

# Dose-escalation trials without monotonicity assumption: A weighted differential entropy approach

Pavel Mozgunov, **Thomas Jaki**

Medical and Pharmaceutical Statistics Research Unit,  
Department of Mathematics and Statistics, Lancaster University, UK

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# Dose escalation

- Limited prior knowledge about toxicities in humans
- Range of  $m$  regimes (doses, combinations, schedules)
- $n$  patients

Goal:

- Find the maximum tolerated regime that corresponds to a controlled level of toxicity, usually  $\gamma \in (0.2, 0.35)$  in oncology trials

# Single agent dose-escalation designs

## Model-based methods

- CRM
- EWOC

## Algorithm based methods

- '3+3' design
- Biased Coin Design

Fundamental assumption: a **monotonic** dose-response relationship

*Cannot be applied to:*

- Combination trials with many treatments
- Scheduling of drugs
- Non-monotonic dose-toxicity relations



## Unknown ordering problem. Example (I)

Let us consider drugs combination dose-escalation trial with

- 3 dose levels of drug  $A$ :  $A_1, A_2, A_3$
- 3 dose levels of drug  $B$ :  $B_1, B_2, B_3$

$(A_1; B_3)$	$(A_2; B_3)$	$(A_3; B_3)$
$(A_1; B_2)$	$(A_2; B_2)$	$(A_3; B_2)$
$(A_1; B_1)$	$(A_2; B_1)$	$(A_3; B_1)$

Even assuming monotonicity one drug being fixed, we cannot order

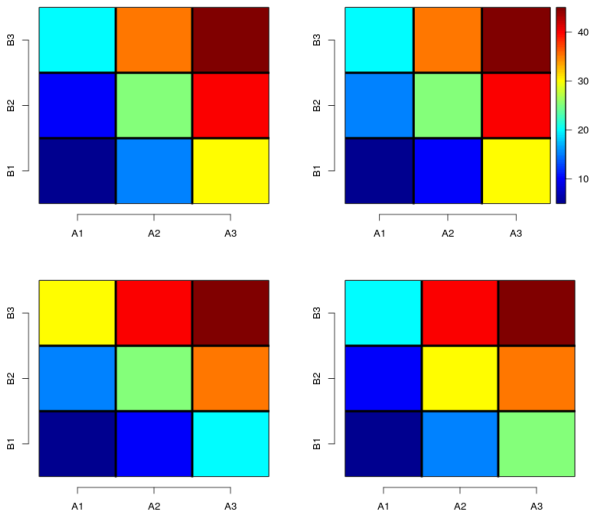
$(A_1; B_2)$  and  $(A_2; B_1)$ ;

$(A_1; B_3)$  and  $(A_2; B_1)$ ;

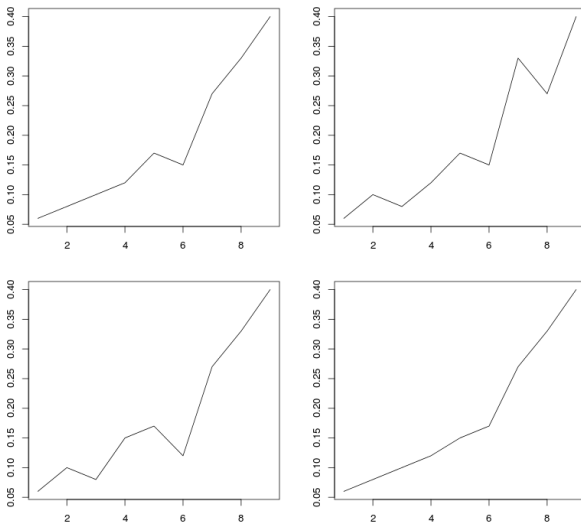
$(A_1; B_3)$  and  $(A_3; B_1)$  and so on...



# Unknown ordering problem. Example (II)



# Unknown ordering problem. Example (III)



# Method for drug combinations

- **Six-parameter model** (*Thall P. et al, 2003*)
- **Up-and-down design** (*Ivanova A, Kim S., 2009*)  
Using the T -statistic
- **Copula regression** (*G.Yin, Y.Yuan, 2009*)  
Parametrization of drug-drug interactive effect
- **POCRM** (*N.Wages, M. Conoway, J. O'Quigley, 2011*)  
Choose several ordering and randomize between them during the trial

General restrictions:

- Strong model assumptions are usually needed
- No diagonal switching is allowed
- Synergistic effect is usually assumed
- Two combinations might be considered only



# Goal

To propose an escalation procedure that **does not require any parametric assumptions** (including monotonicity between regimes).





