## A Bayesian Phase I/II Design for Oncology Clinical Trials of Combining Biological Agents

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#### Outline

- Introduction
- Probability model
- Dose finding algorithm

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- Simulation study
- Conclusion

Introduction

Methods Trial design Simulation Conclusions

Biological agents A lymphoma trial

### **Biological agents**

- The paradigm of oncology drug development is expanding from traditional cytotoxic agents to novel biological (or molecularly targeted) agents.
- Examples of biological agents:
  - Biospecimens targeting a specific tumor pathway.
  - Gene products aiming for DNA repair.
  - Immunotherapies stimulating the immune system to attack a tumor.

Biological agents A lymphoma trial

#### Biological agents versus cytotoxic agents

- Cytotoxic agents
  - Toxicity and efficacy are assumed to monotonically increase with dose.
  - The goal is to find the maximum tolerated dose (MTD).
- Biological agents
  - The toxicity is usually tolerable within the therapeutic dose range and may plateau at higher dose levels.
  - The dose-efficacy curves often follow a non-monotonic pattern.
  - The goal is to find the optimal biological dose (OBD), defined as the dose yielding the most desirable treatment effect.

Biological agents A lymphoma trial

#### Drug-combination Trials

- Treating patients with a combination of agents is becoming common in cancer clinical trials.
- Most existing drug-combination trial designs concern cytotoxic agents (e.g., Thall et al., 2003; Wang and Ivanova, 2005; Yin and Yuan, 2009), thus are not applicable to the trials combining biological agents.
- A phase I/II trial design is imperative for biological agent combination trials because of non-monotonic dose-efficacy and -toxicity relationship.

Biological agents A lymphoma trial

## Motivating trial

- A lymphoma trial combining two novel biological agents to target two components in the PI3K/AKT/mTOR signaling pathway.
  - Agent A is a PI3K kinase inhibitor.
  - Agent B inhibits mTOR kinase downstream in the pathway.
- 4 doses of agent A combined with 4 doses of agent B.
- Goal: to find the biologically optimal dose combination (BODC), defined as the dose combination with the highest efficacy and tolerable toxicity.

Introduction

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Biological agents A lymphoma trial

### Motivating trial

## РІЗК Agent A Akt Rheb mTOR Agent B 4E-BP1 S6K1 Cell growth, proliferation, survival, immune regulation, angiogenesis

#### Targeting PI3K/AKT/mTOR signaling Pathways in Lymphoma

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Biological agents A lymphoma trial

#### Proposed design

We propose a phase I/II trial design to identify the BODC.

- A change-line model is used to reflect the property that the dose-toxicity surface of the combinational agents may plateau at higher dose levels.
- A logistic model with quadratic terms is used to accommodate the possible non-monotonic pattern for the dose-efficacy relationship.
- We devise a novel adaptive dose-finding algorithm to encourage sufficient exploration of the two-dimensional dose space.

Model for toxicity and efficacy Likelihood and prior specifications

#### Notation

- Consider a trial of combinational biological agents
  - J doses of agent A:  $a_1 < a_2 < \cdots < a_J$
  - K doses of agent B:  $b_1 < b_2 < \cdots < b_K$
  - $(a_j, b_k)$ : combination of dose  $a_j$  and dose  $b_k$
  - *p<sub>jk</sub>* and *q<sub>jk</sub>* denote the toxicity and efficacy probabilities of dose combination (*a<sub>j</sub>*, *b<sub>k</sub>*)
- Goal: identify the BODC in the  $J \times K$  dose matrix.

