







To bear in Remembrance...

Whenever a theory appears to you as the only possible one, take this as a sign that you have neither understood the theory nor the problem which it was intended to solve.

Even though it's *applied* science we're dealin' with, it still is – *science!*



Karl R. Popper

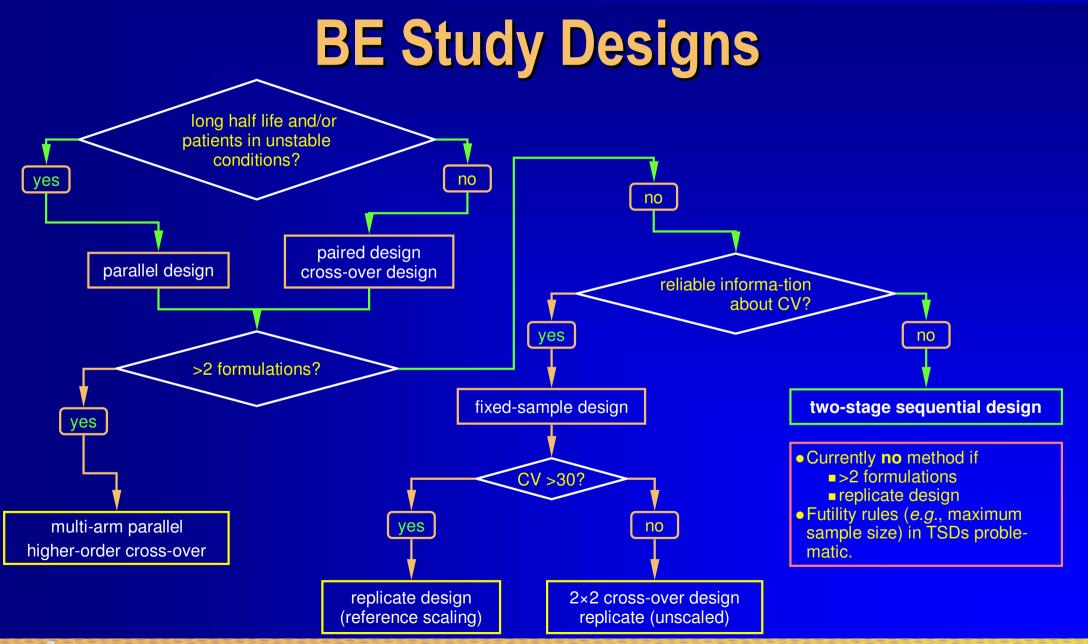


Leslie Z. Benet



Introduction • AOD • GSD • TSD • Case Studies • Outlook









Add-on / Two-Stage Designs

Sometimes properly designed studies fail due to

- 'true' bioinequivalence,
- pure chance (producer's risk),
- poor study conduct (increasing variability),
- false (mainly over-optimistic) assumptions about the CV and/or T/R-ratio – leading to a too small sample size (insufficient power).

•The sample size is planned based on assumptions...





Add-on / Two-Stage Designs

- Dealing with inconclusive BE studies (confidence interval) not entirely with the acceptance range) Repeat the study in a larger sample size. Optionally perform a meta-analysis of pooled data. Only acceptable if at least one study demonstrates BE. Recruit a second group of subjects and pool data? Discussed at Bio-International Conferences (1989, 1992) and guidelines from the 1990s. The patient's risk must be preserved!
 - Among rivaling methods the one with with the highest power should be selected.





Terminology

•Add-On Designs

Sample sizes of both groups have a lower limit.

Group Sequential Designs

Sample sizes of both groups are pre-specified.

Adaptive Two-Stage Sequential Designs

- Groups sizes are (generally) not limited.
- Sample size of the second group is re-estimated from the first group's data.

H Schütz

Two-stage designs in bioequivalence trials Eur J Clin Pharmacol (2015) <u>DOI: 10.1007/s00228-015-1806-2</u>





Definition

 For an overview see Schwartz & Denne, Dragalin, Chow & Chang, and Chin

A study design is called *adaptive* if statistical methodology allows modification of a design element (*e.g.*, the sample size) at an interim analysis with full control of the type I error (TIE).

Schwartz TA and JS Denne JS (2003) Common threads between sample size recalculation and group sequential procedures Pharm Stat 2, 263–271. DOI: 10.1002/pst.068 Dragalin V (2006) Adaptive Designs: Terminology and Classification Drug Info J 40, 425–435 Chow S-C and M Chang (2012) Adaptive Design Methods in Clinical Trials 2nd edn. Chapman & Hall/CRC, Boca Raton Chin R (2012) Adaptive and Flexible Clinical Trials Chapman & Hall/CRC, Boca Raton





Add-On Designs: Guidelines

General conditions

- Intention to perform an AOD has to be stated in the protocol,
- the same batches of products, and
- the same clinical and bioanalytical methods have to be employed in both groups.

