

# Utility-based optimization of phase II / phase III clinical development

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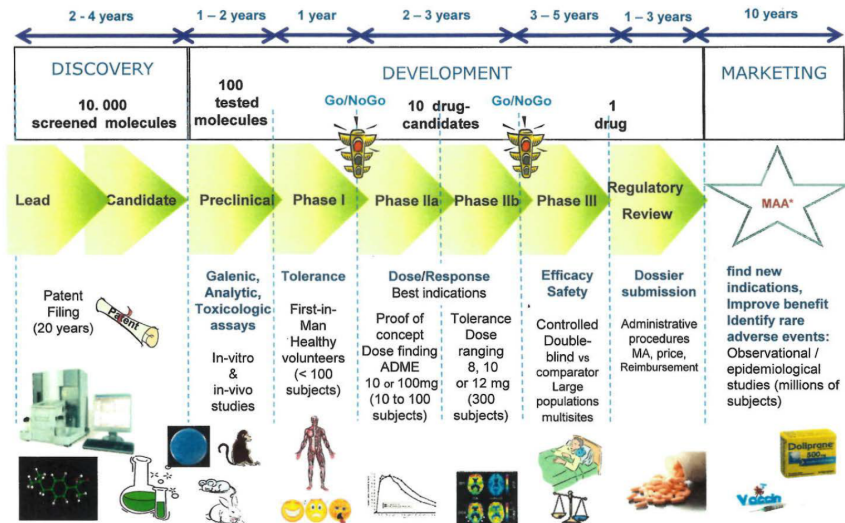
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# Introduction

# Drug Development in Pharmaceutical industries





# Thesis subject

- Dose-finding step is fundamental for phase III in clinical research
- Adaptive designs have already been used in the past and recently in order to select doses
- Optimizing phase II by optimizing the allocation of patients to doses has been poorly explored
  - Several utility functions were proposed and explored through simulations

## Expected goals and objectives of this thesis



- 1 Phase II objectives / Phase II designs
- 2 Optimal Design approach
- 3 Utility Approach

## 1) Phase II Designs / Objectives

- proof of concept study
  - 1 dose, several doses ?
- if several doses
  - which design: balanced ? or not ?
  - which method of analysis

## 2) Optimal Design approach

- D-optimality / C-optimality / Probability of success (POS)
- Multiple doses case and design performance depends of dose response profile, dose of interest
  - Balanced vs optimal design
- Simulations

## 3) Utility approach

- More flexible than "Optimal designs"
  - enables to account for: safety issues, economical/financial aspects, penalties, ...
- Example of utility functions
  - several doses and a placebo

# Mathematical formalization-Notations

- $Y_{d,i} \sim N(m(d; \theta), \sigma^2)$ ,  $i = 1, \dots, n_d$ ,  $n_d$  number of patients per dose group
- Sigmoid-Emax model:  $m(d; \theta) = \frac{\theta_1 \cdot d^{\theta_3}}{\theta_2^{\theta_3} + d^{\theta_3}}$ 
  - $d$  is the dose
  - $m(d; \theta)$  is the true effect for dose  $d$
  - $\theta_1 = E_{max}$  is the maximum effect compared with placebo
  - $\theta_2 = ED_{50}$  is the dose with half of the maximum effect
  - $\theta_3 = g$  (or Hill exponent) is a parameter reflecting the shape of the dose-effect curve
- $\delta$  is the relative effect:  $\delta = \frac{m(d; \theta)}{E_{max}} = \frac{d^g}{d^g + ED_{50}^g}$
- fixed total sample size:  $n_2 + N_3 = N_{total} = constant = 2000$
- $f$  parameter representing patients distribution between phase II and phase III
- $n_2 = f \times N_{total}$  and  $N_3 = (1 - f) \times N_{total}$

## Utility function example

- $U5 = 1(\text{success}) \times (1 - c \times \delta)$
- $U9 = 1(\text{success}) \times (1 - c \times (\frac{d_k}{d_{max}})^2)$  (where  $d_k$  is the dose and  $d_{max}$  is the highest dose)

Sponsor's strategy:

- after phase II:
  - compute  $\mathbb{E}(U(d)|\text{phaseII})$  for each dose  $d$
  - compute  $d_* = \arg \max_d \mathbb{E}(U(d)|\text{phaseII})$
  - decide if worth going into phase III: if  $POS(d_*) \geq 0.30$
- before phase II:
  - choose  $n_2 (= f \times N_{tot})$  sample size of phase II
  - choose the design  $w$

$$(w_*) = \arg \max_{w,f} \mathbb{E}_{w,f}^{(\text{phaseII})} \mathbb{E}(U(d_*)|\text{phaseII})$$

or

$$(f_*) = \arg \max_{w,f} \mathbb{E}_{w,f}^{(\text{phaseII})} \mathbb{E}(U(d_*)|\text{phaseII})$$

# Probability of success

- success means<sup>1</sup>  $\bar{\Delta}(d) \geq 1.96 \times \sqrt{2SE^2}$  (with  $SE^2 = s^2/(N_3/2) = 2s^2/N_3$  and  $\bar{\Delta}(d)$  is the difference between the dose and the placebo after phase III)

- true POS:

$$POS = \mathbb{P}(\bar{\Delta}(d) \geq 1.96 \times \sqrt{2SE^2}) = \Phi \left( \frac{m(d; \theta_0) - 1.96 \times \sqrt{2SE^2}}{\sqrt{2SE^2}} \right)$$

- POS computed by sponsor for dose selection; uses the point estimate:

$$POS = \Phi \left( \left( m(d; \hat{\theta}) - 1.96 \times \sqrt{2SE^2} \right) / \sqrt{2SE^2} \right)$$

# Computation of expectation

$\mathbb{E}(U(d_*)|phasell)$  is a function,  $\mathcal{U}$ , of  $\hat{\theta} \Rightarrow$   
 $\mathbb{E}_{w,f}^{(phasell)} \mathbb{E}(U(d_*)|phasell) = \mathbb{E}(\hat{\theta})\mathcal{U}(\hat{\theta})$  with  $\hat{\theta} \sim N(\theta_0, \mathcal{I}_{\theta_0}^{-1})$

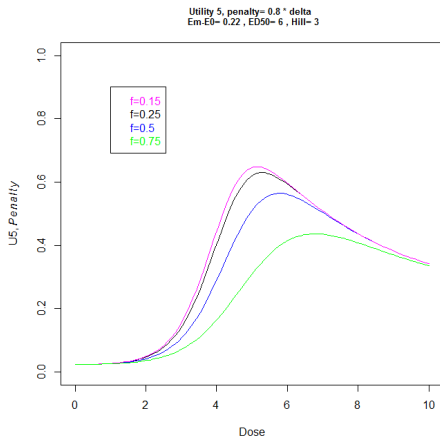
Then  $\mathbb{E}_{w,f}^{(phasell)} \mathbb{E}(U(d_*)|phasell)$  estimated by:

$$\frac{1}{N_{sim}} \sum_{r=1}^{N_{sim}} \mathcal{U}(\hat{\theta}_r)$$

where the  $\hat{\theta}_r$  are sampled from  $N(\theta_0, \mathcal{I}_{\theta_0}^{-1})$



"Theoretical" Utility 5 depends on the size of the phase III:



$$U_9 = 1(\text{success}) \times (1 - c \times (d_k/d_{max})^2)$$

Dose 2

True value : 0.0079

Estimated value → `summary(compile$estd2)`

Min.	1st Qu.	Median	Mean	3rd Qu.	Max.
0.00000	0.00769	0.02652	0.56800	0.12700	165.60000

Dose 4

True value : 0.0503

Estimated value → `summary(compile$estd4)`

Min.	1st Qu.	Median	Mean	3rd Qu.	Max.
0.00000	0.03095	0.06719	0.62000	0.19460	166.00000

Dose 6

True value : 0.11

Estimated value → `summary(compile$estd6)`

Min.	1st Qu.	Median	Mean	3rd Qu.	Max.
0.0000	0.0467	0.1055	0.6585	0.2459	166.2000