Model for toxicity and efficacy Likelihood and prior specifications

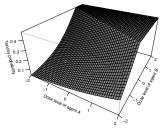
#### Change-line model for toxicity

We model toxicity probability  $p_{jk}$  using a change-line model:

$$logit(p_{jk}) = (\beta_0 + \beta_1 a_j + \beta_2 b_k) I(\beta_0 + \beta_1 a_j + \beta_2 b_k \le \omega) + \omega I(\beta_0 + \beta_1 a_j + \beta_2 b_k > \omega)$$

- I(·): indicator function
- β<sub>1</sub> > 0 and β<sub>2</sub> > 0 such that p<sub>jk</sub> initially increases with the doses of A and B
- When it reaches a plateau, the toxicity probability: e<sup>ω</sup>/(1 + e<sup>ω</sup>).
- We did not include an interactive effect for the two agents because the estimation of that needs large sample

Figure : Surface of the toxicity probabilities



Model for toxicity and efficacy Likelihood and prior specifications

## Logistic model for efficacy

Assume the efficacy probability  $q_{jk}$  follows a logistic model

$$logit(q_{jk}) = \gamma_0 + \gamma_1 a_j + \gamma_2 b_k + \gamma_3 a_j^2 + \gamma_4 b_k^2$$

- The quadratic terms render the model adequate flexibility to capture the non-monotonic pattern.
- We model the marginal distributions of toxicity and efficacy.
- Joint modeling is possible, but small sample size  $\rightarrow$  cannot reliably estimate the correlation parameter.

Model for toxicity and efficacy Likelihood and prior specifications

#### Likelihood

Suppose that at a certain stage of the trial

- $n_{jk}$  patients are treated at the paired dose  $(a_j, b_k)$
- x<sub>jk</sub> and y<sub>jk</sub> patients have experienced toxicity and efficacy, respectively.
- The marginal likelihood for the toxicity data x is

$$\mathcal{L}(\mathbf{x}|\omega,oldsymbol{eta}) \propto \prod_{j=1}^J \prod_{k=1}^K p_{jk}^{x_{jk}} (1-p_{jk})^{n_{jk}-x_{jk}};$$

for the efficacy data  ${\boldsymbol{y}}$  is

$$\mathcal{L}(\mathbf{y}|\boldsymbol{\gamma}) \propto \prod_{j=1}^J \prod_{k=1}^K q_{jk}^{y_{jk}} (1-q_{jk})^{n_{jk}-y_{jk}}.$$

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Model for toxicity and efficacy Likelihood and prior specifications

• The posterior distribution is

$$f(\omega,oldsymbol{eta},oldsymbol{\gamma}|\mathbf{x},\,\mathbf{y}) \propto L(\mathbf{x}|\omega,oldsymbol{eta})L(\mathbf{y}|oldsymbol{\gamma})f(\omega)f(oldsymbol{eta})f(oldsymbol{\gamma})$$

where  $f(\omega)$ ,  $f(\beta)$ , and  $f(\gamma)$  denote the prior distributions for  $\omega$ ,  $\beta$ , and  $\gamma$ , respectively.

• Vague priors are used:

 $\gamma_0 \sim \text{Cauchy}(0,10), \quad \gamma_1, \cdots, \gamma_4 \sim \text{Cauchy}(0, 2.5). \quad \beta_0 \sim \text{Cauchy}(0, 10), \quad \beta_1, \beta_2 \sim \text{Gamma}(0.5, 0.5) \quad \omega \sim N(0, 4)$ 

**Overview** Stage I: Run-in period Stage II: Systematic dose finding

#### Trial design

Our design is conducted in two stages:

- Stage I (run in): We escalate doses along the diagonal to explore the dose-combination space quickly and collect some preliminary data.
- Stage II (dose finding): Based on observed efficacy and toxicity data, we continuously update the posterior estimates of toxicity and posterior means of efficacy and assign patients to the most appropriate dose.

Def: dose  $(a_j, b_k)$  is deemed safe if  $Pr(p_{jk} < \phi | D) > \delta$ ; otherwise toxic.

•  $\phi$  is the target toxicity upper limit and  $\delta$  is a prespecified safety cutoff.

Overview Stage I: Run-in period Stage II: Systematic dose finding

## Stage I: Run-in period

The trial starts with the treatment of the first cohort of patients at the lowest dose  $(a_1, b_1)$ .

- 11 If current dose is safe, escalate the dose along the diagonal. If  $(a_1, b_1)$  is deemed toxic, terminate the trial.
- 12 Stage I completes when either current dose is deemed toxic or the highest dose combination is reached. Stage II starts.

