



#### **Bayesian Methods for Clinical Trials**

Lecture 1: Clinical trials and the role of statistics

Libby Daniells & Pavel Mozgunov & Thomas Jaki MRC Biostatistics Unit February 24, 2025

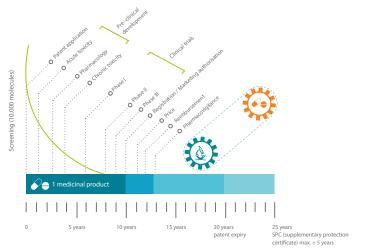
## A dose of reality

Ben Goldacre, Guardian 1-09-08:

- Before 1935 doctors were basically useless
- 1935-1995: antibiotics, dialysis, transplants, intensive-care units, heart surgery, every drug you've ever heard of
- 1995-now: the low-hanging fruit of medical research has all been harvested, and the industry is rapidly running out of new drugs

#### The development process

#### PHASES OF THE RESEARCH AND DEVELOPMENT PROCESS



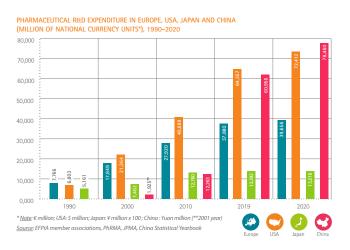


## Drug development

#### Development of a novel medicinal product

- takes 10-15 years
- costs several hundred million euros on average
  - largest contributors are confirmatory (phase III) trials
  - often involve thousands of patients with follow-up period frequently lasting years

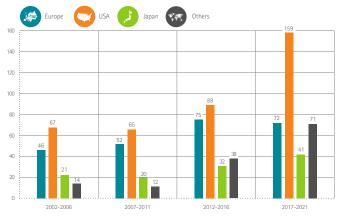
## Cost on R&D in Pharmaceutical industry





#### New molecular entities

#### NUMBER OF NEW CHEMICAL AND BIOLOGICAL ENTITIES (2002-2021)

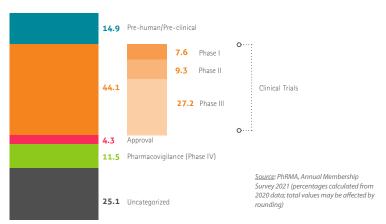


Source: SCRIP – EFPIA calculations (according to nationality of mother company)



## Cost on R & D in Pharmaceutical industry

#### ALLOCATION OF R&D INVESTMENTS BY FUNCTION (%)





#### Success rates

According to a recent review (Wong, Siah & Lo, Biostatistics, 2019), between 2000 and 2015

- 41.0% of confirmatory clinical trials overall and
- 64.5% of confirmatory clinical trials in oncology

have been unsuccessful.

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have been unsuccessful.

- 13.4% of treatments entering Phase I receive approval
- In Oncology only 3.4% of treatments entering Phase I receive approval

#### Consequences

- Avoid going straight into large and expensive phase III trials
- Take more care during phases I and II
- Explore the potential of "new" statistical methods:
  - Sequential designs
  - Adaptive designs
  - Bayesian methods



# Fixed Sample Designs



- no learning during conduct of trial
- clear separation between development phases



# A general approach for fixed sample designs

- Parallel group comparison of efficacy
- E: Experimental, C: Control
- $n = n_E + n_C$
- Superiority trial (advantage for E sought)

 $H_0$ : E no different from C vs  $H_1$ : E superior to C

# A general approach for fixed sample designs

 $\theta$ : measure of advantage of E over C

 $\theta >$  0: E superior  $\theta =$  0: No difference  $\theta <$  0: E inferior

 $\theta_R$ : target measure of advantage of E over C

(target effect)

 $\alpha$ :  $\mathbb{P}\{\text{Reject } H_0 | \theta = 0\}$ Type I error rate (one-sided)

1 –  $\beta$ :  $\mathbb{P}\{\text{Reject } H_0 | \theta = \theta_R\}$ Power to detect the target effect

#### Score Test

- $x_1, x_2, \ldots$  each with density  $f(x|\theta)$
- Log-likelihood of  $\theta$  is

$$I(\theta|x_1,\ldots,x_n) = \sum_{i=1}^n \log f(x_i|\theta)$$

Expanding a Taylor's expansion gives

$$I(\theta) = I(0) + \theta I'(0) + \frac{1}{2}\theta^2 I''(0) + \dots$$
  
 $\approx const + \theta B - \frac{1}{2}\theta^2 V \text{ (for small } \theta)$ 

where B = l'(0) and V = -l''(0).

