



# Design for cost minimisation of a phase II clinical study

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# Introduction

- Designs for PKPD studies mainly focus on improving the precision of parameter estimation
  - By optimising dose, dosing regimen and/or sampling schedule
- Upper boundary of the design space → most precise estimates
- A cost penalty has been incorporated in optimal design methods but as a design constraint [1-4]
  - Studies are penalised for number of patients and blood samples but not for study failure
- An empirical value of power is usually chosen *a priori*, often 80%

[1] Mentré M et al. *Biometrika*. 1997;84:429–442.

[2] Retout S et al. *Communication in Statistics*. 2009;8:3351–3368.

[3] Gagnon R et al. *Journal of Biopharmaceutical Statistics*. 2005;15:143–163.

[4] Bazzoli C et al. [www.page-meeting.org/?abstract=1710](http://www.page-meeting.org/?abstract=1710).

# Phase II clinical studies

- Drug tested in target patients for the first time
- Explore dose effect relationship
- Population PK explored in phase I study of healthy volunteers, and then applied to design a phase II study
- PK of healthy volunteers (prior) = PK of patients (target)?

# The balance between cost and failure

- If we don't consider cost then the upper boundary of ethical constraints provides the best design
- Penalising cost reduces precision and increases failure
- Setting power *a priori* is arbitrary, what is the best power?
- What does power mean from a cost perspective?
- Does cost  $\equiv$  power?

# Aims

- To determine if an optimal design exists that
  - Naturally balances the cost of a clinical study with the probability of study success
    - Without arbitrary constraints on the design space
    - Without the need to define power *a priori*
- To determine the influence of different cost structures on the design

# Design variables

$$\xi = \{N_p, N_s, DDD, T_s\}$$

$N_p$  = number of patients

$N_s$  = number of samples per patient

$DDD$  = defined daily dose

$T_s$  = blood sampling times conditioned on  $N_s$

# Expenditure

- For each patient:

Expenditure for samples = sampling days  $\times N_s \times C_s$

Expenditure for drug = study duration (days)  $\times DDD \times C_d$

Expenditure for pre-investigation, housing, food, ... =  $C_p$

- Expenditure of a study:

$$X(\xi) = \$[N_p \times (C_p + \text{Expenditure for samples} + \text{Expenditure for drug})]$$

# Cost of a study

$$\text{Cost} = \begin{cases} X(\xi) & ; \quad \text{study successful} \\ X(\xi) + X(\xi_0) + X(TP) & ; \quad \text{study failed} \end{cases}$$

$X(\xi_0)$  : cost to redo the study using a previous empirical (and more intensive) design