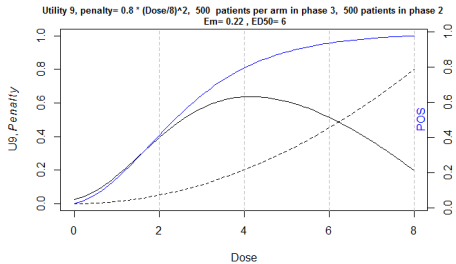
Dose 8

True value : 0.1547

Estimated value → `summary(compile$estd8)`

Min.	1st Qu.	Median	Mean	3rd Qu.	Max.
0.00000	0.05419	0.13660	0.68820	0.29060	166.40000

⇒ Remove the 'Hill' exponent  $g$ :



According to the 15000 simulations with the two parameter model ( $E_{max}$ ,  $ED_{50}$ ):

- Prob(go)=0.77
- Prob(choosing dose 2)=7.4%
- Prob(choosing dose 4)=59.40%
- Prob(choosing dose 6)=27.34%
- Prob(choosing dose 8)=5.82%

d=2 true value -> 0.055

estimate

Min.	1st Qu.	Median	Mean	3rd Qu.	Max.
0.0002888	0.0221700	0.0392300	0.0401700	0.0552400	0.1432000

d=4 true value -> 0.088

estimate

Min.	1st Qu.	Median	Mean	3rd Qu.	Max.
0.0002888	0.0343100	0.0581000	0.0591200	0.0802100	0.2096000

d=6 true value -> 0.11

estimate

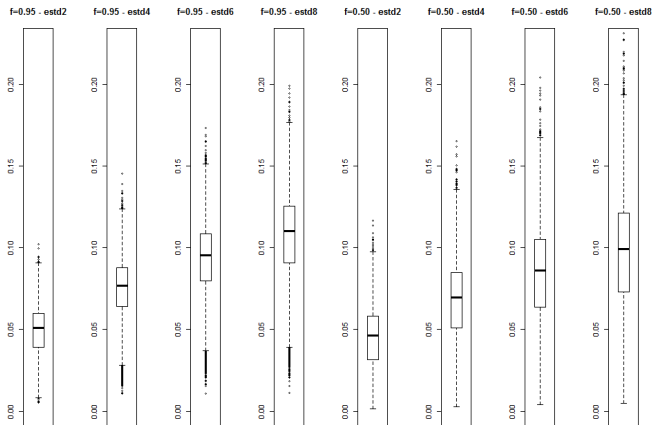
Min.	1st Qu.	Median	Mean	3rd Qu.	Max.
0.0002888	0.0426600	0.0714500	0.0729400	0.0988600	0.2648000

d=8 true value -> 0.12

estimate

Min.	1st Qu.	Median	Mean	3rd Qu.	Max.
0.0002888	0.0486200	0.0823000	0.0840500	0.1140000	0.3049000

- Effects are not very well estimated when "Hill" exponent was removed
- However, 'theta' is fairly well estimated, we have:
  - Theta.true:  $\log(E_m) = -1.514128$  and  $\log(ED_{50}) = 1.791759$
  - Mean estimate:  $\log(E_m) \rightarrow -1.52357$  and  $\log(ED_{50}) \rightarrow 1.780353$



## Non Optimal Designs

Point 2			
Emax	U9_2param	Sigmoid1	Plateau
2 parameters			
		w=(0.2,0.2,0.2,0.2,0.2),f=0.10	w=(0.2,0.2,0.2,0.2,0.2),f=0.10
		Go=55%	Go=69%
		doses=0.16 0.47 0.28 0.09	doses=0.65 0.22 0.09 0.04
		POS(go)=79%	POS(go)=90%
		E(U)=0.2884105	E(U)=0.5211206
		w=(0.2,0.2,0.2,0.2,0.2),f=0.25	w=(0.2,0.2,0.2,0.2,0.2),f=0.25
		Go=78%	Go=88%
		doses=0.08 0.60 0.27 0.05	doses=0.63 0.25 0.09 0.03
		POS(go)=81%	POS(go)=90%
		E(U)= 0.436873	E(U)=0.6665605
		w=(0.2,0.2,0.2,0.2,0.2),f=0.50	w=(0.2,0.2,0.2,0.2,0.2),f=0.50
		Go=91%	Go=96%
		doses=0.04 0.68 0.24 0.03	doses=0.61 0.30 0.08 0.02
		POS(go)=82%	POS(go)=90%
		E(U)=0.5274331	E(U)=0.7395557



