

# Sample Size Estimation for Bioequivalence Studies

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# $\alpha$ and $\beta$

- All formal decisions are subjected to two ‘types’ of error:

- $\alpha$ : Probability of Type I Error (aka Risk Type I)
- $\beta$ : Probability of Type II Error (aka Risk Type II)

Example from the justice system – which presumes that the defendant is *not* guilty:

Verdict	Defendant innocent	Defendant guilty
Presumption of innocence rejected <b>guilty</b>	Error type I	Correct
Presumption of innocence accepted <b>not guilty</b>	Correct	Error type II

# $\alpha$ and $\beta$

- In statistical terminology

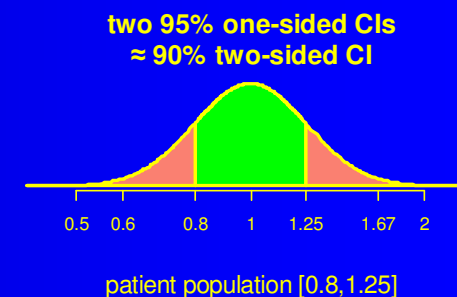
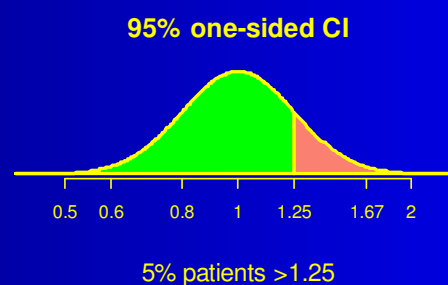
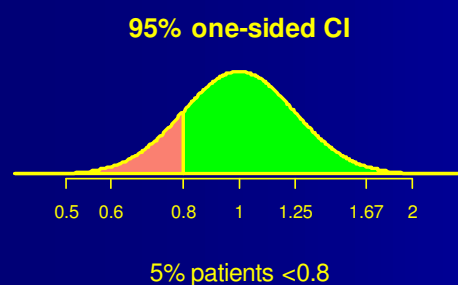
Decision	Null hypothesis true	Null hypothesis false
Null hypothesis rejected	Error type I	Correct ( $H_a$ )
Failed to reject null hypothesis	Correct ( $H_0$ )	Error type II

- In BE-testing the Null hypothesis is **bioinequivalence** ( $\mu_1 \neq \mu_2$ )!

Decision	Null hypothesis true	Null hypothesis false
Null hypothesis rejected	Patient's risk	Correct (BE)
Failed to reject null hypothesis	Correct (not BE)	Producer's risk

$\alpha \dots$ 

- Patient's Risk to be treated with an **inequivalent** formulation ( $H_0$  falsely **rejected**)
  - BA of the test compared to reference in a *particular* patient is risky **either** below 80% **or** above 125%.
  - If we keep the risk of *particular patients* at  $\alpha$  0.05 (5%), the risk of the entire *population of patients* (<80% **and** >125%) is  $2\alpha$  (10%) – expressed as a confidence interval:  $100(1 - 2\alpha) = 90\%$ .



... and  $\beta$ 

- **Producer's Risk** to get no approval of an **equivalent** formulation ( $H_0$  falsely **not rejected**)
  - *Fixed* in study planning to  $\leq 0.2$  (20%), where  $\text{power} = 1 - \beta = \geq 80\%$
  - If power is set to 80%, **one** out of **five** studies will fail just by chance!

$\alpha$ 0.05	BE
not BE	$\beta$ 0.20

←  $0.20 = 1/5$

- *A posteriori* (*post hoc*) power does not make sense!  
**Either** a study has demonstrated BE **or** not.

# Sample Size (Guidelines)

## ● Minimum

- 12 WHO, EU, CAN, NZ, AUS, AR, MZ, ASEAN States, RSA, Russia (“Red Book”), **Ukraine**
- 12 USA ‘A pilot study that documents BE can be appropriate, provided its design and execution are suitable and a sufficient number of subjects (e.g., 12) have completed the study.’
- 18 Russia (2008)
- 20 RSA (MR formulations)
- 24 Saudia Arabia (12 to 24 if statistically justifiable)
- 24 Brazil, USA (in replicate designs intended for RSABE)
- ‘Sufficient number’ Japan

# Sample Size (Guidelines)

- **Maximum**

- **Generally not specified (decided by IEC/IRB and/or local Authorities).**

- **ICH E9, Section 3.5 states:**

*‘The number of subjects in a clinical trial should always be large enough to provide a reliable answer to the questions addressed.’*

# No Sample Size 'Calculation'

- It is not possible to *directly* calculate the required sample size.
- Power is calculated instead; the smallest sample size which fulfills the minimum target power is used.
  - Example:  
 $\alpha$  0.05, target power 80% ( $\beta$  0.2),  
expected T/R 0.95,  $CV_{\text{intra}}$  30%  $\rightarrow$   
minimum sample size 39 (power 80.6%),  
rounded *up* to the next even number in  
a  $2 \times 2 \times 2$  study (power 81.6%).

