Design and Sample Size Determination for Multipledose Randomized Phase II Trials for Dose Optimization

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► Background – Project Optimus

- In 2022, FDA initiated Project Optimus "to reform the dose optimization and dose selection paradigm in oncology drug development."
- Paradigm shifting from maximum tolerated dose (MTD) to optimal biological dose (OBD).



Targeted Therapies



 MTD-based dose finding is often appropriate to inform RP2D Safety alone is not sufficient to inform optimal RP2D

Background – Design strategies to find OBD

- Trial designs to compare multiple dosages (FDA, 2023)
 - The trial should be sized to allow for sufficient assessment of activity, safety, and tolerability for each dosage.
 - The trial does not need to be powered to demonstrate statistical superiority of a dosage or statistical non-inferiority among the dosages



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Background – Two-stage decision-making paradigm



Determination of OBD admissible dose set

 The objective is to identify a dose set that satisfy certain safety and efficacy requirements (i.e., OBD admissible set A).

Identification of the OBD

- The OBD will be selected from A based on the totality of activity, safety and tolerability data.
 - "Relevant nonclinical and clinical data, as well as the dose- and exposure- response relationships for safety and efficacy should be evaluated to select a dosage(s) for clinical trial(s)" (FDA guidance)
 - Unlikely/impossible to formulate statistical decision rules to capture all quantitative and qualitative considerations relevant to the final OBD selection

Background – Two-stage decision-making paradigm



• The objective is to identify a dose set that satisfy certain safety and efficacy requirements (i.e., OBD admissible set **A**).



Goal: formulize the deign in this step to ensure the trial satisfies certain statistical properties, including type I error and power.

└→ Statistical modelling – set ups

- Consider a multiple-dose randomized trial, where a total of $J \times n$ patients are equally randomized to J doses, $d_1 < d_2 < \cdots < d_I$.
 - In most applications, J = 2 or 3, and d_J is often the MTD or maximum administered dose.



- Let Y_T and Y_E denote the binary toxicity and efficacy endpoints, respectively.
 - Example of *Y_T*: dose-limiting toxicity, dichotomized total toxicity burden.
 - Example of *Y_E*: objective responses, efficacy surrogate endpoints.

└→ Statistical modelling – assumptions

- Let $\pi_{T,j} = \Pr(Y_T = 1|d_j)$ and $\pi_{E,j} = \Pr(Y_E = 1|d_j)$ denotes the probability of the occurrence of toxicity and efficacy events.
- Assuming that $\pi_{T,j}$ and $\pi_{E,j}$ are **non-decreasing** with respect to the increasing of dose levels.
 - Randomized dose optimization trials have the same drug but with ordered doses
- For toxicity endpoint, we assume
 - \$\phi_{T,0}\$: the null toxicity rate that is high and deemed unacceptable;
 - φ_{T,1}: the alternative toxicity rate that is low and deemed acceptable;
- For efficacy endpoint, we let
 - $\phi_{E,0}$: the **null efficacy** rate that is **low** and deemed unacceptable;
 - $\phi_{E,1}$: the alternative efficacy rate that is high and deemed acceptable



OBD admissible

Statistical modelling – global type I error

- Consider H_0 : None of the doses is the OBD.
 - **Challenge**: consists of multiple hypotheses $(\sum_{j=1}^{J+1} j)$.

$$H_{0}(s,k): \underbrace{\pi_{T,1} = \pi_{T,2} = \cdots = \pi_{T,s} = \phi_{T,1}}_{\text{acceptable toxicity}} < \underbrace{\pi_{T,s+1} = \cdots = \pi_{T,k} = \pi_{T,k+1} = \cdots = \pi_{T,J} = \phi_{T,0}}_{\text{unacceptable toxicity}};$$

$$\underbrace{\pi_{E,1} = \pi_{E,2} = \cdots = \pi_{E,s} = \pi_{E,s+1} = \cdots = \pi_{E,k} = \phi_{E,0}}_{\text{unacceptable efficacy}} < \underbrace{\pi_{E,k+1} = \cdots = \pi_{E,J} = \phi_{E,1}}_{\text{acceptable efficacy}},$$

where
$$s, k \in \{0, 1, \dots, J\}$$
 with $s \leq k$.