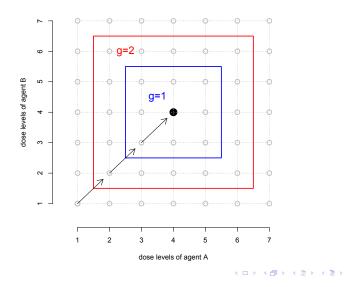
Overview Stage I: Run-in period Stage II: Systematic dose finding

#### g-degree admissible dose set

Assume that the current dose combination is  $(a_j, b_k)$ ,

- Define g-degree neighbors of (a<sub>j</sub>, b<sub>k</sub>), denoted by N<sub>g</sub>, as dose combinations {(a<sub>j'</sub>, b<sub>k'</sub>)} whose dose levels are different from (a<sub>j</sub>, b<sub>k</sub>) no more than g levels, i.e., N<sub>g</sub> = {(a<sub>j'</sub>, b<sub>k'</sub>) : |j' j| ≤ g and |k' k| ≤ g}.
- Further define a g-degree admissible dose set A<sub>g</sub>, which is a subset of the g-degree neighbors N<sub>g</sub> satisfying the pre-specified safety requirement Pr(p<sub>i'k'</sub> < φ<sub>T</sub>|D) > δ.

Overview Stage I: Run-in period Stage II: Systematic dose finding



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Overview Stage I: Run-in period Stage II: Systematic dose finding

## Stage II: Systematic dose finding

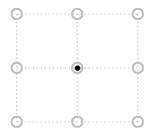
- II1 Based on the observed data, identify  $\mathcal{A}_{g^*}$  as the nonempty set of safe neighbors of  $(a_j, b_k)$  with minimum degree  $g^*$ . If  $\mathcal{A}_{g^*}$ does not exist (i.e., all experimental doses are deemed toxic), terminate the trial.
- II2 Among the doses in  $A_{g^*}$ , identify the dose  $(a_{j^*}, b_{k^*})$  with the highest posterior mean of efficacy  $\hat{q}_{j^*k^*}$ .

Overview Stage I: Run-in period Stage II: Systematic dose finding

First-degree neighbors of current dose combination,  $\mathcal{N}_1$ 

Current dose

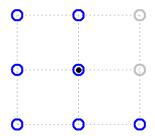
First-degree neighbors



Overview Stage I: Run-in period Stage II: Systematic dose finding

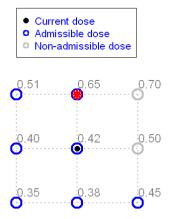
# First-degree admissible dose set of current dose combination, $\mathcal{A}_1$

- Current dose
- Admissible dose
- Non-admissible dose



Overview Stage I: Run-in period Stage II: Systematic dose finding

The dose  $(a_{j^*}, b_{k^*})$  with the highest posterior mean of efficacy  $\hat{q}_{j^*k^*}$ 



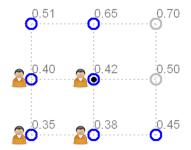
Overview Stage I: Run-in period Stage II: Systematic dose finding

- The commonly used algorithm is to assign the next cohort of patients to  $(a_{j^*}, b_{k^*})$ .
- Problem: this greedy algorithm is easily trapped in locally optimal doses due to
  - small sample size
  - model misspecification
- Solution: a novel dose-finding algorithm to adaptively encourage the exploration of untried doses

Overview Stage I: Run-in period Stage II: Systematic dose finding



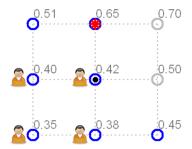
- Admissible dose
- Non-admissible dose



Overview Stage I: Run-in period Stage II: Systematic dose finding

Current dose

- Admissible dose
- Non-admissible dose

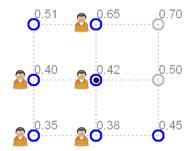


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Overview Stage I: Run-in period Stage II: Systematic dose finding

