# 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
- Simulation study
- Conclusion

## 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 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.

## **Drug-combination Trials**

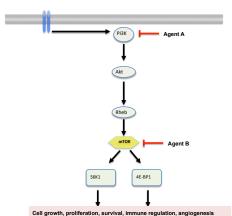
- Treating patients with a combination of agents is becoming common in cancer clinical trials.
- Most existing drug-combination trial designs concern cytotoxic agents (e.g., Thall et al., 2003; Wang and Ivanova, 2005; Yin and Yuan, 2009), thus are not applicable to the trials combining biological agents.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

## 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.

## Motivating trial

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



## 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.

#### 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:  $b_1 < b_2 < \cdots < b_K$
  - $(a_i, b_k)$ : combination of dose  $a_i$  and dose  $b_k$
  - $p_{jk}$  and  $q_{jk}$  denote the toxicity and efficacy probabilities of dose combination  $(a_j, b_k)$
- Goal: identify the BODC in the  $J \times K$  dose matrix.

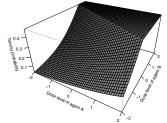
## Change-line model for toxicity

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

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

- $I(\cdot)$ : indicator function
- $\beta_1 > 0$  and  $\beta_2 > 0$  such that  $p_{jk}$  initially increases with the doses of A and B
- When it reaches a plateau, the toxicity probability:  $e^{\omega}/(1+e^{\omega})$ .
- 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



## Logistic model for efficacy

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

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

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

#### Likelihood

Suppose that at a certain stage of the trial

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

$$L(\mathbf{x}|\omega, \boldsymbol{eta}) \propto \prod_{i=1}^J \prod_{k=1}^K \rho_{jk}^{x_{jk}} (1-\rho_{jk})^{n_{jk}-x_{jk}};$$

for the efficacy data y is

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

• The posterior distribution is

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

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

Vague priors are used:

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

#### 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  $\Pr(p_{jk} < \phi | \mathcal{D}) > \delta$ ; otherwise toxic.

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

#### Stage I: Run-in period

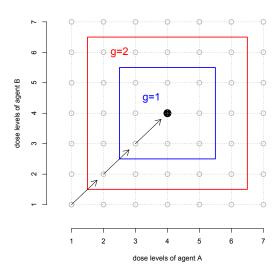
The trial starts with the treatment of the first cohort of patients at the lowest dose  $(a_1, b_1)$ .

- I1 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.

#### g-degree admissible dose set

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

- Define g-degree neighbors of  $(a_j, b_k)$ , denoted by  $\mathcal{N}_g$ , as dose combinations  $\{(a_{j'}, b_{k'})\}$  whose dose levels are different from  $(a_j, b_k)$  no more than g levels, i.e.,  $\mathcal{N}_g = \{(a_{j'}, b_{k'}) : |j' j| \le g \text{ and } |k' k| \le g\}.$
- Further define a g-degree admissible dose set  $\mathcal{A}_g$ , which is a subset of the g-degree neighbors  $\mathcal{N}_g$  satisfying the pre-specified safety requirement  $Pr(p_{j'k'} < \phi_T | \mathcal{D}) > \delta$ .

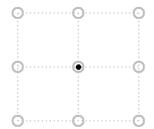


## Stage II: Systematic dose finding

- 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  $\mathcal{A}_{g^*}$ , identify the dose  $(a_{j^*}, b_{k^*})$  with the highest posterior mean of efficacy  $\hat{q}_{j^*k^*}$ .

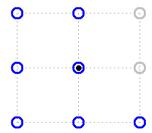
## First-degree neighbors of current dose combination, $\mathcal{N}_1$

- Current dose
- First-degree neighbors



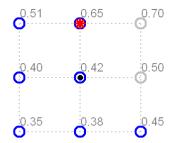
## First-degree admissible dose set of current dose combination, $A_1$

- Current dose
- Admissible dose
- Non-admissible dose



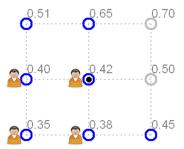
# The dose $(a_{j^*}, b_{k^*})$ with the highest posterior mean of efficacy $\hat{q}_{j^*k^*}$

- Current dose
- Admissible dose
- Non-admissible dose

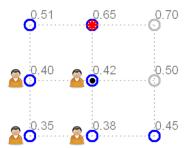


- The commonly used algorithm is to assign the next cohort of patients to  $(a_{j^*}, b_{k^*})$ .
- Problem: this greedy algorithm is easily trapped in locally optimal doses due to
  - small sample size
  - model misspecification
- Solution: a novel dose-finding algorithm to adaptively encourage the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

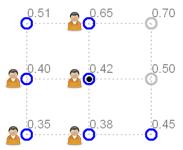
- Current dose
- Admissible dose
- Non-admissible dose



