

# Identifying multiple MTD combinations in two-dimensional dose-finding

Nolan A. Wages

University of Virginia

April 15, 2015



Cancer Center

# Outline of Talk

- ▶ Challenges associated with Drug combination studies
- ▶ Method for locating multiple MTD combinations
- ▶ Operating characteristics
- ▶ Conclusions
  - ▶ Extension to other two-dimensional problems

# Drug combinations

## Notation

- ▶ Consider a two-drug combination study
  - ▶ Agent A has  $I$  dose levels:  $A_1 < \dots < A_I$
  - ▶ Agent B has  $J$  dose levels:  $B_1 < \dots < B_J$
- ▶  $(A_j, B_k)$  is the combination of Agent A at dose level  $j$  and Agent B at dose level  $k$
- ▶ Probability of DLT at combination  $(A_i, B_j)$  is denoted  $\pi_{ij}$
- ▶ **Goal:** find maximum tolerated dose combinations (MTDC's)

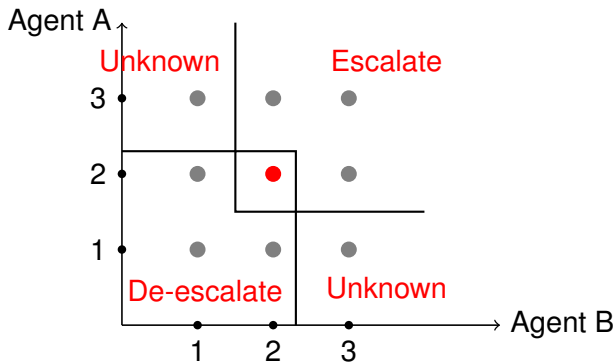
# Drug Combination Studies

## Challenges

1. Toxicity order of the dose combinations is only **partially** known
  - ▶ If current combination is safe, may not be clear where to go next.
2. Dimension of the problem may be large
  - ▶ Many combinations to consider
3. Multiple MTD combinations may exist in the two-dimensional space
  - ▶ MTD equivalence contour

# Drug Combination Studies

## Challenges



# Drug Combination Studies

## Design Considerations

- ▶ Is the objective of the trial to find a...
  1. single MTDC?
  2. MTD equivalence contour containing multiple combinations?

# Recent Methods

## Single MTD Combination

- ▶ Hirakawa, Hamada, Matsui (*Stat Med*, 2013)
- ▶ Braun and Jia (*Stat Biopharm Res*, 2013)
- ▶ Jin, Huo, Yin, Yuan (*Pharm Stat*, 2015)
- ▶ Mander and Sweeting (*Stat Med*, 2015)

# Existing Methods

## Multiple MTD Combinations

- ▶ Thall, Millikan, Mueller, Lee (*Biometrics*, 2003)
- ▶ Ivanova and Wang (*Stat Med*, 2004)
- ▶ Wang and Ivanova (*Biometrics*, 2005)
- ▶ Yuan and Yin (*Stat Med*, 2008)
- ▶ Tighiouart, Piantidosi, Rogatko (*Stat Med*, 2014)



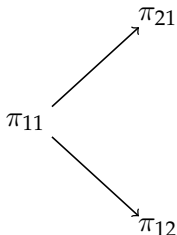
# Review of Harrington *et al.*, 2013

**Table 1** | Summary of features for various dual-agent dose escalation study designs

Study	Number of model parameters	Stages	Outcomes	Response values	Number of RP2D combinations
<i>Model-based designs</i>					
Wang and Ivanova (2005) <sup>60</sup>	3	2	Toxicity	Binary	Minimum number of doses of drug A or drug B
Yin and Yuan (2009) <sup>62</sup>	3	2	Toxicity	Binary	1
Yin and Yuan (2009) <sup>63</sup>	3	2	Toxicity	Binary	1
Kramar <i>et al.</i> (1999) <sup>61</sup>	2	2	Toxicity	Binary	1
Su (2010) <sup>64</sup>	1	3	Toxicity	Binary	1
Thall <i>et al.</i> (2003) <sup>31</sup>	6	2	Toxicity	Binary	3
Conaway <i>et al.</i> (2004) <sup>68</sup>	K	2	Toxicity	Binary	1
Wages <i>et al.</i> (2011) <sup>70</sup>	M <sup>‡</sup>	2	Toxicity	Binary	1
Wages <i>et al.</i> (2011) <sup>69</sup>	M <sup>‡</sup>	1	Toxicity	Binary	1
Braun and Wang (2010) <sup>80</sup>	6	1	Toxicity	Binary	1
Bailey <i>et al.</i> (2009) <sup>67</sup>	≥3 <sup>§</sup>	1	Toxicity	Binary	1