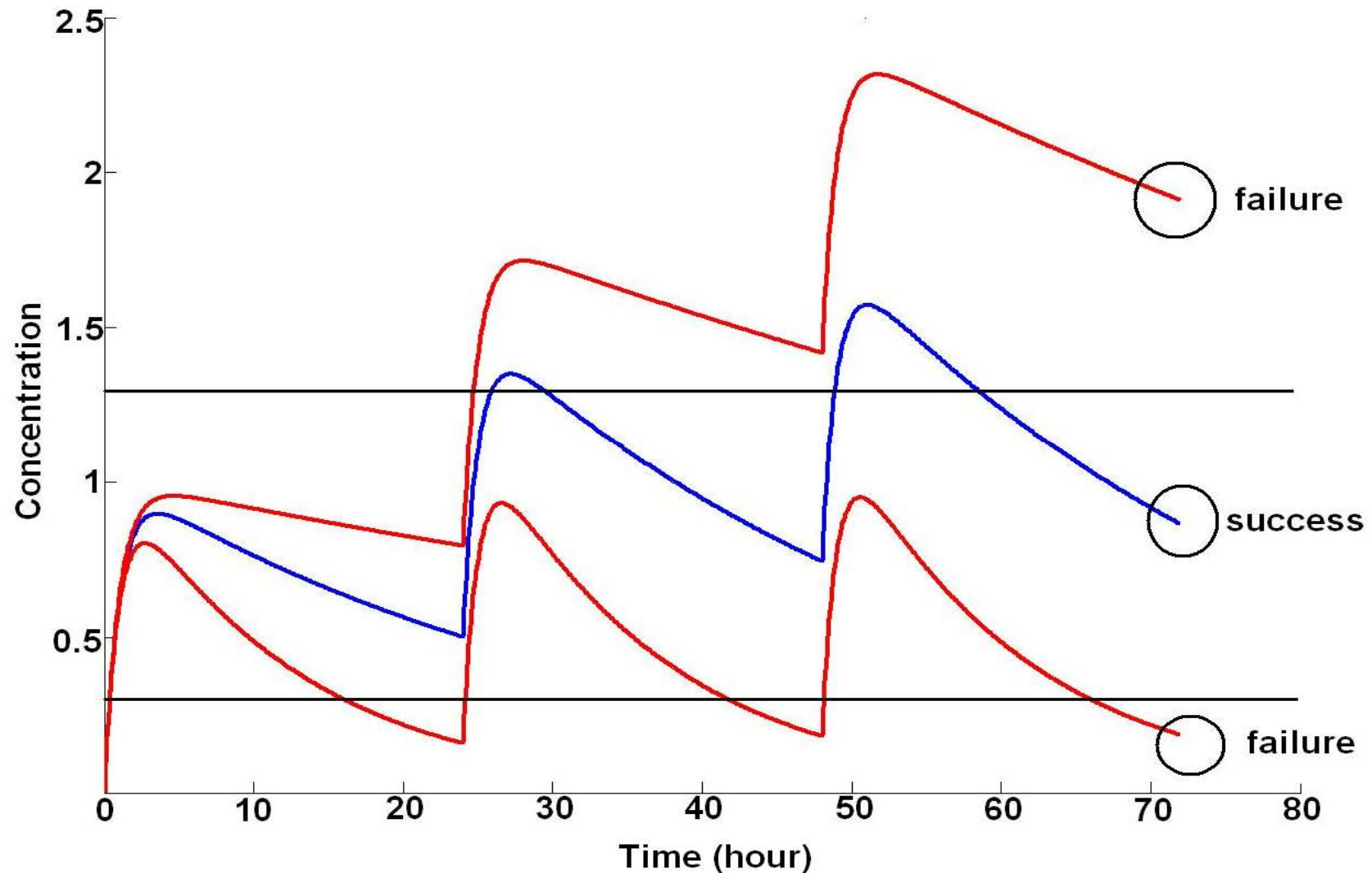
$X(TP)$  : cost for time penalty



# Hypothetical example

- Phase II clinical study for a drug
- All patients received the same dose of drug given orally
- Dosing schedule = 3 doses at 24 hours dose interval
- Therapeutic range of the trough concentration for the 3<sup>rd</sup> dose is defined based on prior biomarker data [0.3 unit/L, 1.3 unit/L]
- The study is successful if  $\geq 60\%$  of patients have trough concentration within the range

# Success criterion for a patient



# The Model

$$C_{ij} = f(Ka_i, CL_i, V_i, t_{ij}, DDD) \times e^{\left(\varepsilon_{p_{ij}}\right)} + \varepsilon_{a_{ij}}$$

$$f(Ka_i, k_i, t_{ij}, DDD) = \frac{DDD \times Ka_i}{V_i \times (Ka_i - k_i)} \times \left( \frac{\left[ e^{(-k_i \times t_{ij})} \right] \left[ 1 - e^{(-dn \times k_i \times di)} \right]}{1 - e^{(-k_i \times di)}} - \frac{\left[ e^{(-Ka_i \times t_{ij})} \right] \left[ 1 - e^{(-dn \times Ka_i \times di)} \right]}{1 - e^{(-Ka_i \times di)}} \right)$$

$$k_i = \frac{CL_i}{V_i} \quad \boldsymbol{\theta}_i = (Ka_i \quad CL_i \quad V_i)^T$$

$$\log(\boldsymbol{\theta}_i) \sim N(\log \bar{\boldsymbol{\theta}}, \boldsymbol{\Omega}) \quad \varepsilon_{p_{ij}} \stackrel{iid}{\sim} N(0, \sigma_p^2) \quad \varepsilon_{a_{ij}} \stackrel{iid}{\sim} N(0, \sigma_a^2)$$

# Describing Uncertainty

- Population PK parameters:  $\Phi_0 = (\bar{\theta}, \Omega, \sigma^2)$

- Hyperprior distribution

$$\bar{\theta} \sim N(\boldsymbol{\mu}, \boldsymbol{\Sigma}) \quad \Omega \sim IW(\mathbf{R}, \nu) \quad \sigma^2 \sim IG(a, b)$$

- Hyperprior parameter:  $\mathbf{H} = \{\boldsymbol{\mu}, \boldsymbol{\Sigma}, \mathbf{R}, \nu, a, b\}$
- If the point estimates and the variance-covariance of the population PK parameters are available, the values of hyperparameters can be computed<sup>[1]</sup>

[1] Dokoumetzidis et al. *Journal of Biopharmaceutical Statistics*. 2008;18:662–676

# Simulation Study

- Population PK estimates from phase I study:

$$\hat{\boldsymbol{\theta}} = (1, 0.03, 1)^T \quad \hat{\boldsymbol{\Omega}} = \begin{bmatrix} 0.1 & 0 & 0 \\ 0 & 0.1 & 0 \\ 0 & 0 & 0.1 \end{bmatrix} \quad \hat{\sigma}_p^2 = 0.1 \quad \hat{\sigma}_a^2 = 0.05$$