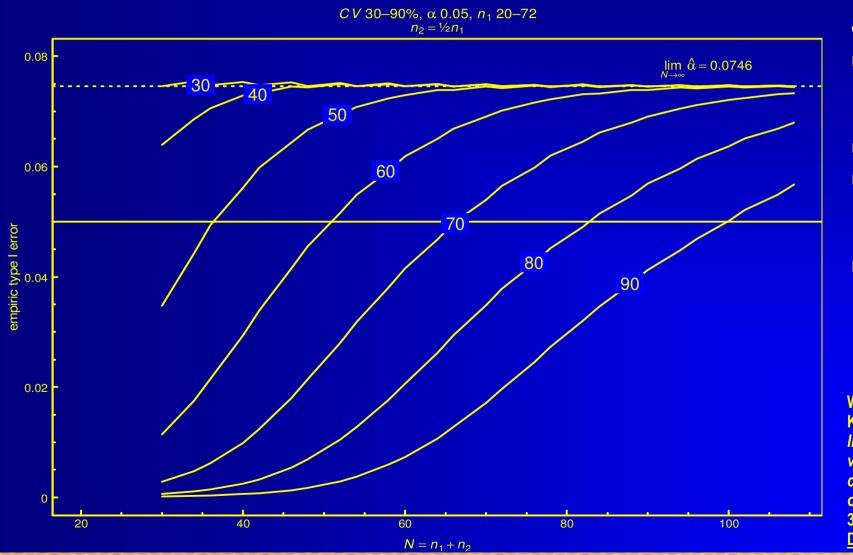
Currently only stated in GLs of Japan, Argentina, Mexico, and Korea

The patient's risk might be seriously compromised!





Add-on / Two-Stage Designs



Japan (2012)

- Ist group (n₁) ≥20 evaluated with α 0.05 (90% CI)
- ■ 2^{nd} group $(n_2) \ge \frac{1}{2}n_1$
- Pooled data evaluated with α 0.05 (90% CI)
- Inflation of the patient's risk (up to 7.5%)!

Wonnemann M, Frömke C, and A Koch (2015) Inflation of the Type I Error: Investigations on Regulatory Recommendations for Bioequiva-Ience of Highly Variable Drugs Pharm Res 32(1), 135–43 DOI: 10.1007/s11095-014-1450-z





Group Sequential Designs

- Long and accepted tradition in clinical research (mainly phase III)
 - Based on work by Armitage et al. (1969), McPherson (1974), Pocock (1977), O'Brien and Fleming (1979), Lan and DeMets (1983), ...
 - Developed for superiority testing, normal distributed data with known variance, fixed and equal sizes of groups.
 - First proposal by Gould (1995) in the field of BE did not get regulatory acceptance in Europe.

AL Gould Group Sequential Extension of a Standard Bioequivalence Testing Procedure J Pharmacokin Biopharm 23(1), 57–86 (1995) DOI: 10.1007/BF02353786





Group Sequential Designs: GLs

- Australia (2004), Canada (Draft 2009) - Application of Bonferroni's correction (α 0.025). **Theoretical TIE** \leq 0.0494. For CVs and samples sizes typical in BE ≤ 0.04 . •Canada (2012) Pocock's α 0.0294. $\square n_1$ based on 'most likely variance' + additional subjects to compensate for expected dropout-rate. Total sample size based on 'worst-case scenario'.
 - If $n_2 \neq n_1$ relevant inflation of the TIE is possible!





Adaptive TS Sequential Designs

- Methods by Potvin et al. (2008) first validated framework in the context of BE
 - Supported by the 'Product Quality Research Institute' (members: FDA/CDER, Health Canada, USP, AAPS, PhRMA...).
 - Inspired by conventional BE testing and Pocock's α 0.0294 for Group Sequential Designs.

Potvin D, DiLiberti CE, Hauck WW, Parr AF, Schuirmann DJ, and RA Smith Sequential design approaches for bioequivalence studies with crossover designs Pharm Stat 7(4), 245–62 (2008) <u>DOI: 10.1002/pst.294</u>





Adaptive TS Sequential Designs

Two 'types' of TS Sequential Designs

- 1. The same adjusted α is applied in both stages (regardless whether a study stops already in the first stage or proceeds to the second stage).
 - Based on Group Sequential Design.
 - In publications called Method B.
- 2. An unadjusted α may be used in the first stage (dependent on interim power).
 - Based on conventional BE testing + GSD.
 - In publications called Method C, D, C/D.





Review of Guidelines

•EMA (Jan 2010)

Acceptable; Potvin et al. Method B preferred (?)