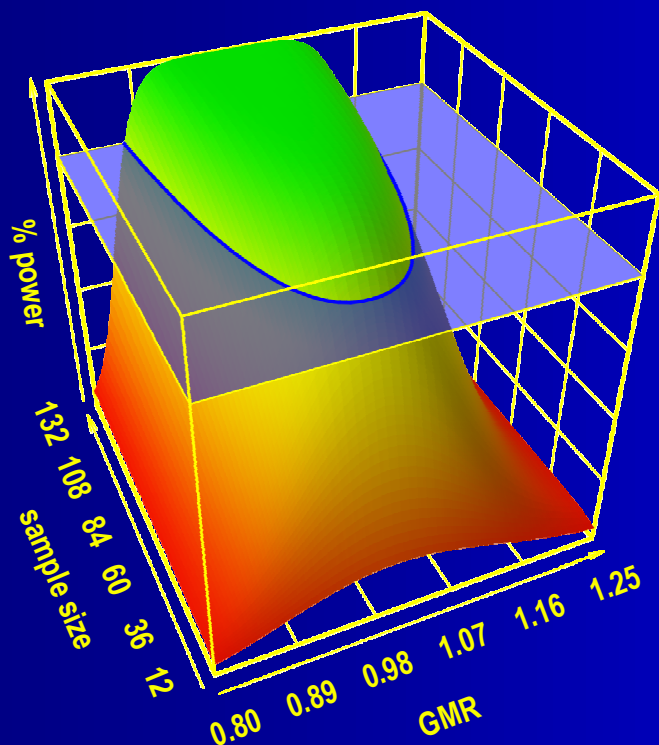
n	Power (%)
36	77.24
37	78.39
38	79.53
39	80.56
40	81.58



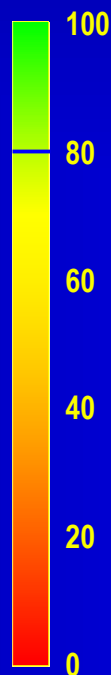
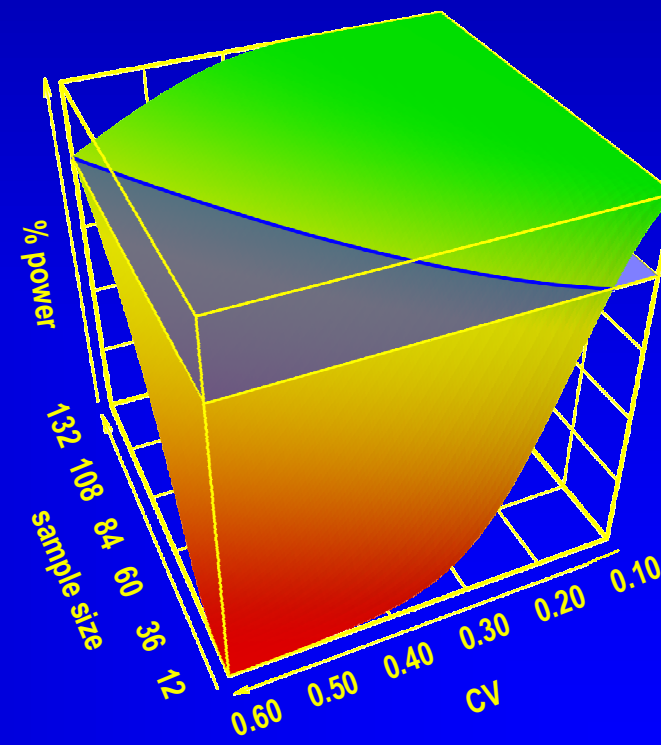
# Power Surfaces

2x2x2 crossover study, target power 80%

CV 0.30



GMR 0.95



# Power vs. Sample Size

- **How many subjects are enough?**
  - Most guidelines recommend 80 – 90% power.
  - If a study is planned for  $\leq 70\%$  power, problems with the ethics committee are possible (ICH E9).
  - If a study is planned for  $>90\%$  power (especially with low variability drugs), additional problems with regulators are possible ('forced bioequivalence').
  - Some subjects ('alternates') should be added to the estimated sample size according to the expected drop-out rate – especially for studies with more than two periods or multiple-dose studies.