# Problem formulation

- Toxicity probabilities  $Z_1, \dots, Z_m$  are random variables with Beta prior  $B(\nu_j + 1, \beta_j - \nu_j + 1)$ ,  $\nu_j > 0, \beta_j > 0$
- $n_j$  patients assigned to the regime  $j$  and  $x_j$  toxicities observed
- Beta posterior  $f_{n_j} B(x_j + \nu_j + 1, n_j - x_j + \beta_j - \nu_j + 1)$
- Let  $0 < \alpha_j < 1$  be the unknown parameter in the neighbourhood of which the probability of toxicity is concentrated
- Target toxicity  $\gamma$



# Information theory concepts

## 1) A statistical experiment of estimation of a toxicity probability.

The Shannon differential entropy (DE)  $h(f_n)$  of the PDF  $f_n$  is defined as

$$h(f_n) = - \int_0^1 f_n(p) \log f_n(p) dp \quad (1)$$

with the convention  $0 \log 0 = 0$ .



# Information theory concepts

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## 2) A statistical experiment of a sensitive estimation.

The weighted Shannon differential entropy (WDE),  $h^{\phi_n}(f_n)$ , of the RV  $Z^{(n)}$  with positive weight function  $\phi_n(p) \equiv \phi_n(p, \alpha, \gamma)$  is defined as

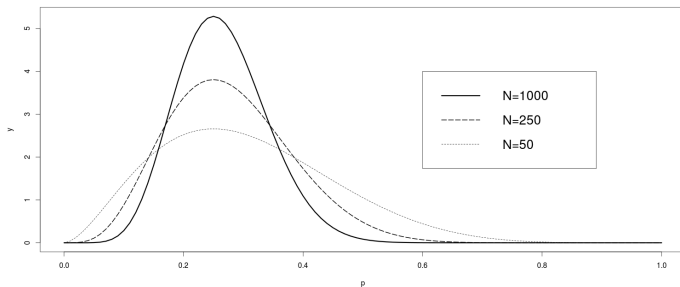
$$h^{\phi_n}(f_n) = - \int_0^1 \phi_n(p) f_n(p) \log f_n(p) dp. \quad (2)$$



# Weight Function

The Beta-form weight function

$$\phi_n(p) = \Lambda(\gamma, x, n) p^{\gamma\sqrt{n}} (1-p)^{(1-\gamma)\sqrt{n}}. \quad (3)$$



# Escalation criteria

The difference of informations in two statistical experiments:

## Theorem

Let  $h(f_n)$  and  $h^{\phi_n}(f_n)$  be the DE and WDE corresponding to PDF  $f_n$  when  $x \sim \alpha n$  with the weight function  $\phi_n$  given in (3). Then

$$\lim_{n \rightarrow \infty} (h^{\phi_n}(f_n) - h(f_n)) = \frac{(\alpha - \gamma)^2}{2\alpha(1 - \alpha)} \equiv \Delta. \quad (4)$$

Therefore, for a regime  $d_j$ ,  $j = 1, \dots, m$ , we obtained that

$$\Delta_j \equiv \frac{(\alpha_j - \gamma)^2}{2\alpha_j(1 - \alpha_j)}.$$

Criteria:

$$\Delta_j = \inf_{i=1, \dots, m} \Delta_i.$$



# Estimation

Consider the mode of the posterior distribution  $f_{n_j}$

$$\hat{p}_j^{(n)} = \frac{x_j + \nu_j}{n_j + \beta_j}.$$

Then the following "plug-in" estimator  $\hat{\Delta}_j^{(n)}$  may be used

$$\hat{\Delta}_j^{(n)} = \frac{(\hat{p}_j^{(n)} - \gamma)^2}{\hat{p}_j^{(n)}(1 - \hat{p}_j^{(n)})}. \quad (5)$$

# Escalation design

Let  $d_j(i)$  be a regime  $d_j$  recommended for cohort  $i$ .