		d_1	d_2			d_1	d_2
H ₀ (0,0)	Тох	$\phi_{T,0}$	$\phi_{T,0}$	<i>H</i> ₀ (0,0)	Тох	$\phi_{T,1}$	$\phi_{T,0}$
	Eff	$\phi_{E,1}$	$\phi_{E,1}$		Eff	$\phi_{E,0}$	$\phi_{E,1}$
H ₀ (0,1)	Тох	$\phi_{T,0}$	$\phi_{T,0}$	$H_0(0,1)$	Тох	$\phi_{T,1}$	$\phi_{T,0}$
	Eff	$\phi_{E,0}$	$\phi_{E,1}$		Eff	$\phi_{E,0}$	$\phi_{E,0}$
H ₀ (0,2)	Тох	$\phi_{T,0}$	$\phi_{T,0}$	μ (0.2)	Тох	$\phi_{T,1}$	$\phi_{T,1}$
	Eff	$\phi_{E,0}$	$\phi_{E,0}$	$H_0(0,2)$	Eff	$\phi_{E,0}$	$\phi_{E,0}$

Statistical modelling – global type I error

- Consider H_0 : None of the doses is the OBD.
 - **Challenge**: consists of multiple hypotheses $(\sum_{j=1}^{j+1} j)$.

$$H_{0}(s,k): \underbrace{\pi_{T,1} = \pi_{T,2} = \cdots = \pi_{T,s} = \phi_{T,1}}_{\text{acceptable toxicity}} < \underbrace{\pi_{T,s+1} = \cdots = \pi_{T,k} = \pi_{T,k+1} = \cdots = \pi_{T,J} = \phi_{T,0}}_{\text{unacceptable toxicity}};$$

$$\underbrace{\pi_{E,1} = \pi_{E,2} = \cdots = \pi_{E,s} = \pi_{E,s+1} = \cdots = \pi_{E,k} = \phi_{E,0}}_{\text{unacceptable efficacy}} < \underbrace{\pi_{E,k+1} = \cdots = \pi_{E,J} = \phi_{E,1}}_{\text{acceptable efficacy}},$$

where $s, k \in \{0, 1, \dots, J\}$ with $s \leq k$.

• Define *global type I error* that encompasses all $H_0(s, k)$ as follows:

$$\alpha = \Pr(\text{reject } H_0 | H_0)$$

$$= \max_{0 \le s \le J, \ s \le k \le J} \{\alpha(s,k)\}.$$

where $\alpha(s, k) = \Pr(\operatorname{reject} H_0(s, k) | H_0(s, k)).$

Statistical modelling – generalized powers

- Consider H₁: At lease one does is the OBD admissible.
 - **Challenge**: consists of multiple hypotheses ($\sum_{j=1}^{J} j$).

$$H_{1}(u,v): \underbrace{\pi_{T,1} = \pi_{T,2} = \dots = \pi_{T,u} = \pi_{T,u+1} = \dots = \pi_{T,v} = \phi_{T,1}}_{\text{acceptable toxicity}} < \underbrace{\pi_{T,v+1} = \dots = \pi_{T,J} = \phi_{T,0}}_{\text{unacceptable toxicity}};$$

$$\underbrace{\pi_{E,1} = \pi_{E,2} = \dots = \pi_{E,u} = \phi_{E,0}}_{\text{unacceptable efficacy}} < \underbrace{\pi_{E,u+1} = \dots = \pi_{E,v} = \pi_{E,v+1} = \dots = \pi_{E,J} = \phi_{E,1}}_{\text{acceptable efficacy}},$$

where $u, v \in \{0, 1, \dots, J\}$ with u < v.