Canada (May 2012)

Potvin et al. Method C recommended.

•FDA (Jun 2012)

Potvin *et al.* Method C/D recommended. API specific guidance: Loteprednol

Russia (2013)

Acceptable; Potvin et al. Method B preferred (?)





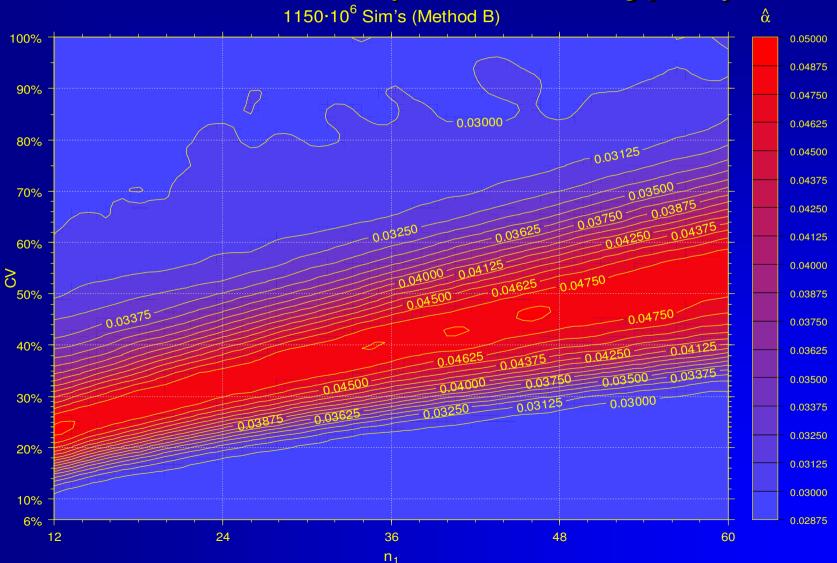
Potvin et al. (Method B – type 1) Evaluate BE at stage 1 (α 0.0294) BE met? no yes Evaluate power at stage 1 using α -level of 0.0294 ≥80%? yes no Estimate sample size based on CV_{intra}, T/R 0.95, α 0.0294; continue to stage 2 Evaluate BE at stage 2 using pooled data from both stages (α 0.0294) Fail Pass or fail Pass



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Potvin et al. (Method B – type 1)





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Potvin et al. (Method B – type 1) $1150 \cdot 10^6$ Sim's (Method B) $1-\hat{\beta}$ 100% 1.000 0.986 0.972 90% 0.958 0.944 80% 0.930 0.916 70% 0.902 0.888 60% 0.874 0.818 20 0.860 50% 0.846 X 0162 0.832 40% 0.818 0.8320.846 0.804 - 0.860 30% 0.888 0.902 0.790 -0.874 0.930 -0.916 0.776 0.972 0 958 0.944 0.986 20% 0.762 0.748 .000 0.734 10% .000 6% 0.720 24 12 36 48 60

 n_1





Technical Aspects

- Only one Interim Analysis (after stage 1).
- Potvin *et al.* used a simple power estimation based on the shifted central *t*-distribution. Use software (avoid approximations). Example 2: method
- Should be termed 'Interim Power Analysis' – not 'Bioequivalence Assessment' in the protocol.

method	% power
approx. (shifted central t)	50.49
approx. (noncentral t)	52.16
exact (Owen's Q)	52.51

- No post hoc Power only a validated method to guide the decision tree.
- No adjustment for T/R-ratio observed in stage 1!



Technical Aspects (cont'd)

- No futility rule preventing to go into stage 2 with a high sample size!
 - Must be clearly stated in the protocol (unfamiliar to the IEC because common in Group Sequential Designs).
- Pocock's α 0.0294 is used in stage 1 and in the pooled analysis (data from stages 1 + 2),
 - *i.e.*, the $100(1 2 \times \alpha) = 94.12\%$ CI is calculated.
- Overall TIE preserved at ≤0.05.





Technical Aspects (cont'd) + EMA modification

- If the study is stopped after stage 1, the statistical model is:
 - fixed: sequence + period + treatment
 - + subject (sequence)

If the study continues to stage 2, the model for the combined analysis is:

- fixed: stage + sequence + sequence(stage)
 - + subject(sequence × stage) + period(stage)
 - + treatment

No poolability criterion! Combining is always allowed – even if a significant difference between stages is observed. No need to test this effect.