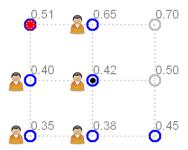
- Current dose
- Admissible dose
- Non-admissible dose



- Current dose
- Admissible dose
- Non-admissible dose



- Current dose
- Admissible dose
- Non-admissible dose



## 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  $(a_{j^*},b_{k^*})$ .
  - Otherwise,  $\begin{cases} &\text{If } \hat{q}_{j^*k^*} > \left(\frac{N-n}{N}\right)^{\alpha} &\text{treat the next cohort at } (a_{j^*},b_{k^*}),\\ &\text{If } \hat{q}_{j^*k^*} \leq \left(\frac{N-n}{N}\right)^{\alpha} &\text{remove dose } (a_{j^*},b_{k^*}) \text{ from } \mathcal{A}_{g^*}\\ &\text{and go to step II2}. \end{cases}$ 
    - N: prespecified maximum sample size
    - $n = \sum_{i,k} n_{jk}$ : the total number of patients treated in the trial
    - $\bullet$   $\alpha$  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.

## 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  $\alpha = 2$ .

## 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.

Table: Scenario 1

	Agent A										
Agent	Tox	cicity 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		

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

	Agent A											
			Greedy	design								
В	1	2	3	4	_	1	2	3	4			
4	23.8 <sub>14.1</sub>	31.0 <sub>15.9</sub>	10.89.4	8.9 <sub>8.5</sub>		18.2 <sub>9.5</sub>	21.5 <sub>10.0</sub>	7.8 <sub>5.3</sub>	21.8 <sub>26.5</sub>			
3	$3.5_{3.9}$	$5.5_{6.0}$	$1.2_{6.9}$	$1.1_{4.6}$		$4.5_{3.0}$	$4.3_{3.0}$	$1.1_{9.5}$	$2.2_{3.2}$			
2	$0.9_{2.3}$	$2.7_{8.1}$	$0.8_{3.7}$	$0.5_{2.3}$		$1.2_{1.6}$	$4.2_{11.4}$	$0.9_{1.6}$	$0.6_{1.9}$			
_1	0.7 <sub>7.6</sub>	$2.1_{2.8}$	$1.0_{2.1}$	$0.9_{1.8}$		$0.5_{8.4}$	2.2 <sub>1.9</sub>	1.4 <sub>2.1</sub>	$2.1_{1.2}$			

Table: Scenario 2

:		Agent A										
Agent	Tox	icity p	robab	_	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			

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 desigr	ı		Greedy design						
В	1	2	3	4		1	2	3	4			
4	1.62.1	3.23.2	$4.1_{6.4}$	17.0 <sub>13.7</sub>		2.5 <sub>1.6</sub>	$3.1_{2.3}$	3.9 <sub>3.7</sub>	30.1 <sub>32.0</sub>			
3	$2.5_{2.1}$	$2.8_{4.3}$	$7.1_{9.2}$	$33.1_{18.5}$		$2.4_{2.3}$	$3.1_{2.3}$	$9.0_{13.9}$	$17.9_{9.3}$			
2	$0.7_{1.6}$	$1.5_{7.8}$	$3.4_{5.3}$	$9.6_{8.5}$		$0.8_{0.9}$	$1.1_{9.0}$	$3.0_{2.6}$	$8.2_{5.1}$			
1	$0.3_{7.3}$	$0.8_{1.6}$	$2.5_{2.7}$	$6.0_{5.7}$		$0.1_{7.7}$	$0.6_{0.9}$	$2.2_{2.3}$	$7.1_{3.9}$			

Table: Scenario 3

-	Agent A										
Agent	Tox	cicity p	robab	ility	Effi	сасу р	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			

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

	Agent A										
	1	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>	1.36.1	39.1 <sub>13.8</sub>	$7.1_{5.2}$	3.333.6	$0.9_{9.8}$			
3	$7.8_{5.5}$	$2.7_{5.0}$	$3.1_{8.6}$	$2.2_{4.5}$	$7.3_{3.9}$	$2.6_{2.9}$	$3.5_{13.2}$	$2.9_{3.9}$			
2	$5.3_{5.0}$	$1.9_{8.2}$	$1.5_{4.5}$	$3.1_{3.4}$	$3.9_{2.7}$	$3.0_{12.0}$	$1.8_{2.5}$	$3.9_{3.6}$			
_1	5.5 <sub>10.2</sub>	$2.3_{3.6}$	$1.7_{2.7}$	2.9 <sub>3.0</sub>	8.6 <sub>16.1</sub>	$2.5_{2.0}$	2.5 <sub>1.8</sub>	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
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## Thank you!