		d_1	d_2		
U(0.1)	Тох	$\phi_{T,1}$	$\phi_{T,0}$	-	
$\pi_1(0,1)$	Eff	$\phi_{E,1}$	$\phi_{E,1}$		· OPD admissible
	Тох	$\phi_{T,1}$	$\phi_{T,1}$	-	
$H_1(0,2)$	Eff	$\phi_{E,1}$	$\phi_{E,1}$		
U(1,2)	Тох	$\phi_{T,1}$	$\phi_{T,1}$	-	
$H_1(1,2)$	Eff	$\phi_{E,0}$	$\phi_{E,1}$		

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Statistical modelling – generalized powers

- Consider H_1 : At lease one does is the OBD admissible.
 - **Challenge**: consists of multiple hypotheses ($\sum_{i=1}^{J} j$).
 - Additional challenge: the standard definition of power, which reject the H_0 , is not sufficient to account for the characteristics of dose optimization.

		d_1	<i>d</i> ₂
$\mathcal{U}(0,1)$	Тох	$\phi_{T,1}$	$\phi_{T,0}$
$\pi_1(0,1)$	Eff	$\phi_{E,1}$	$\phi_{E,1}$
		\checkmark	
Selection			\checkmark
		\checkmark	\checkmark

It is important to account for the quality of the admissible dose selection!

Statistical modelling – generalized powers

• Generalized power I:

 $\beta_1(u, v) = \Pr\left(\begin{array}{c} \operatorname{reject} H_0 \text{ and all dose in } A \text{ are truly} \\ \operatorname{safe and efficacious} \end{array} \middle| H_1(u, v) \right).$

where A denotes the admissible dose set selected by the design.

• Generalized power II:

 $\beta_2(u, v) = \Pr\left(\begin{array}{c} \operatorname{reject} H_0 \text{ and at leaset one dose in } A \text{ are truly} \\ \text{safe and efficacious} \end{array} \middle| H_1(u, v) \right).$

	eta_1	β_2
Tolerate false positive	×	\checkmark

- Both generalized powers are stricter than the standard power.
- The choice of power depends on the characteristics of the trial and user's tolerability of false positive.
- Power II is a good option when reducing the sample size is of top priority.

Statistical modelling – generalized powers

- Consider *H*₁: At lease one does is the OBD admissible.
 - **Challenge**: consists of multiple hypotheses ($\sum_{j=1}^{J} j$).

$$H_{1}(u,v): \underbrace{\pi_{T,1} = \pi_{T,2} = \cdots = \pi_{T,u} = \pi_{T,u+1} = \cdots = \pi_{T,v} = \phi_{T,1}}_{\text{acceptable toxicity}} < \underbrace{\pi_{T,v+1} = \cdots = \pi_{T,J} = \phi_{T,0}}_{\text{unacceptable toxicity}};$$

$$\underbrace{\pi_{E,1} = \pi_{E,2} = \cdots = \pi_{E,u} = \phi_{E,0}}_{\text{unacceptable efficacy}} < \underbrace{\pi_{E,u+1} = \cdots = \pi_{E,v} = \pi_{E,v+1} = \cdots = \pi_{E,J} = \phi_{E,1}}_{\text{acceptable efficacy}},$$

where $u, v \in \{0, 1, ..., J\}$ with u < v.

 We define global power I and power II that encompasses all H₁(u, v) as follows

$$\beta_i = \min_{u,v \in \{0, \dots, J\}, u < v} \beta_i(u, v), \text{ for } i = 1, 2.$$

Statistical modelling – least favorable set

Theorem 1. Define the least favorable set
$$\widetilde{H}_1 = \{H_1(j), j = 1, \cdots, J\}$$
, where

$$H_1(j) = \begin{pmatrix} \pi_{T,1} = \cdots = \pi_{T,j-1} = \phi_{T,1} \\ \pi_{E,1} = \cdots = \pi_{E,j-1} = \phi_{E,0} \\ safe but futile \end{pmatrix} \xrightarrow{\pi_{T,j} = \phi_{T,1}}_{safe and efficacious} \xrightarrow{\pi_{T,j+1} = \cdots = \pi_{T,J} = \phi_{T,0}}_{toxic and efficacious} \end{pmatrix}$$