# Partial Order Continual Reassessment Method

Wages, Conaway, O'Quigley (2011)



$$\pi_{11} < \pi_{12} < \pi_{21}$$

or

$$\pi_{11} < \pi_{21} < \pi_{12}$$

- ▶ POCRM has been applied in several ongoing/completed studies\*
- ▶ **Question:** Can POCRM select more than one MTD combination?

\*Wages, Conaway, Slingsluff, Williams, Portell, Hwu, Petroni. *Ann Oncol* 2015; [epub ahead of print].

# Example

## Two agents A and B

- ▶ Phase I trial of two small molecule inhibitors in patients in relapsed/refractory mantle cell lymphoma.
- ▶ Agent A has 2 dose levels:  $A_1 < A_2$
- ▶ Agent B has 4 dose levels:  $B_1 < B_2 < B_3 < B_4$

		Doses of B			
		$B_1$	$B_2$	$B_3$	$B_4$
Doses of A	$A_2$	$(A_2, B_1)$	$(A_2, B_2)$	$(A_2, B_3)$	$(A_2, B_4)$
	$A_1$	$(A_1, B_1)$	$(A_1, B_2)$	$(A_1, B_3)$	$(A_1, B_4)$

# Primary Objective

## Find Multiple MTD's

- ▶ Goal: find an MTDC for each dose of Agent A
- ▶ Locate  $j^* \in \{1, \dots, 4\}$  such that  $(A_i, B_{j^*})$  has DLT probability closest to the target rate  $\phi$  for each  $i$  ( $i = 1, 2$ )
  - ▶ i.e. find an MTD combination in each row  $i$  such that

$$(A_i, B_{j^*}) = \arg \min |\pi_{ij} - \phi|$$

# Combination-toxicity Relationships

## Assumptions

- ▶ Toxicity increases with increasing dose of each agent, holding the other agent fixed
- ▶ DLT probabilities increase up rows and across columns of matrix

		$B_1$	$B_2$	$B_3$	$B_4$
↑	$A_2$	$(A_2, B_1)$	$(A_2, B_2)$	$(A_2, B_3)$	$(A_2, B_4)$
↑	$A_1$	$(A_1, B_1)$	$(A_1, B_2)$	$(A_1, B_3)$	$(A_1, B_4)$
TOXICITY INCREASES		→	→	→	→

# Combination-toxicity Relationships

## Affect on MTD Location

- ▶ Row 1 is at least the most toxic MTDC (i.e. the MTDC in row 1 will contain at least the largest dose of Agent B)
- ▶ For instance, suppose MTDC in row 1 is estimated to be  $(A_1, B_3)$ 
  - ▶ If dose of A is fixed at  $A_2$ , MTD level of drug B must be lower than or equal to 3 (i.e.  $B_1, B_2$ , or  $B_3$ )

$(A_2, B_1)$	$(A_2, B_2)$	$(A_2, B_3)$	$(A_2, B_4)$
$(A_1, B_1)$	$(A_1, B_2)$	$(A_1, B_3)$	$(A_1, B_4)$
		↑ MTD for $A_1$	

# Relative Location of MTD's

## Shift Model\*

- ▶ If MTD for  $A_1$  is  $(A_1, B_{j^*})$ , then MTD for  $A_2$  is  $(A_2, B_{j^* - \Delta_2})$ ;  $\Delta_2 \in \{0, 1, 2, 3\}$
- ▶ The truth could be any one of the four possible values for  $\Delta_2$ 
  - ▶ Use the data to estimate the relative location of the MTD between rows
- ▶ Similar strategy has been used for patient heterogeneity\*\*

\* O'Quigley J, Conaway MR. *Stat Science* 2010; 25: 202–16.