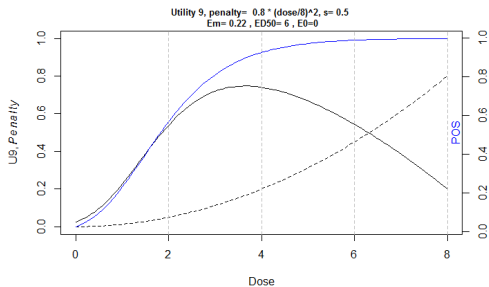
## Optimal Designs

Point 2			
Emax 2 parameters	U9_2param	Sigmoid1	Plateau
		w=(0.00,0.63,0.03,0.00,0.33),f=0.10	w=(0.00,0.46,0.01,0.02,0.51),f=0.10
Go=72%	Go=84%		
doses= 0.13 0.52 0.27 0.07	doses= 0.62 0.25 0.10 0.03		
POS(go)=79%	POS(go)=90%		
E(U)= 0.3874309	E(U)= 0.6358422		
w=(0.00,0.65,0.06,0.00,0.34),f=0.25	w=(0.19,0.21,0.19,0.20,0.21),f=0.25		
Go=91%	Go=88%		
doses= 0.06 0.66 0.24 0.03	doses= 0.62 0.26 0.09 0.03		
POS(go)=81%	POS(go)=90%		
E(U)= 0.526404	E(U)= 0.672623		
w=(0.02,0.68,0.03,0.00,0.27),f=0.50	w=(0.06,0.52,0.01,0.03,0.38),f=0.50		
Go=98%	Go=100%		
doses= 0.02 0.79 0.19 0.01	doses= 0.64 0.32 0.04 0.00		
POS(go)=82%	POS(go)=90%		
E(U)= 0.5921385	E(U)= 0.7839223		

# New Steps

- Global patient allocation between phase II and phase III optimization, rather than dose allocation optimization
- Focus on  $U_9 = 1(\text{success}) \times (1 - c \times (\frac{d_k}{d_{max}})^2)$
- 3-parameter model ( $E_0$ ,  $\log(ED_{50})$  and  $\log(E_m)$ )
- Include a second constraint in the decision rule: the POS must be  $> 0.3$  and the effect difference between placebo and the recommended dose must be  $> 0.04$

## New results



Probability results of selecting each dose, as well as the probability of 'Go':

- $\text{Prob}(\text{Go})=0.7352667$
- $\text{Prob}(\text{choosing dose 2})=32.13347$
- $\text{Prob}(\text{choosing dose 4})=40.39351$
- $\text{Prob}(\text{choosing dose 6})=20.93571$
- $\text{Prob}(\text{choosing dose 8})=6.537311$

Descriptive statistics of the estimates ( $E_0$ ,  $\log(E_m)$  and  $\log(ED_{50})$ ):

```
> summary(U9_simul_complet$X1)  log(Em): true value= -1.514128
```

```
  Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
```

```
-7.6495 -2.4701 -1.5130 -1.5164 -0.5561  3.7237
```

```
> summary(U9_simul_complet$X2)  log(ED50): true value=1.791759
```

```
  Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
```

```
-12.2139 -0.2554  1.8198  1.7958  3.8761 13.3549
```

```
> summary(U9_simul_complet$X3)  E0 : true value=0
```

```
  Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
```

```
-1.858e-01 -3.334e-02 1.242e-04 -5.821e-05 3.365e-02 1.854e-01
```

## Descriptive statistics of the dose effects (dose d versus placebo):

> **summary(U9\_simul\_complet\$estd2) 0.07**

Min. 1st Qu. Median Mean 3rd Qu. Max.  
0.0000912 0.0167214 0.0345835 0.0434328 0.0598453 0.4160290

> **summary(U9\_simul\_complet\$estd4) 0.09**

Min. 1st Qu. Median Mean 3rd Qu. Max.  
0.0001824 0.0263163 0.0509724 0.0619272 0.0839857 0.5185502