Current dose

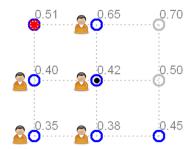
- Admissible dose
- Non-admissible dose



Overview Stage I: Run-in period Stage II: Systematic dose finding

Current dose

- Admissible dose
- Non-admissible dose



Overview Stage I: Run-in period Stage II: Systematic dose finding

### Stage II: Systematic dose finding

- II3 If  $n_{j^*k^*} = 0$  or  $n_{rs} \neq 0$  for all  $(a_r, b_s) \in \mathcal{A}_{g^*}$ , treat the next cohort at dose  $(a_{j^*}, b_{k^*})$ .
  - $\begin{array}{l} \textbf{O} \quad \text{Otherwise,} \\ \left\{ \begin{array}{l} \text{If } \hat{q}_{j^{*}k^{*}} > \left(\frac{N-n}{N}\right)^{\alpha} & \text{treat the next cohort at } \left(a_{j^{*}}, b_{k^{*}}\right), \\ \text{If } \hat{q}_{j^{*}k^{*}} \leq \left(\frac{N-n}{N}\right)^{\alpha} & \text{remove dose } \left(a_{j^{*}}, b_{k^{*}}\right) \text{ from } \mathcal{A}_{g^{*}} \\ & \text{and go to step II2.} \end{array} \right. \end{array} \right.$ 
    - N: prespecified maximum sample size
    - n = ∑<sub>j,k</sub> n<sub>jk</sub>: the total number of patients treated in the trial
      α is a known tuning parameter.
- II4 Repeat steps II2-4 until exhaustion of the sample size. Select as the BODC the dose combination with the highest  $\hat{q}_{jk}$ among all safe doses.

Setup Results

#### Simulation setup

- Consider 4 dose levels for each agent:
  - Dose levels of A and B are (0.075, 0.15, 0.225, 0.3) and (0.08, 0.16, 0.24, 0.32), respectively.
- The maximum sample size was 15 cohorts of size 3.
- Set the target toxicity upper limit  $\phi = 0.3$  and the safety cutoff  $\delta = 0.4$ .
- Set the tuning parameter  $\alpha = 2$ .

Setup Results

#### Simulation setup

- We compared the proposed design with a greedy design that is otherwise identical except that it uses the greedy dose-assignment rule (i.e., always assign the next cohort to the dose with the highest estimate of efficacy).
- 2000 simulated trials under each scenario.

Setup Results

#### Simulation results

#### Table : Scenario 1

		Agent A										
Agent	Tox	icity p	robab	ility	Efficacy probability							
В	1	2	3	4		1	2	3	4			
4	.25	.25	.25	.25		.42	.60	.38	.32			
3	.15	.25	.25	.25		.19	.44	.20	.18			
2	.10	.25	.25	.25		.12	.29	.15	.10			
1	.05	.10	.15	.25		.05	.22	.10	.08			

Setup Results

#### Simulation results

Table : The selection percentage and the percentage of patients treated at each dose combination (shown as the subscripts) for scenario 1.

	-						-						
	Agent A												
		Proposed	design	Greedy design									
В	1	2	3	4	-	1	2	3	4				
4	23.814.1	31.0 <sub>15.9</sub>	$10.8_{9.4}$	8.9 <sub>8.5</sub>		$18.2_{9.5}$	$21.5_{10.0}$	7.8 <sub>5.3</sub>	21.826.5				
3	<b>3.5</b> <sub>3.9</sub>	$5.5_{6.0}$	$1.2_{6.9}$	$1.1_{4.6}$		<b>4.5</b> <sub>3.0</sub>	4.3 <sub>3.0</sub>	$1.1_{9.5}$	2.23.2				
2	0.92.3	2.7 <sub>8.1</sub>	0.83.7	0.52.3		$1.2_{1.6}$	$4.2_{11.4}$	$0.9_{1.6}$	0.61.9				
1	0.77.6	$2.1_{2.8}$	$1.0_{2.1}$	$0.9_{1.8}$		$0.5_{8.4}$	2.2 <sub>1.9</sub>	$1.4_{2.1}$	$2.1_{1.2}$				

Setup Results

#### Simulation results

#### Table : Scenario 2

	Agent A										
Agent	Toxicity probability					Efficacy probability					
В	1	2	3	4		1	2	3	4		
4	.25	.25	.25	.25		.10	.29	.29	.42		
3	.15	.25	.25	.25		.25	.35	.43	.60		
2	.10	.25	.25	.25		.12	.24	.32	.39		
1	.05	.10	.15	.25		.05	.14	.28	.32		

Setup Results

#### Simulation results

Table : The selection percentage and the percentage of patients treated at each dose combination (shown as the subscripts) for scenario 2.