For any $H_1(u,v)$, with $u,v \in \{0,1,2,\ldots,J\}$ and u < v, there exists an $H_1(j)$ such that $\beta_i(j) \leq \beta_i(u,v), i = 1, 2, \text{ where } \beta_1(j) \text{ and } \beta_2(j) \text{ denote the generalized power I and II under }$ $H_1(j)$, respectively.

toxic and efficacious

The global power can be equivalently defined as follows

$$\beta_i = \min_{j \in \{1, \dots, J\}} \beta_i(j)$$
 for $i = 1, 2$.

► MERIT Design

- MERIT (Multiple-dosE RandomIzed Phase-II Trial)
 - Specify target global type I error α^* and power β^* ;
 - Randomize J×n patients equally to J dosing arm;
 - In any dose arm d_j , if $n_{E,j} \ge m_E$ and $n_{T,j} \le m_T$, we reject H_0 and claim that d_j is OBD admissible, where m_T and m_E are decision boundaries for toxicity and efficacy, respectively.

 ${}^*n_{E,j}$ and $n_{T,j}$ are the total number of patients who experience efficacy and toxicity events in dose arm d_j .

└→ Optimal Design Parameters

- The optimal design parameters (n, m_T, m_E) are determined through numerical search such that the design controls the global type I error and power at nominal values α^{*} and β^{*}.
- Type I error

$$\begin{aligned} \alpha(s,k) &= \Pr(\text{reject } H_0(s,k) | H_0(s,k)) \\ &= 1 - \left\{ \left(1 - \Pr(n_T \le m_T, n_E \ge m_E; n, \phi_{T,1}, \phi_{E,0}) \right)^s \\ &\times \left(1 - \Pr(n_T \le m_T, n_E \ge m_E; n, \phi_{T,0}, \phi_{E,0}) \right)^{k-s} \\ &\times \left(1 - \Pr(n_T \le m_T, n_E \ge m_E; n, \phi_{T,0}, \phi_{E,1}) \right)^{J-k} \right\}. \end{aligned}$$

Power

$$\beta_1(j) = \Pr(n_{E,1} < m_E, \dots, n_{E,j-1} < m_E, n_{T,j+1} > m_T, \dots, n_{T,j} > m_T,$$
$$n_{E,j} \ge m_E, n_{T,j} \le m_T \ |H_1(j)),$$
$$\beta_2(j) = \Pr(n_{E,j} \ge m_E, n_{T,j} \le m_T \ |H_1(j))$$

Practice Consideration

- In small samples, isotonic transformed $\{\tilde{n}_{T,j}, \tilde{n}_{E,j}\}$ should be applied before comparing with boundaries (m_T, m_E) when non-decreasing assumption is sound for toxicity and efficacy.
- In some trials, it maybe desirable to add futility and safety interim monitoring
 - Stop arm *j* for safety if $Pr(\pi_{T,j} > \phi_{T,1} | data) > C_T$,
 - Stop arm *j* for futility if $Pr(\pi_{E,j} < \phi_{E,1} | data) > C_E$, where C_T and C_E are probability cutoffs (e.g. 0.95).
- Whether to including interim monitoring depends on the availability of Y_T and Y_E, and other considerations. Typically, one or two interims are sufficient.

└→ Optimal Design Parameters

• Optimal design parameters (n, m_T, m_E) of MERIT design, when J = 2, $(\phi_{T,0}, \phi_{T,1}) = (0.4, 0.2)$, and $(\phi_{E,0}, \phi_{E,1}) = (0.2, 0.4)$.