> **summary(U9\_simul\_complet\$estd6) 0.11**

Min. 1st Qu. Median Mean 3rd Qu. Max.  
0.0002736 0.0331183 0.0621437 0.0751841 0.1011619 0.5697685

> **summary(U9\_simul\_complet\$estd8) 0.12**

Min. 1st Qu. Median Mean 3rd Qu. Max.  
0.0003648 0.0380519 0.0708128 0.0857639 0.1153829 0.6130136



Non Optimal

	Sigmoid1	Plateau
<b>f=0.10</b>	w = (0.2,0.2,0.2,0.2,0.2 ) Go=51% doses= 0.44, 0.32, 0.17, 0.07 POS(go)=83% POS(global)=42.3% E(U)= 0.315097	w = (0.2,0.2,0.2,0.2,0.2 ) Go=66% doses=0.74, 0.16, 0.07, 0.03 POS(go)=99% POS(global)=65.3% E(U)= 0.568064
<b>f=0.25</b>	w = (0.2,0.2,0.2,0.2,0.2 ) Go=74% doses= 0.32, 0.41, 0.21, 0.06 POS(go)=83% POS(global)=61.4% E(U)= 0.4435682	w = (0.2,0.2,0.2,0.2,0.2 ) Go=86% doses= 0.69, 0.21, 0.08, 0.03 POS(go)=97% POS(global)=83.4% E(U)= 0.7211123
<b>f=0.50</b>	w = (0.2,0.2,0.2,0.2,0.2 ) Go=86% doses= 0.17, 0.52, 0.25, 0.05 POS(go)=77% POS(global)=66.2% E(U)= 0.4625786	w = (0.2,0.2,0.2,0.2,0.2 ) Go=94% doses= 0.61, 0.29, 0.08, 0.02 POS(go)=90% POS(global)=84.6% E(U)= 0.7178017

Optimal

Sigmoid1	Plateau
w = (0.2,0.2,0.2,0.2,0.2 ) , f=0.4008782 Go=83% doses= 0.23, 0.49, 0.23, 0.05 POS(go)=80% POS(global)=66.4 E(U)= 0.4778065	w = (0.2,0.2,0.2,0.2,0.2 ) , f=0.3775344 Go=92% doses=0.65, 0.25, 0.08, 0.02 POS(go)=95% POS(global)=87.4 E(U)=0.745277

# Conclusion and Perspectives

## Main findings:

- Optimizing the dose allocation ratio in stage 2 of the dose-finding study offers very little improvement in regard of significantly increased operational complexity and consequently, this optimization part was removed from the scope of this thesis
- Utility functions depending on the parameters of the dose-response function make things complicated → increase in uncertainty and bad choices after phase II, therefore it is better to reduce the number of parameters in the model
- The sample size of phase II is vital: it is better to have quite enough patients in phase II to make a better choice

## Perspectives:

- Put uncertainty on the penalty of toxicity and rework it properly
- $c$  coefficient mustn't be constant
- $c$  coefficient should become a prior law on the tolerance percentage of a given dose by patients, where tolerance is defined by following the treatment to the end

## **So the idea here is to put efficacy and safety at the same level and avoid arbitrary choices for toxicity:**

- Characterize subject's safety by using a binary safety outcome mimicking the drug limiting toxicity (DLT) concept
- Penalty considered for each subject would depend on the expected probability of DLT / treatment discontinuation at the given dose
- Optimizing the patient allocation ratio between phase II and phase III

**THANK YOU**



**FOR YOUR ATTENTION**

# Appendices

# Mathematical formalization of a Phase II clinical trial I

- $Y_{d,t}$  data representing patients treated with dose  $d$  for different stages, where  $t \in \{1, 2, 3\}$
- $Dose_0$  and  $Dose_t$  the set of selected doses
- $D$  is the decision rule
- $F$  is the data probabilistic model given the parameters
- $\theta_d \rightarrow$  parameter (efficacy of dose  $d$ )
- $Y_{d,t} \sim F(\theta_d, d)$  where  $d \in Dose_t$
- $X_t = \{Y_{d,1}, \dots, Y_{d,t}, Dose_0, \dots, Dose_t\}$  set of observed data and doses
- $D_t : \mathcal{X} \rightarrow w$
- We select the allocation ratio for the next group according to the  $D$  function
- In other words, we have two main steps:
  - Decision step:  $X_t \rightarrow w_t = D_t(X_t)$ , where  $t \in \{1, 2\}$
  - Hazard step:  $X_{t+1} = U_t(X_t, D_t(X_t), Y_{d,t})$ , where  $Y_{d,t} \sim F(\theta_d, d)$