\*\* O'Quigley J, Iasonos A. *Stat Biopharm Res* 2014; 6: 185–197

# Relative Location of MTD's

Shifts of 0 or 1

▶  $\{\Delta_2 = 0\}$

$(A_2, B_1)$	$(A_2, B_2)$	$(A_2, B_3)$	$(A_2, B_4)$
$(A_1, B_1)$	$(A_1, B_2)$	$(A_1, B_3)$	$(A_1, B_4)$

▶  $\{\Delta_2 = 1\}$

$(A_2, B_1)$	$(A_2, B_2)$	$(A_2, B_3)$	$(A_2, B_4)$
$(A_1, B_1)$	$(A_1, B_2)$	$(A_1, B_3)$	$(A_1, B_4)$



# Relative Location of MTD's

Shifts of 2 or 3

▶  $\{\Delta_2 = 2\}$

$(A_2, B_1)$	$(A_2, B_2)$	$(A_2, B_3)$	$(A_2, B_4)$
$(A_1, B_1)$	$(A_1, B_2)$	$(A_1, B_3)$	$(A_1, B_4)$

▶  $\{\Delta_2 = 3\}$

$(A_2, B_1)$	$(A_2, B_2)$	$(A_2, B_3)$	$(A_2, B_4)$
$(A_1, B_1)$	$(A_1, B_2)$	$(A_1, B_3)$	$(A_1, B_4)$

# Working Models

Targeting  $\phi = 0.30$

- ▶ Model  $m = 1 : \{\Delta_2 = 0\}$

$0.06^{\theta_1}$	$0.16^{\theta_1}$	$0.30^{\theta_1}$	$0.45^{\theta_1}$
$0.06^{\theta_1}$	$0.16^{\theta_1}$	$0.30^{\theta_1}$	$0.45^{\theta_1}$

- ▶ Model  $m = 2 : \{\Delta_2 = 1\}$

$0.16^{\theta_2}$	$0.30^{\theta_2}$	$0.45^{\theta_2}$	$0.59^{\theta_2}$
$0.06^{\theta_2}$	$0.16^{\theta_2}$	$0.30^{\theta_2}$	$0.45^{\theta_2}$

Skeleton 1 = `getprior(0.075,0.30,3,4)`

Skeleton 2 = `getprior(0.075,0.30,2,4)`

# Working Models

Targeting  $\phi = 0.30$

- ▶ Model  $m = 3 : \{\Delta_2 = 2\}$

$0.30^{\theta_3}$	$0.45^{\theta_3}$	$0.59^{\theta_3}$	$0.71^{\theta_3}$
$0.06^{\theta_3}$	$0.16^{\theta_3}$	$0.30^{\theta_3}$	$0.45^{\theta_3}$

- ▶ Model  $m = 4 : \{\Delta_2 = 3\}$

$0.30^{\theta_4}$	$0.45^{\theta_4}$	$0.59^{\theta_4}$	$0.71^{\theta_4}$
$0.01^{\theta_4}$	$0.06^{\theta_4}$	$0.16^{\theta_4}$	$0.30^{\theta_4}$

Skeleton 1 = `getprior(0.075,0.30,1,4)`

Skeleton 2 = `getprior(0.075,0.30,4,4)`

# Getting Trial Underway

## Initial Escalation Scheme

- ▶ Choose a path for initial escalation scheme
- ▶ Follow path in the absence of DLT's, until first DLT is observed

	$B_1$	$B_2$	$B_3$	$B_4$
$A_2$	$(A_2, B_1)$	$(A_2, B_2)$	$(A_2, B_3)$	$(A_2, B_4)$
$A_1$	$(A_1, B_1)$	$(A_1, B_2)$	$(A_1, B_3)$	$(A_1, B_4)$

# Modeling Stage

After each cohort inclusion...