			β	1					β	2		
eta^*	$\alpha^* = 0.1 \qquad \qquad \alpha^* = 0.2$			$\alpha^* = 0.1$			$\alpha^* = 0.2$					
	n	m_T	m_E	n	m_T	m_E	n	m_T	m_E	п	m_T	m_E
0.6	26	7	6	23	6	26	25	6	5	18	5	4
0.7	33	9	7	30	8	33	33	8	6	24	7	5
0.8	44	12	8	39	11	44	39	11	8	30	8	5

Operating Characteristics – under power I

• Type I error and power of MERIT design when $(\phi_{T,0}, \phi_{T,1}) = (0.4, 0.2)$, and $(\phi_{E,0}, \phi_{E,1}) = (0.2, 0.4)$.



Operating Characteristics – under power II

• Type I error and power of MERIT design when $(\phi_{T,0}, \phi_{T,1}) = (0.4, 0.2)$, and $(\phi_{E,0}, \phi_{E,1}) = (0.2, 0.4)$.



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MERIT: Multiple-dose Randomized Phase II Trial Design for Dose Optimization and Sample Size Determination

PID: 1126; Version: V1.1.0.0 ; Last Updated: 2/18/2023

Peng Yang and Ying Yuan

Trial Setting Operating Characteristics Trial Conduct Reference **MERIT Design** 0 Number of Doses: ○ 2 ○ 3 ○ 4 **Toxicity Rates:** Null $\phi_{T,0}$ Alternative $\phi_{T,1}$ 0.4 0.2 **Efficacy Rates:** Null $\phi_{E,0}$ Alternative $\phi_{E,1}$ 0.2 0.4

Department of Biostatistics, The University of Texas MD Anderson Cancer Center

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Global Type I Error Rate:	0
0.2	
Generalized Power: Power I Power II 	
0.8	
Inlucde toxicity and futility monitoring	



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Peng Yang and Ying Yuan

Department of Biostatistics, The University of Texas MD Anderson Cancer Center

Trial Setting	Operating Characteristics Trial Conduct	R	leference
Number of	Doses:	8	MERIT Design
Toxicity Rat	es:		Download MERIT Design
0.4	0.2		In this trial, the toxicity rates of 0.4 and 0.2 are considered unacceptable and acceptable, respectively, while the efficacy rates of 0.2 and 0.4 are considered unacceptable and acceptable, respectively. In order to control the global Type I error rate at 0.2 and achieve a global generalized power I of 0.8, a minimum sample size of 44 per
Efficacy Ra	tes: Alternative $\phi_{E,1}$		arm is required. The generalized power I is defined as the probability of rejecting the null hypothesis that none of the doses are considered optimal biological doses (OBD) admissible and all doses identified as OBD admissible are truly safe and efficacious given the alternative hypothesis that at least one dose is OBD admissible. At the end of the trial, perform isotonic regression on toxicity and efficacy data across all doses. A dose will be considered OBD admissible if the isotonically transformed number of toxicity $z = 13$ and the isotonically
0.2	0.4		transformed number of efficacy >= 13.

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Global Type I Error Rate:		Global Type I Error Rate:			0
0.2		0.2			
Generalized Power: ● Power I ○ Power II		Generalized Power: Power I Power II			
0.8		0.8			
Inlucde toxicity and futility monitoring		Inluce toxicity and futility me	nitorin	g	
		Interim Times:			0
		Input the fraction of the total separated by space.	sample	e size at interims,	
		Efficacy		Toxicity	
		1/2		1/3 2/3	
	•	Stopping Criteria:Stop for futility if $p(\pi_{E,j} < \phi_E)$ 0.95Stop for toxicity if $p(\pi_{T,j} > \phi)$ 0.95	₁ data _{1,1} data	$a(t) > C_E$, where C_E $b(ta) > C_T$, where C_T	

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Num	ber o	of Doses	:
0 2	03	0 4	

Toxicity Rates:

Null $\phi_{T,0}$	Alternative $\phi_{T,1}$
0.4	0.2
Efficacy Rates:	
Null $\phi_{E,0}$	Alternative $\phi_{E,1}$
0.2	0.4