# Examples of utility functions I

- 1  $U1 = -\gamma N_T + 1(\text{success}) \times (R - c(d^g / (ED_{50}^g + d^g) - 0.95))^2$
- 2  $U2 = -\gamma N_T + 1(\text{success}) \times R_{max}(1 - \delta)$
- 3  $U3 = -\gamma N_T + 1(\text{success}) \times R_{max}(1 - \delta)^2$
- 4  $U4 = 1(\text{success}) \times (1 - c(d^g / (ED_{50}^g + d^g) - 0.95))^2$  (without  $-\gamma N_T$  and R)
- 5  $U5 = 1(\text{success}) \times (1 - c \times \delta)$
- 6  $U6 = 1(\text{success}) \times (1 - c \times \delta)^2$  (without  $-\gamma N_T$  and R)
- 7  $U7 = -\gamma N_T + 1(\text{success}) \times (R - \mathbf{1}.c(d^g / (ED_{50}^g + d^g) - 0.95))^2$   
 (where **1.** is an indicator which doesn't allow to take into account the quantity " $c(d^g / (ED_{50}^g + d^g) - 0.95)^2$ " unless we exceed  $ED_{95}$ )



## Examples of utility functions II

- 8  $U8 = 1(\text{success}) \times (1 - \mathbf{1} \cdot c(d^g / (ED_{50}^g + d^g) - 0.95)^2)$  (without  $-\gamma N_T$  and R, where  $\mathbf{1}$  is an indicator which doesn't allow to take into account the quantity " $c(d^g / (ED_{50}^g + d^g) - 0.95)^2$ " unless we exceed  $ED_{95}$ )
- 9  $U9 = 1(\text{success}) \times (1 - c \times (\frac{d_k}{d_{max}})^2)$  (where  $d_k$  is the dose and  $d_{max}$  is the highest dose)

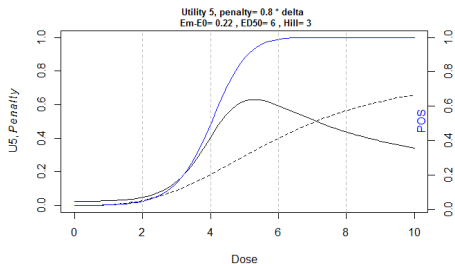
Reminder:  $\delta = d^g / (ED_{50}^g + d^g)$  and the  $\mathbf{1}$  indicator is mathematically translated by  $1(\frac{d^g}{ED_{50}^g + d^g} > 0.95) = 1$  if  $\frac{d^g}{ED_{50}^g + d^g} > 0.95$  and  $= 0$  if

$$\frac{d^g}{ED_{50}^g + d^g} \leq 0.95.$$

## Examples of utility functions III

- Focus on two utility functions:  $U_5$  and  $U_9$
- Computational and optimization problems  $\rightarrow$  move away from Bayesian context
- Analyse phase II with a parametric model ( $E_{max}$  with a parameter  $\theta$ )

$$U_5 = 1(\text{success}) \times (1 - c \times \delta)$$



**Non optimal****sigmoid1**
 $w = (0.2, 0.2, 0.2, 0.2, 0.2), f = 0.25$ 

Go=63%

doses= 0.47, 0.21, 0.19, 0.14

POS(go)=45%

E(U)= 0.1600304

**plateau**
 $w = (0.2, 0.2, 0.2, 0.2, 0.2), f = 0.25$ 

Go=94%

doses= 0.66, 0.21, 0.04, 0.09

POS(go)=90%

E(U)= 0.40004825

## Optimal design

sigmoid1	Plateau
$w = (0.21, 0.21, 0.21, 0.19, 0.18)$ , $f =$	$w = (0.22, 0.18, 0.19, 0.21, 0.20)$ , $f =$
0.25	0.25
Go=:63%	Go=:94%
doses= 0.46, 0.22, 0.18, 0.14	doses= 0.65, 0.21, 0.05, 0.09
POS(go)=45%	POS(go)=90%
E(U)= 0.1638747	E(U)= 0.3996662