- The procedure starts from  $\hat{\Delta}_j^{(0)}$
- $l$  cohorts were already assigned

The  $(l + 1)^{th}$  cohort of patients will be assigned to regime  $k$  such that

$$d_j(l + 1) : \hat{\Delta}_k^{(l)} = \inf_{i=1, \dots, m} \hat{\Delta}_i^{(l)}, \quad l = 0, 1, 2, \dots, C.$$

We adopt regime  $d_j(C + 1)$  as the final recommended regime.

# Alternative angle

One can consider

$$\hat{\Delta}_j^{(n)} = \frac{(\hat{p}_j^{(n)} - \gamma)^2}{\hat{p}_j^{(n)}(1 - \hat{p}_j^{(n)})}$$

as a **loss function** for a parameter defined on  $(0, 1)$ .

- Loss function penalize  $\hat{p}_j^{(n)}$  close to 0 to 1 and **'pushes' the allocation away from bounds** to the neighbourhood of  $\gamma$
- Does not include any definition of safety  $\rightarrow$  safety constraint is needed



# Safety constrain (I)

Considers regime  $d_j$  as safe if at the moment  $n$  its PDF satisfies

$$\int_{\gamma^*}^1 f_{n_j}(p) dp \leq \theta_n \quad (6)$$

where

- $\gamma^*$  is some threshold after which all regimes above are declared to have excessive risk,  $\gamma^* = \gamma + 0.2$
- $\theta_n$  is the level of probability that controls the overdosing
  - Note that this depends on  $n$

# Why is a time-varying SC is needed?

If  $\beta = 1$  and  $\theta_n = \theta = 0.50$  then regimes with prior mode  $\geq 0.40$  will never be considered since

$$\int_{0.45}^1 f_0(p|x=0)dp = 0.5107 > 0.50$$

Requirements to the function  $\theta_n$

- $\theta_n$  is a decreasing function of  $n$
- $\theta_0 = 1$
- $\theta_N \leq 0.3$
- $\rightarrow \theta_n = 1 - rn$



# Choice of SC parameters

	<i>r</i>							
	0.010	0.015	0.020	0.025	0.030	0.035	0.040	0.045
$\gamma^* = 0.55$	0.00	0.32	4.32	18.47	36.15	49.06	61.49	75.70
	26.47	26.65	26.40	26.05	26.85	25.03	24.10	20.23
$\gamma^* = 0.50$	0.15	2.50	17.76	38.75	52.74	63.06	74.94	87.22
	26.27	26.22	26.53	27.24	25.46	23.30	19.35	17.10
$\gamma^* = 0.45$	1.13	12.72	35.72	56.49	67.16	<b>77.55</b>	86.53	93.49
	26.15	26.02	26.81	25.18	22.26	<b>21.75</b>	15.16	11.05
$\gamma^* = 0.40$	7.47	37.95	59.49	70.52	80.53	88.32	94.18	97.63
	26.04	25.91	24.90	21.98	17.66	14.47	8.05	3.51
$\gamma^* = 0.35$	33.98	58.22	74.42	84.14	90.52	94.86	97.90	99.20
	25.65	24.54	20.45	15.55	13.77	7.21	3.25	0.70
$\gamma^* = 0.30$	55.51	77.02	87.21	92.99	96.50	98.55	99.37	99.83
	24.21	18.09	14.40	11.42	7.13	0.95	0.08	0.04

**Table:** Top row: Proportion of no recommendations for toxic scenario. Bottom row: Proportion of correct recommendations.  $10^6$  simulations.