# Hierarchy of Designs

- The more ‘sophisticated’ a design is, the more information can be extracted.

- Hierarchy of designs:

Full replicate (TRTR | RTRT or TRT | RTR), ↗

Partial replicate (TRR | RTR | RRT) ↗

Standard 2×2×2 cross-over (TR | RT) ↗

Parallel (R | T)

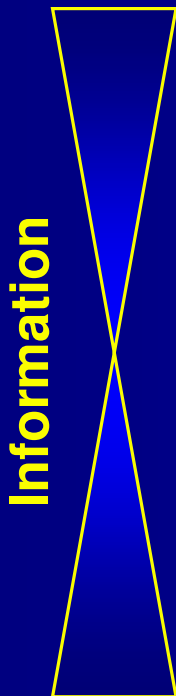
- Variances which can be estimated:

Parallel: total variance (between + within subjects)

2×2×2 Xover: + between, within subjects ↗

Partial replicate: + within subjects of reference ↗

Full replicate: + within subjects of reference and test ↗



# Coefficient(s) of Variation

- From any design one gets variances of designs which are lower in the hierarchy as well.

- Total CV% from a 2×2×2 crossover used in planning a parallel design study:

- Intra-subject CV% (within)

$$\longrightarrow CV_{\text{intra}} \% = 100 \cdot \sqrt{e^{MSE_W} - 1}$$

- Inter-subject CV% (between)

$$\searrow CV_{\text{inter}} \% = 100 \cdot \sqrt{e^{\frac{MSE_B - MSE_W}{2}} - 1}$$

- Total CV% (pooled)

$$\downarrow CV_{\text{total}} \% = 100 \cdot \sqrt{e^{\frac{MSE_B + MSE_W}{2}} - 1}$$

# Coefficient(s) of Variation

- However, CVs of *higher* design levels not available.

- If only mean  $\pm$  SD of reference is available...

- ‘Rule of thumb’  $CV_{intra} \approx 60\%$  of  $CV_{total}$  not correct.

- A crossover study must not be planned based on  $CV_{total}$ .

- Examples (2x2x2 studies)

drug	formulation	design	n	PK metric	$CV_{intra}$	$CV_{inter}$	$CV_{total}$
methylphenidate	biphasic MR	SD	12	$AUC_{0-t}$	7.0	19.1	20.4
laroxetine	XR	MD	32	$AUC_{0-\tau}$	25.2	55.1	62.1
lansoprazole	DR	SD	47	$C_{max}$	47.0	25.1	54.6

- Pilot study unavoidable, unless a TSD is used.

# Tools

- Sample Size Tables (Phillips, Diletti, Hauschke, Chow, Julious, ...)
- Approximations (Diletti, Chow, Julious, ...)
- General purpose (SAS, S+, R, StaTable, ...) and specialized software (nQuery Advisor, PASS, FARTSSIE, StudySize, ...)
- Exact method (*R*-package *PowerTOST*)  
Currently only *PowerTOST* provides sample size estimation for reference-scaling (EMA, FDA) and FDA's method for NTIDs.

# Sensitivity Analysis

- ICH E9 (1998)