- Probability to go to phase III = 63.2%
- Probability of choosing dose 2, if go, = 46.6078916%
- Probability of choosing dose 4, if go, = 20.9224815%
- Probability of choosing dose 6, if go, = 18.9657026%
- Probability of choosing dose 8, if go, = 13.5039243%

**sigmoid1**

$w = (0.2, 0.2, 0.2, 0.2, 0.2)$ ,  $f = 0.25$   
 Go=:63%  
 doses= 0.47, 0.21, 0.19, 0.14  
 POS(go)=45%  
 $E(U) = 0.1600304$

**sigmoid1**

$w = (0.2, 0.2, 0.2, 0.2, 0.2)$ ,  $f = 0.50$   
 Go=:68%  
 doses=0.27, 0.23, 0.31, 0.19  
 POS(go)=57%  
 $E(U) = 0.2202242$

**sigmoid1**

$w = (0.2, 0.2, 0.2, 0.2, 0.2)$ ,  $f = 0.75$   
 Go=:70%  
 doses= 0.12, 0.11, 0.37, 0.41  
 POS(go)=66%  
 $E(U) = 0.2210461$

**sigmoid1**

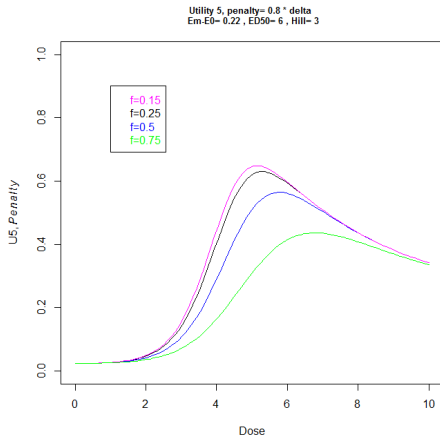
$w = (0.2, 0.2, 0.2, 0.2, 0.2)$ ,  $f = 0.95$   
 Go=:44%  
 doses= 0.01, 0.00, 0.03, 0.96  
 POS(go)=33%  
 $E(U) = 0.06320928$

By increasing the phase II, by taking  $n_2 = 2000$  patients, and keeping  $N_3 = 1500$  patients, we obtain:

- Prob(dose 2) = 17%
- Prob(dose 4) = 36%
- Prob(dose 6) = 41%
- Prob(dose 8) = 9%



"Theoretical" utility depends on the size of the phase III:



With no modeling approach (simulate the stage 1 of phase II, without simulating all the patients, one can simulate a mean and a variance per dose group, for example  $\bar{x} \sim N(\mu_2, \frac{s}{n_2})$ , where  $\mu_2$  is the empirical mean for the dose 2 group,  $s$  is the residual variance, and  $n_2$  is the number of patients for this dose 2 group), we have the following results:

- Prob(go)=0.96
- Prob(choosing dose 2)=38.6%
- Prob(choosing dose 4)=38.5%
- Prob(choosing dose 6)=18.3%
- Prob(choosing dose 8)=4.7%

## Simulated means are consistent with theoretical means:

```
Placebo > summary(U9_simul_complet$X1)
  Min.   1st Qu.   Median     Mean   3rd Qu.    Max.
-0.1829000 -0.0343400 -0.0007815 -0.0008171  0.0328200  0.2141000
```

```
Dose 2 > summary(U9_simul_complet$X2)
  Min. 1st Qu.  Median     Mean 3rd Qu.    Max.
-0.13830  0.01995  0.05359  0.05351  0.08699  0.25770
```

```
Dose 4 > summary(U9_simul_complet$X3)
  Min. 1st Qu.  Median     Mean 3rd Qu.    Max.
-0.09561  0.05504  0.08896  0.08892  0.12350  0.28900
```

```
Dose 6 > summary(U9_simul_complet$X4)
  Min. 1st Qu.  Median     Mean 3rd Qu.    Max.
-0.11750  0.07646  0.11000  0.10970  0.14320  0.29400
```

```
Dose 8 > summary(U9_simul_complet$X5)
  Min. 1st Qu.  Median     Mean 3rd Qu.    Max.
-0.08588  0.09172  0.12570  0.12560  0.15900  0.33680
```