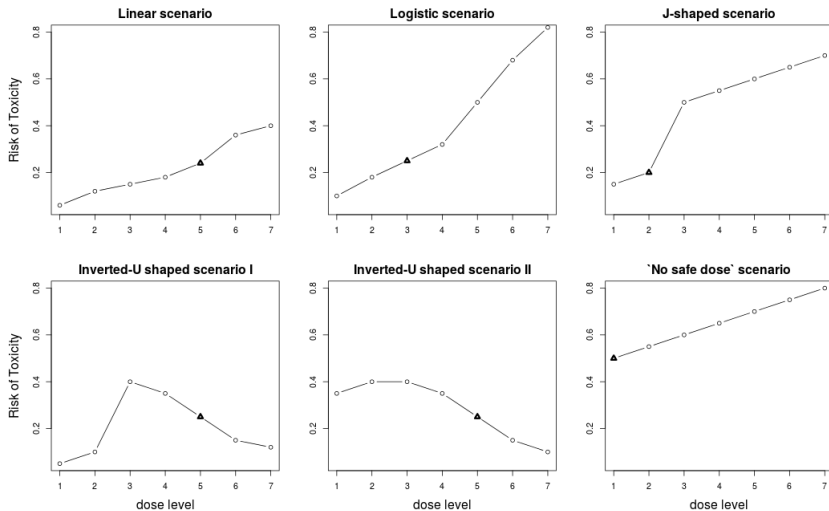
# Simulations

For simulations below the following parameters were chosen:

- The cohort size  $c = 1$
- Total sample size  $N = 20$
- Number of regimes  $m = 7$
- The target probability  $\gamma = 0.25$
- Safety constraint

$$\theta_n = \begin{cases} 1 - 0.035n, & \text{if } 0.035 \times n \leq 0.7; \\ 0.3, & \text{otherwise.} \end{cases}$$

# Investigated scenarios



**Figure:** Considering response shapes. The TD is marked as triangle.

# Specifying the prior

Assumptions:

- Vague beliefs about toxicity risk
- Prior belief: regimes have been correctly ordered monotonically
- A escalation to be started from  $d_1$

The prior for regime  $d_j$  ( $1 \leq j \leq 7$ ) is specified through the mode  $\hat{p}_j^{(0)} = \frac{\nu_j}{\beta_j}$ .

Starting from the bottom:  $\hat{p}_1^{(0)} = \gamma$ .

The vector of modes  $\hat{\mathbf{p}}$  for all regimes is defined

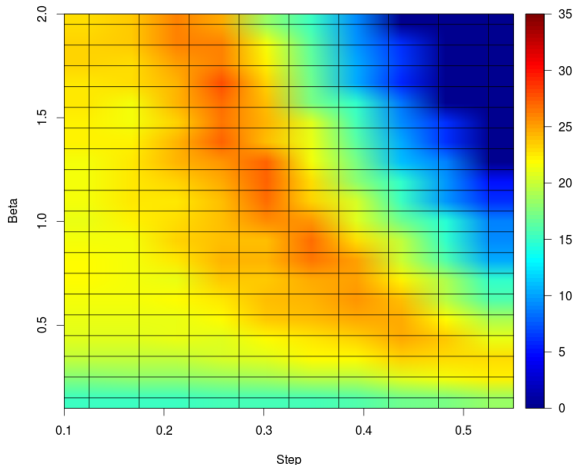
$$\hat{\mathbf{p}} = [0.25, 0.3, 0.35, 0.4, 0.45, 0.5, 0.55]^T.$$

Vague prior  $\rightarrow \beta_j = \beta = 1$  for  $j = 1, \dots, m$ .

Is there a unique set of prior parameters that lead to the equivalent performance?



# Choice of prior



**Figure:** Proportion of correct recommendations:  $\beta$  = number of patients and difference between the risk of toxicity on lowest and highest dose across six scenarios.