	Agent A										
Agent		Propos	ed desigi	ı		Greedy design					
В	1	2	3	4	1	2	3	4			
4	$1.6_{2.1}$	3.2 <sub>3.2</sub>	$4.1_{6.4}$	17.0 <sub>13.7</sub>	$2.5_{1.6}$	$3.1_{2.3}$	3.9 <sub>3.7</sub>	$30.1_{32.0}$			
3	$2.5_{2.1}$	2.84.3	$7.1_{9.2}$	$33.1_{18.5}$	2.42.3	$3.1_{2.3}$	9.0 <sub>13.9</sub>	17.9 <sub>9.3</sub>			
2	$0.7_{1.6}$	$1.5_{7.8}$	3.4 <sub>5.3</sub>	9.6 <sub>8.5</sub>	0.80.9	$1.1_{9.0}$	3.0 <sub>2.6</sub>	8.2 <sub>5.1</sub>			
1	0.37.3	$0.8_{1.6}$	$2.5_{2.7}$	$6.0_{5.7}$	$0.1_{7.7}$	0.60.9	$2.2_{2.3}$	$7.1_{3.9}$			

Setup Results

#### Simulation results

#### Table : Scenario 3

	Agent A										
Agent	Toxicity probability					Effi	cacy p	robab	ility		
В	1	2	3	4	_	1	2	3	4		
4	.17	.25	.45	.55		.60	.35	.32	.28		
3	.12	.16	.25	.43		.42	.30	.28	.25		
2	.08	.10	.19	.22		.35	.28	.22	.20		
1	.05	.08	.12	.18		.25	.23	.19	.16		

Setup Results

#### Simulation results

Table : The selection percentage and the percentage of patients treated at each dose combination (shown as the subscripts) for scenario 3.

	Agent A											
		Proposed	design			Greedy design						
В	1	2	3	4	1	2	3	4				
4	46.3 <sub>18.9</sub>	6.8 <sub>5.5</sub>	3.4 <sub>5.2</sub>	$1.3_{6.1}$	39.1 <sub>13.8</sub>	$7.1_{5.2}$	3.3 <sub>3.6</sub>	0.99.8				
3	$7.8_{5.5}$	$2.7_{5.0}$	$3.1_{8.6}$	2.24.5	7.3 <sub>3.9</sub>	2.62.9	$3.5_{13.2}$	2.9 <sub>3.9</sub>				
2	$5.3_{5.0}$	$1.9_{8.2}$	$1.5_{4.5}$	$3.1_{3.4}$	3.9 <sub>2.7</sub>	$3.0_{12.0}$	$1.8_{2.5}$	3.9 <sub>3.6</sub>				
1	$5.5_{10.2}$	$2.3_{3.6}$	$1.7_{2.7}$	2.9 <sub>3.0</sub>	8.616.1	$2.5_{2.0}$	$2.5_{1.8}$	4.9 <sub>2.9</sub>				

#### Conclusions

- Our proposed design explicitly accounts for the unique features of the biological agents, i.e., dose-efficacy and -toxicity relationships may take non-monotonic patterns.
- The proposed design adaptively encourages dose exploration in the two-dimensional dose space.
- Our design identifies the BODC with substantially higher selection percentage and allocates more patients to the target dose combination than the greedy design.
- In the case that efficacy plateaus, a similar change-line model can be used.

#### Reference

 Cai, C., Yuan, Y. and Ji, Y. (2014) A Bayesian Phase I/II Design for Oncology Clinical Trials of Combining Biological Agents. *Journal of the Royal Statistical Society: Series C*, 63, 159-173.

## Thank you !