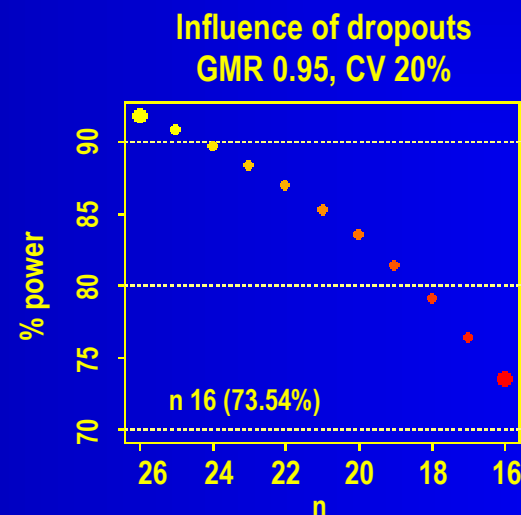
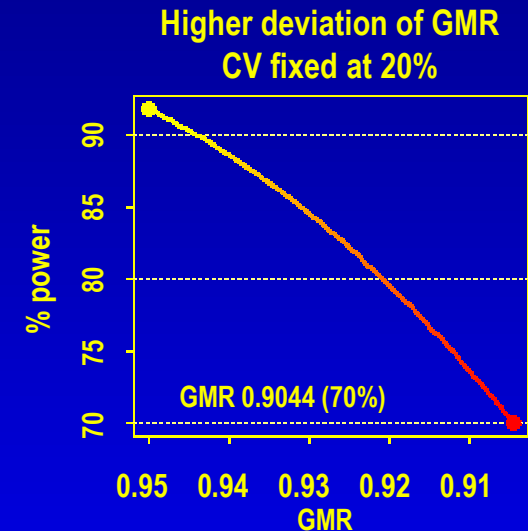
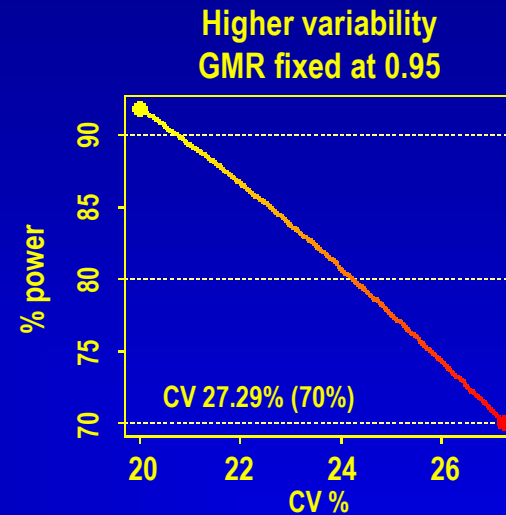
- Section 3.5 Sample Size, paragraph 3

- The method by which the sample size is calculated should be given in the protocol [...]. The basis of these estimates should also be given.
- It is important to investigate the sensitivity of the sample size estimate to a variety of deviations from these assumptions and this may be facilitated by providing a range of sample sizes appropriate for a reasonable range of deviations from assumptions.
- In confirmatory trials, assumptions should normally be based on published data or on the results of earlier trials.

# Sensitivity Analysis

## ● Example

- CV 20%, GMR 0.95, AR 80 – 125%, target power 90%, min. acceptable 70%.
- Estimated sample size: 26 (91.76% power)
- Acceptable relative deviations:
  - CV +36.4%
  - GMR -4.80%
  - n -38.5%





# Sensitivity Analysis

- Has to be done *before* the study (*a priori*).
- The Myth of retrospective (*a posteriori*) Power:<sup>1,2,3</sup>
  - High power *does not* further support the claim of already demonstrated bioequivalence.
  - Low power *does not* invalidate the conclusion of claimed bioequivalence.

1. Lenth RV (2000)

*Two Sample-Size Practices that I don't recommend*

<http://www.math.uiowa.edu/~rlenth/Power/2badHabits.pdf>

2. Hoening JM, Heisey DM (2001)

*The Abuse of Power: The Pervasive Fallacy of Power Calculations for Data Analysis*

The American Statistician 55(1):19–24

[http://www.vims.edu/people/hoenig\\_jm/pubs/hoenig2.pdf](http://www.vims.edu/people/hoenig_jm/pubs/hoenig2.pdf)

3. Bacchetti P (2010)

*Current sample size conventions: Flaws, harms, and alternatives*

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*Thank You!*

# Sample Size Estimation for BE Studies

*Questions?*



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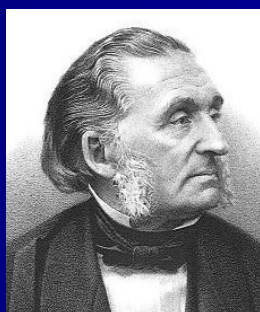
# To bear in Remembrance...

**Power.** That which statisticians are always calculating but never have.

**Power:** That which is wielded by the priesthood of clinical trials, the statisticians, and a stick which they use to beta their colleagues.

**Power Calculation –** A guess masquerading as mathematics.

***Stephen Senn***



**You should treat as many patients as possible with the new drugs while they still have the power to heal.**

***Armand Trousseau***