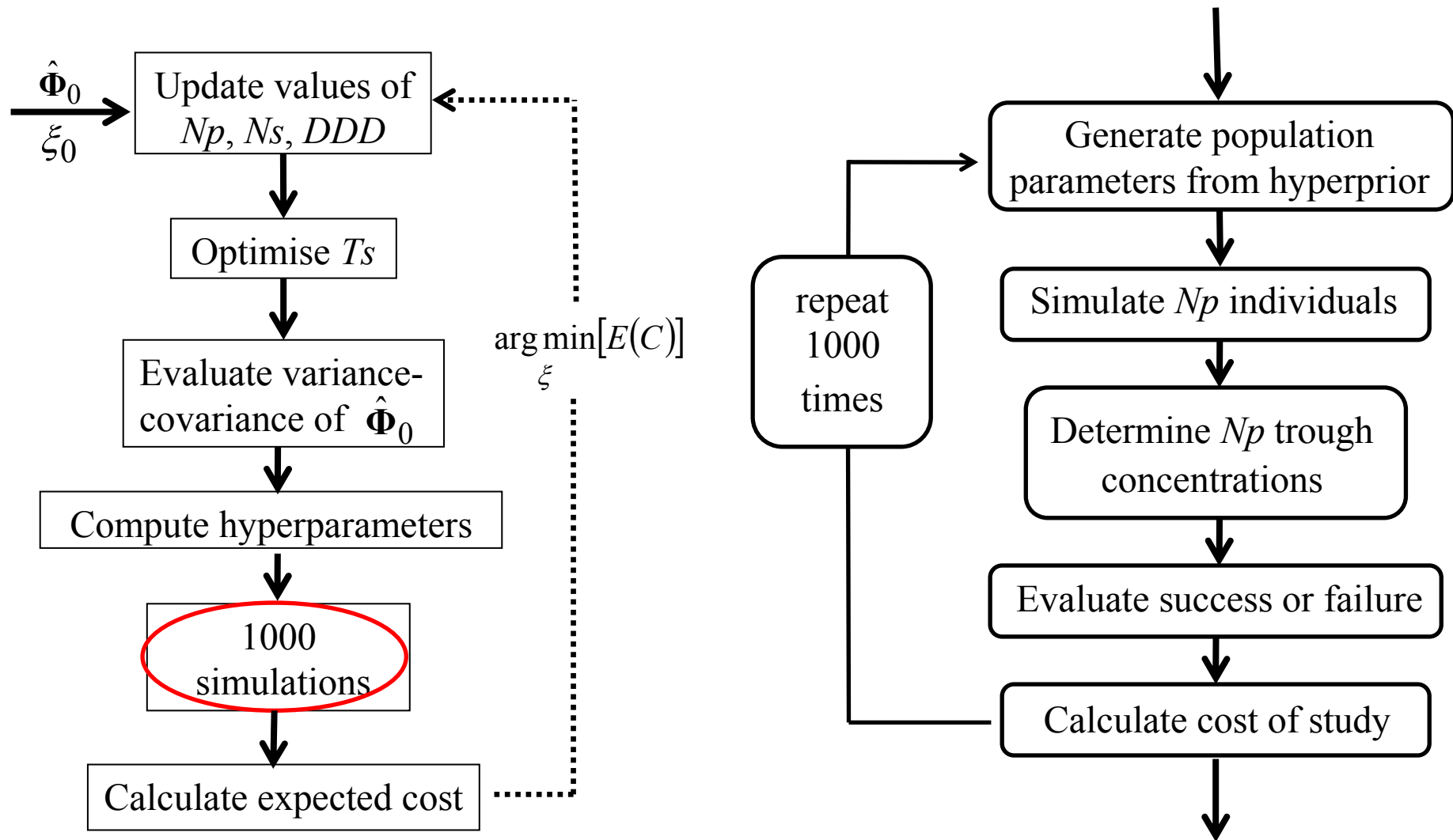
- Hyperprior distribution

$$\bar{\boldsymbol{\theta}} \sim N(\boldsymbol{\mu}, \boldsymbol{\Sigma}) \quad \boldsymbol{\Omega} \sim IW(\mathbf{R}, \nu) \quad \sigma_p^2 \sim IG(a, b) \quad \sigma_a^2 = 0.05$$

# Assumptions

- We consider that ethical constraints and recruitment issues can be handled by penalising the cost per blood sample
- There was one elementary design for the study, which means one sampling schedule for all patients
- A failed study would be repeated with an empirical design

# Procedure



# Simulation Study

	<b>Unit cost</b>	<b>Empirical design</b>	<b>Upper bound</b>
Patient	\$10000	70	100
Blood sample	\$100 \$500 \$1000	8	35
<i>DDD</i>	\$10	1	6



# Result

	<b>Cs</b>	<b>Np</b>	<b>Ns</b>	<b>DDD</b>	<b>\$</b>	<b>Power</b>
No time penalty	100	33	18	3	582,520	0.918
	500	46	8	3	1,185,771	0.890
	1000	58	6	3	1,884,100	0.893
With time penalty	100	38	17	3	618,980	0.968
	500	53	8	3	1,279,500	0.953
	1000	63	6	3	2,012,600	0.932

# Power

- Design for cost minimisation naturally results in study with appropriate power
- High cost  $\neq$  high power & high power  $\neq$  high cost even when the design is optimised
- Setting power *a priori* did not ensure the best design
- Cost minimisation design is a more sensible way to design study

# Conclusions

- There exists an optimal design that naturally balances the cost of a clinical study with the probability of study success
  - Without arbitrary constraints on the design space
  - Without the need to define the power *a priori*
- The design changed with different cost structure

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