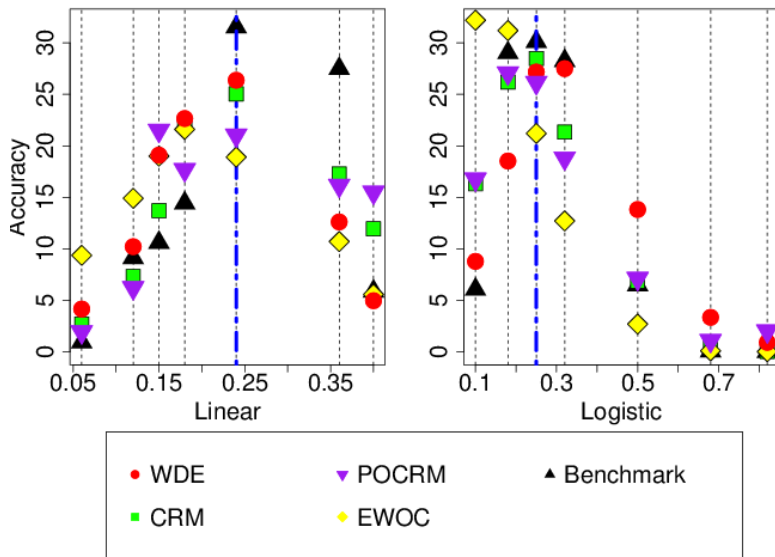
# Alternative methods

We have also investigated

- Continual reassessment method (CRM)
  - Partial ordering continual reassessment method (POCRM)
    - All correct orderings used in simulation are incorporated in the model.*
- Escalation with overdose control (EWOC)
  - A target 25<sup>th</sup> percentile is used.*
- Non-parametric optimal benchmark



# Simulation results. Ordering is correctly specified



# Simulation results. Ordering is wrongly specified.

	$d_1$	$d_2$	$d_3$	$d_4$	$d_5$	$d_6$	$d_7$	No	TR	$\bar{N}$
True	0.05	0.10	0.40	0.35	0.25	0.15	0.12			
WDE <sub>SC</sub>	14.11	19.13	11.77	18.27	27.90	8.50	0.23	0.15	4.26	19.99
CRM <sub>SC</sub>	4.26	19.90	17.70	6.31	2.84	3.00	46.10	0.31	3.26	19.92
POCRM <sub>SC</sub>	2.87	11.39	11.75	9.32	19.11	33.94	11.62	0.24	4.29	19.99
EWOC <sub>SC</sub>	7.18	24.90	18.60	3.79	2.52	3.79	30.60	6.62	2.73	18.89

	$d_1$	$d_2$	$d_3$	$d_4$	$d_5$	$d_6$	$d_7$	No	TR	$\bar{N}$
True	0.35	0.40	0.40	0.35	0.25	0.15	0.10			
WDE <sub>SC</sub>	15.57	12.65	13.31	18.27	27.92	8.90	0.58	9.96	5.81	19.73
CRM <sub>SC</sub>	47.41	2.51	0.97	0.48	0.72	0.40	30.10	27.30	4.27	15.96
POCRM <sub>SC</sub>	16.81	5.98	5.66	12.42	20.10	23.13	10.23	9.67	5.14	19.46
EWOC <sub>SC</sub>	30.75	1.26	0.78	0.47	0.47	0.31	9.78	56.15	3.30	11.02



# Simulation results. Highly toxic scenarios.

	$d_1$	$d_2$	$d_3$	$d_4$	$d_5$	$d_6$	$d_7$	No	TR	$\bar{N}$
True	0.15	0.20	0.50	0.55	0.60	0.65	0.70			
WDE <sub>SC</sub>	38.07	44.65	6.59	3.44	1.48	0.28	0.02	5.47	5.94	19.77
CRM <sub>SC</sub>	37.47	37.85	17.41	2.92	0.36	0.07	0.00	3.92	5.10	19.41
POCRM <sub>SC</sub>	33.57	37.76	13.27	2.55	0.54	1.33	6.04	4.95	6.06	19.82
EWOC <sub>SC</sub>	51.00	26.11	11.01	0.88	0.13	0.00	0.00	10.87	3.60	16.82
True	0.50	0.55	0.60	0.65	0.70	0.75	0.80	No		
WDE <sub>SC</sub>	13.63	5.53	2.45	0.88	0.27	0.06	0.00	77.17	8.02	14.28
CRM <sub>SC</sub>	32.24	0.32	0.08	0.00	0.00	0.00	0.00	67.36	5.33	10.30
POCRM <sub>SC</sub>	15.18	0.57	0.12	0.04	0.01	3.06	0.08	80.94	7.12	12.59
EWOC <sub>SC</sub>	16.17	0.00	0.12	0.00	0.00	0.00	0.00	83.71	3.07	6.05

# Conclusions

The WDE-based method

- **performs comparably** to the model-based methods **when the ordering is specified correctly** scenarios
- **outperform** them in **wrongly specified** setting

However, WDE-based method

- **experience problems** in scenarios with **no safe doses** or with sharp jump in toxicity probability at the bottom.
- **The time-varying safety constrain** in the proposed form *can overcome overdosing problems* and increase the accuracy of the original method



# Further development

- Phase II
- Generalized weight function
- Consistency conditions

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