1. Use adaptive model selection to choose working model most consistent with data
2. Update estimates of DLT probabilities,  $\hat{\pi}_{ij}$ , for each combination
3. Recommend a combination in each row with  $\hat{\pi}_{ij}$  closest to target rate  $\phi$

$$S = \{(A_1, B_{j_1}^*), (A_2, B_{j_2}^*)\}$$

4. Randomize next patient to treatment in  $S$

## Extension To More Than 2 Rows

- ▶ If MTD for  $A_1$  is  $(A_1, B_{j^*})$ , then MTD for  $A_i$  is  $(A_i, B_{j^* - \Delta_i})$
- ▶ Each model  $m$  consists of multiple  $\Delta$  values  $\{\Delta_2, \Delta_3, \dots, \Delta_I\}$ , where  $\Delta_2 \leq \Delta_3 \leq \dots \leq \Delta_I$
- ▶ For instance,  $m : \{\Delta_2 = 2, \Delta_3 = 3\}$

$(A_3, B_1)$	$(A_3, B_2)$	$(A_3, B_3)$	$(A_3, B_4)$	$(A_3, B_5)$	$(A_3, B_6)$
$(A_2, B_1)$	$(A_2, B_2)$	$(A_2, B_3)$	$(A_2, B_4)$	$(A_2, B_5)$	$(A_2, B_6)$
$(A_1, B_1)$	$(A_1, B_2)$	$(A_1, B_3)$	$(A_1, B_4)$	$(A_1, B_5)$	$(A_1, B_6)$

# Simulation Studies

- ▶ Evaluated performance over four  $3 \times 6$  combination-toxicity scenarios in Wang and Ivanova (2005)
- ▶ Sample size  $N = 60$ ; Target DLT rate;  $\phi = 0.20$
- ▶ Compared to Wang and Ivanova (2005), Yuan and Yin (2008).

# Scenarios in Wang and Ivanova (2005)

	$s_1$	$s_2$	$s_3$	$s_4$	$s_5$	$s_6$
Scenario 1						
Level 3	0.08	0.13	<b>0.20</b>	0.29	0.40	0.53
Level 2	0.05	0.08	0.13	<b>0.20</b>	0.29	0.40
Level 1	0.03	0.05	0.08	0.13	<b>0.20</b>	0.29
Scenario 2						
Level 3	0.05	0.08	0.11	0.15	<b>0.21</b>	0.29
Level 2	0.04	0.06	0.09	0.13	<b>0.18</b>	0.25
Level 1	0.04	0.05	0.08	0.11	0.15	<b>0.21</b>
Scenario 3						
Level 3	<b>0.20</b>	0.30	0.41	0.53	0.65	0.70
Level 2	0.10	<b>0.20</b>	0.25	0.32	0.41	0.50
Level 1	0.03	0.05	0.13	<b>0.20</b>	0.27	0.35
Scenario 4						
Level 3	<b>0.20</b>	0.40	0.47	0.56	0.65	0.76
Level 2	0.08	0.13	<b>0.20</b>	0.32	0.41	0.50
Level 1	0.03	0.05	0.08	0.13	0.17	<b>0.20</b>



# Assessing Performance

## Combined PCS

- ▶ In each row of the matrix, we are interested in how often each dose of Agent B is selected as the MTD
- ▶ Use the combined proportion of correct selection (CPCS) for each row  $i$

Results in Scenario 1 of WI (2005)

0.25	0.23	<b>0.33</b>	0.18	0.02	0.00
0.00	0.04	0.26	<b>0.54</b>	0.14	0.01
0.00	0.03	0.13	0.53	<b>0.27</b>	0.04

$$\text{CPCS} = 0.33 + 0.54 + 0.27 = 1.14$$

# Some Results

Method	Scenario			
	1	2	3	4
Proposed	1.31	1.62	1.73	1.36
WI (2005)	1.14	1.54	1.32	0.99
YY (2008)	1.03	1.25	1.46	1.04

## Concluding Remarks

- ▶ **R** package **pocrm** can be used to implement design
  - ▶ in the process of updating simulation functions
- ▶ Method has good properties in identifying multiple MTDCs
  - ▶ compares favorably with alternative methods in the area (WI, 2005; YY, 2008)
- ▶ Method could be applied to other two-dimensional dose-finding problems
  - ▶ rows could represent different treatment schedules or prognosis groups