{02}, \ldots\}$
- ullet  $V\sim$  number of incorrectly rejected  $H_0$ 's

#### Then, FWER

- $\bullet = P(V \ge 1) = 1 P(V = 0)$
- ~ probability of making at least one type I error in the family (i.e. rejecting any of the null hypotheses)
- can be weak, strong, with / without Individual Outcome (IO) control (i.e. 4 options)

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## Proposed Type I error control

```
Option 1 (Weak FWER): \sum_i FP_i \leq \alpha
Option 2 (Strong FWER): max(\sum_i FP_i, WP_1, WP_2) \leq \alpha
Option 3 (Weak IH): max(FP_1, FP_2) \leq \alpha/2
Option 4 (Strong IH):
```

- $max(FP_1, FP_2) \leq \alpha/2$
- $max(WP_1, WP_2) \le \alpha$

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## Weak FWER

- probability of type I error in  $R_1$  or  $R_{23}$
- reject  $H_0^-$  (via  $R_1$ ) or reject  $H_0^+$  (via  $R_{23}$ )
- $\sum$  (False positives)  $\leq \alpha$

$$R_1(p_0^-) + R_2(p_0^-, p_0^+) + R_3(p_0^-, p_0^+) \le \alpha$$

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# **Expected Sample Size**

$$E(N) = N_1^- + N_1^+ + N_2[1 - PET(p^-)] + N_{2e}^+ P(X_1^- < k_1^-)[(1 - PET(p^+))]$$

where

$$PET(p^{-}) = P(X_{1}^{-} < k_{1}^{-}) + P(X_{1}^{-} \ge k^{-})$$

$$PET(p^{+}) = P(X_{1}^{+} < k_{1}^{+}) + P(X_{1}^{+} \ge k_{n}^{+})$$

Overall probability of early termination

$$PET = P(X_1^- \ge k^-) + P(X_1^- < k_1^-)[P(X_1^+ \ge k_e^+) + P(X_1^+ < k_1^+)]$$

Note:  $PET \neq PET(p^{-}) + PET(p^{+})$ 

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# Results (J & H)

#### Operating characteristics given $k_1^- = 2, k_1^+ = 1, k^- = 4, k_e^+ = 4, k_e^+ = 5$

p <sup>-</sup>	p <sup>+</sup>	Power: Route 1	Power: Routes 2 + 3	$E(N)_{Simon}$	$E(N)_{Adaptive}$	E(N) <sub>Adaptive</sub> /
		(unselected)	(positives)			E(N) <sub>Simon</sub>
0.03	0.03	0.067	0.012	74.61	65.79	0.881
0.03	0.10	0.067	0.373	85.21	76.91	0.902
0.03	0.15	0.067	0.624	88.36	80.21	0.907
0.10	0.15	0.755	0.624	127.66	80.03	0.626
0.10	0.25	0.755	0.807	129.78	80.44	0.619
0.15	0.30	0.952	0.852	136.99	80.10	0.584

Significance,  $\alpha = 0.079$ 

#### Design

$$(k_1^-k_1^+)/(N_1^-N_1^+) \to (k_e^+/N_{2e}^+)|(k^-k^+)/(N^-N^+)$$
  
(2 1)/(34 14)  $\to$  (5/50) | (4 4)/(53 27)

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# Optimal design results

Weak FWER (
$$p_0 = 0.03, \alpha = 0.05, \beta = 0.2$$
)

$ ho_1^-$	$ ho_1^+$	ESS	Design $(k_1^-k_1^+)/(N_1^-N_1^+) \to (k_e^+/N_{2e}^+) (k^-k^+)/(N^-N^+) $
0.10	0.10	110.2	$(3\ 2)/(44\ 34) \to (7/104) \mid (9\ 4)/(135\ 53)$
0.10	0.15	77.9	$(2\ 2)/(32\ 21)  o (6/67) \mid (7\ 3)/(106\ 29)$
0.10	0.25	59.9	$(2\ 1)/(34\ 8) \to (4/29) \mid (6\ 2)/(87\ 9)$
0.15	0.15	46.9	$(2\ 1)/(20\ 12)  o (4/43) \mid (6\ 2)/(66\ 21)$
0.15	0.25	32.5	$(1\ 1)/(12\ 7) \to (4/28) \mid (4\ 2)/(43\ 11)$
0.15	0.35	27.8	$(1\ 1)/(11\ 5) \to (3/15) \mid (4\ 2)/(47\ 7)$
0.25	0.25	18.4	$(1\ 1)/(6\ 6) \to (3/24) \mid (3\ 2)/(23\ 13)$
0.25	0.40	13.4	$(1\ 1)/(6\ 4) \to (2/9) \mid (3\ 2)/(23\ 5)$

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## Conclusions

- Extension to randomised Phase II / III trials
- Study different outcomes: PFS, OS
- Weak FWER: smaller study with sufficient control for early phase trial.
- Strong FWER: late stage definitive study.
- Choice of error control: what clinicians / trialists prefer

An optimal stratified Simon two-stage design D. Parashar, J. Bowden, C. Starr, L. Wernisch, A. Mander Pharm. Statistics (2016), doi: 10.1002/pst.1742.



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