Point 3			
No Model	U9_non_param	Sigmoid1	Plateau
		w=(0.2,0.2,0.2,0.2,0.2,0.2),f=0.10	w=(0.2,0.2,0.2,0.2,0.2,0.2),f=0.10
		Go=92%	Go=95%
		doses=0.48 0.30 0.16 0.06	doses=0.61 0.26 0.10 0.04
		POS(go)=65%	POS(go)=90%
		E(U)=0.4361716	E(U)=0.7124915
		w=(0.2,0.2,0.2,0.2,0.2,0.2),f=0.25	w=(0.2,0.2,0.2,0.2,0.2,0.2),f=0.25
		Go=96%	Go=98%
		doses=0.39 0.37 0.18 0.05	doses=0.61 0.29 0.09 0.02
		POS(go)=68%	POS(go)=90%
		E(U)=0.4759959	E(U)= 0.7486909
		w=(0.2,0.2,0.2,0.2,0.2,0.2),f=0.50	w=(0.2,0.2,0.2,0.2,0.2,0.2),f=0.50
		Go=99%	Go=99%
		doses=0.33 0.44 0.20 0.03	doses=0.63 0.30 0.06 0.01
		POS(go)=70%	POS(go)=90%
		E(U)=0.5090461	E(U)=0.7748988

## Point 3

No Model	U9_non_param	Sigmoid1	Plateau
		w=(0.59,0.08,0.09,0.13,0.11),f=0.10	w=(0.23,0.19,0.20,0.19,0.19),f=0.10
		Go=97%	Go=95%
		doses= 0.46 0.33 0.16 0.05	doses= 0.61 0.26 0.10 0.03
		POS(go)=65%	POS(go)=90%
		E(U)= 0.4713943	E(U)= 0.7180307
		w=(0.48,0.12,0.14,0.13,0.13),f=0.25	w=(0.2,0.2,0.2,0.2,0.2),f=0.25
		Go=99%	Go=99%
		doses= 0.39 0.40 0.18 0.03	doses= 0.61 0.29 0.09 0.02
		POS(go)=68%	POS(go)=90%
		E(U)= 0.4980988	E(U)= 0.7566064
		w=(0.23,0.19,0.19,0.18,0.20),f=0.50	w=(0.55,0.25,0.10,0.07,0.03),f=0.50
		Go=99%	Go=100%
		doses= 0.33 0.44 0.20 0.03	doses= 0.65 0.30 0.05 0.00
		POS(go)=70%	POS(go)=89%
		E(U)= 0.5139042	E(U)= 0.7858431

By considering  $w = c(0.06, 0.52, 0.01, 0.03, 0.38)$  as a starting value for the algorithm:

**Plateau**

$w=(0.17,0.28,0.016,0.015,0.25),f=0.25$

$Go=91\%$

doses= 0.62 0.27 0.09 0.02

$POS(go)=90\%$

$E(U)= 0.6961431$

## Utility definitions

- With or without economic consideration (i.e. costs, phase II & phase III, potential gains if success)
- This type of utility,  $U = 1 \times Penalty$ , if successful, 0, otherwise, was privileged:
  - The dose that maximizes the average utility is the one that maximizes  $\mathbb{E}1(success) \times Penalty(dose, \theta) = POS(dose, \theta, N_3) \times Penalty(dose, \theta)$
  - The sponsor chooses, after phase II, the dose with the average utility estimation:  $POS(dose, \hat{\theta}, N_3) \times Penalty(dose, \hat{\theta})$
- It is better to avoid a penalty depending on the model (on  $\theta$ )
  - Moreover, if the sponsor has a constant total number, as soon as the molecule is a little efficient, the sponsor has an interest in putting as many patients as possible in phase III so that the low dose has a POS close to 1 (even if the effect is very small), and since the penalty is minimal for the low dose  $\Rightarrow$  the low dose will be optimal