Global Type I Error Rate:	0
0.2	
Generalized Power: Power I O Power II	
0.8	
Inlucde toxicity and futility monitoring	
Interim Times:	0

MERIT Design

0

🛃 Download MERIT Design

Design Description

In this trial, the toxicity rates of 0.4 and 0.2 are considered unacceptable and acceptable, respectively, while the efficacy rates of 0.2 and 0.4 are considered unacceptable and acceptable, respectively. In order to control the global Type I error rate at 0.2 and achieve a global generalized power I of 0.8, a minimum sample size of 45 per arm is required. The generalized power I is defined as the probability of rejecting the null hypothesis that none of the doses are considered optimal biological doses (OBD) admissible and all doses identified as OBD admissible are truly safe and efficacious given the alternative hypothesis that at least one dose is OBD admissible. At the end of the trial, perform isotonic regression on toxicity and efficacy data across all doses. A dose will be considered OBD admissible if the isotonically transformed number of toxicity <= 13 and the isotonically transformed number of efficacy >= 13.

During the trial, the toxicity and efficacy of each dose arm will be monitored independently using the stopping criteria outlined in Table 1. If the isotonically transformed toxicity and efficacy acrosss topping boundaries, enrollment in that particular dose arm will be suspended.

Table 1. Stopping boundaries for toxicity and e CSV Excel PDF Print	fficacy. Sea	rch:
# of patients treated	Stop if # toxicity >=	Stop if # efficacy <=
15	6	NA
23	NA	5
30	10	NA
Showing 1 to 3 of 3 entries Note: 'NA' means that this endpoint will not be	used to make go/no-go decision a	Previous 1 Next t the interim

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MERIT: Multiple-dose Randomized Phase II Trial Design for Dose Optimization and Sample Size Determination

PID: 1126; Version: V1.1.0.0 ; Last Updated: 2/18/2023

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Trial Sett	ing O	peratin	g Chara	acteris	tics	Trial Conduct	Reference	
Enter	Simulatio	n Scen	arios:					Operating Characteristics
Add a scenario				scenario	Save sce	enarios		
For eacl	h scenario,	enter tru	e toxicit	y and eff	icacy rat	e of each dose leve	l:	
		Tox(d1)	Eff(d1)	Tox(d2)	Eff(d2)			
	Scenario 1	0.40	0.40	0.40	0.40			
	Scenario 2	0.40	0.20	0.40	0.20			
	Scenario 3	0.20	0.20	0.20	0.20			
	Scenario 4	0.20	0.40	0.40	0.40			
	Scenario 5	0.20	0.40	0.20	0.40			
	Scenario 6	0.20	0.20	0.20	0.40			
Number	of simulati	ons				Set seed		
5000						123		
			O F	Run Simu	lation			

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Operating Characteristics										
Copy CSV Excel Print						Search:				
Scenarios	$\frac{1}{\nabla}$	Metrics	$\frac{1}{\nabla}$	Values	$\frac{\Delta}{\nabla}$	Average sample size				
1		Type I error		0.075		45				
2		Type I error		0.000		45				
3		Type I error		0.104		45				
4		Power		0.821		45				
5		Power		0.982		45				
6		Power		0.805		45				

Showing 1 to 6 of 6 entries

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Figure 1. Type I error and power of MERIT design when unacceptable and acceptable toxicity rates are 0.4 and 0.2, and unacceptable and acceptable efficacy rates are 0.2 and 0.4. The horizontal dashed lines represent the nominal values of type I error (0.2) and power I (0.8).



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Discussion

- In this project, we proposed a MERIT design for dose optimization by controlling generalized type I error and power while optimizing the sample size.
- The implementation of MERIT is simple, which only involve comparison of the observed number of toxicity and efficacy with prespecified decision boundaries.
- We have developed R shiny app on <u>www.trialdesign.org</u> to facilitate the application of MERIT design.





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Reference

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└→ Thank you