#### Score Test

- As *B* is a sum  $-\sum_{i=1}^{n} \frac{d}{d\theta} \log f(x_i|\theta)$  it is asymptotically normally distributed.
- For large n and small  $\theta$

$$B \sim \textit{N}(\theta \textit{V}, \textit{V})$$
 and  $\frac{B}{\sqrt{\textit{V}}} \sim \textit{N}(0, 1)$ 

- B is the efficient score statistic for  $\theta$
- For various response types B and V can be derived

# Example: Normal data

treatment	Е	С
mean	$\mu_{E}$	$\mu_{\mathcal{C}}$
standard deviation (known)	$\sigma$	$\sigma$

- $\theta = \mu_E \mu_C$
- $B = \frac{n_C n_E}{n\sigma^2} (\bar{X}_E \bar{X}_C)$
- $V = \frac{n_C n_E}{n\sigma^2} = \frac{n}{4\sigma^2}$  for  $n_E = n_C$

## Example: Normal data

$$Z = \frac{B}{\sqrt{V}} = \frac{\bar{X}_E - \bar{X}_C}{\sigma} \sqrt{\frac{n_C n_E}{n}}$$
$$= \frac{\bar{X}_E - \bar{X}_C}{\sigma \sqrt{\frac{1}{n_E} + \frac{1}{n_C}}}$$

And hence the Score test is equivalent to the Z-test in this situation.

# Sample size

$$\mathbb{P}\{B > c \mid \theta = 0\} = \alpha$$

$$\mathbb{P}\left\{\frac{B}{\sqrt{V}} > \frac{c}{\sqrt{V}} \mid \theta = 0\right\} = \alpha$$

$$1 - \Phi\left(\frac{c}{\sqrt{V}}\right) = \alpha \quad \text{as } \frac{B}{\sqrt{V}} \sim N(0, 1)$$

so that  $c = z_{1-\alpha}\sqrt{V}$ , where  $z_{\gamma}$  is the  $\gamma$  percentile of the standard normal distribution.

# Sample size

Similarly

$$\mathbb{P}\{B > c \mid \theta = \theta_R\} = 1 - \beta$$
$$1 - \Phi\left(\frac{c - \theta_R V}{\sqrt{V}}\right) = 1 - \beta$$

as for 
$$\theta = \theta_R$$
,  $\frac{Z - \theta_R V}{\sqrt{V}} \sim N(0, 1)$ .

Therefore

$$c - \theta_R V = z_\beta \sqrt{V}$$

$$V = \left(\frac{z_{1-\alpha} + z_{1-\beta}}{\theta_R}\right)^2$$

when combining the results.



## Sample Size

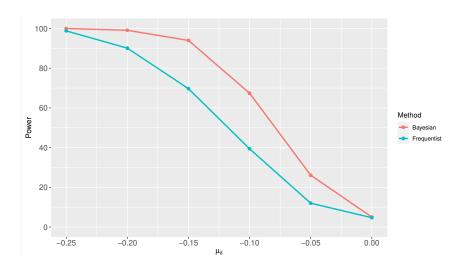
$$V = \left(\frac{z_{1-\alpha} + z_{1-\beta}}{\theta_R}\right)^2$$

- This formula has general validity
- V is a function of n and unknown parameters, which have to be estimated to obtain a sample size
- The relationship of V to n is the most approximate part of the procedure

# When do traditional trials struggle: Rare diseases

- Neuroferritinopathy is an ultra-rare disease
- Fewer than 100 cases reported since its identification in 2001
- Some observational data available

# When do traditional trials struggle: Rare diseases

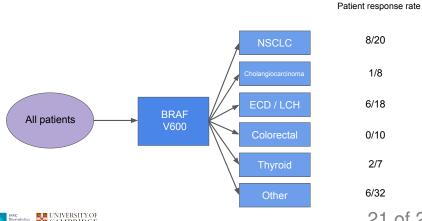




#### Basket trials in oncology – An example

Hyman et al. (2015) reported a recent basket trial, which has been designed to evaluate the efficacy of vemurafenib in BRAF-V600.

A total of 122 patients with BRAF-V600 mutations were enrolled, of which 95 entered the 6 modules





#### References