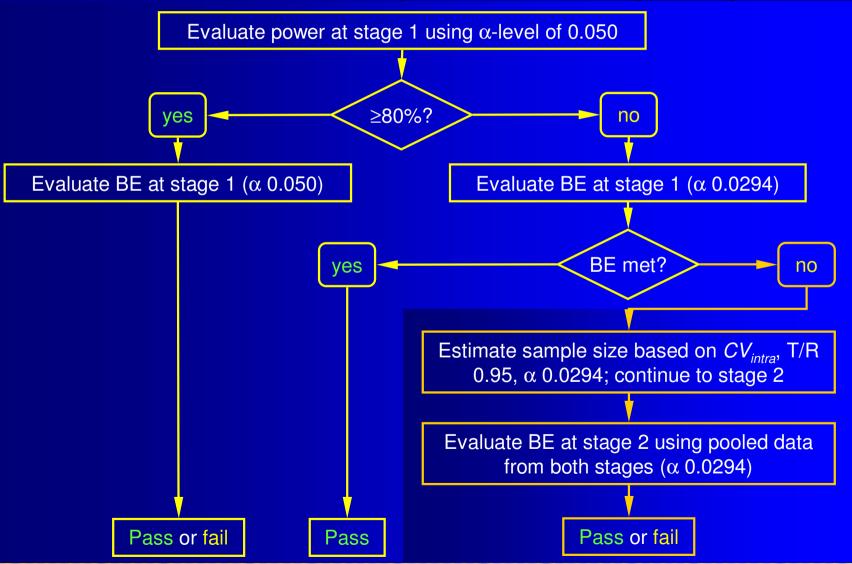


Technical Aspects (cont'd) + EMA modification Incomprehensible why this modification was introducted by EMA's Biostatistical Working Party Simulations performed or "gut feeling"? Modification shown to be irrelevant. Furthermore no difference whether subjects were treated as a fixed or random term (unless T/R >1.20).

Karalis V and P Macheras On the Statistical Model of the Two-Stage Designs in Bioequivalence Assessment J Pharm Pharmacol 66(1), 48–52 (2014) DOI: 10.1111/jphp.12164









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Potvin et al. (Method C – type 2) 1150.10⁶ Sim's (Method C) $\hat{\alpha}$ 100% 0.0520 0.0506 0.0500 90% 0.0492 0.0478 80% 0.0338 0.0464 70% 0.0324 0.0450 0.0352 0.0380 0.0436 0.0450 0.042 60% 0.040 0.0464 0.0478 0.0422 20 0.0436 50% 0.0394 0.0408 0.0394 40% 0.036 0.0380 0.0492 0.0366 30% 0.0478 2 2 0.0352 0.0500 0.0500 20% 0.0338 0.0324 0.0500 10% / 0.0500 -6% 0.0310 24 12 36 48 60

 n_1





Pros & Cons

- Method C (*if power* ≥80%) is a conventional BE study; no penalty in terms of α needs to be applied.
- Method C proceeds to stage 2 less often and has smaller average total sample sizes than Method B for cases where the initial sample size is reasonable for the CV.
- If the size of stage 1 is low for the actual CV both methods proceed to stage 2 almost all the time; total sample sizes are similar.
- Method B slightly more conservative than C.





Type 1/2

Recommendations

- Type 2 preferred due to slightly higher power than type 1 (FDA, HPFB). Type 1 for EMA (?)
- Plan the study as if the CV is known
 - If assumptions turn out to be true = no penalty
 - If lower power (CV higher than expected), BE still possible in first stage (penalty; 94.12% CI) or continue to stage 2 as a 'safety net'.
- Don't jeopardize! Small sample sizes in the first stage don't pay off. Total sample sizes are ~10–20% higher.





TSDs: Alternatives

Methods by Potvin *et al.* (2008) limited to *T/R* of 0.95 and 80% power

Follow-up publications (*T*/*R* 0.95...0.90, 80...90% power)

reference	type	method	T/R	target power	CV	$lpha_{adj}$	max. TIE
Potvin e <i>t al.</i>	1	В	0.95	80%	10–100%	0.0294	0.0485
	2	С					0.0510
Montague <i>et al.</i>	2	D	0.90			0.0280	0.0518
Fuglsang	1	В	0.95 0.90	90%	10–80%	0.0284	0.0501
	2	C/D				0.0274	0.0503
	2	C/D				0.0269	0.0501

Montague TH, Potvin D, DiLiberti CE, Hauck WW, Parr AF, and DJ Schuirmann

Additional results for 'Sequential design approaches for bioequivalence studies with crossover designs' Pharm Stat 11(1), 8–13 (2011) DOI: 10.1002/pst.483

A Fuglsang

Sequential Bioequivalence Trial Designs with Increased Power and Controlled Type I Error Rates AAPS J 15(3), 659–61 (2013) DOI: 10.1208/s12248-013-9475-5





TSDs: Alternatives

Slight inflation of the TIE in some 'type 2' designs could easily be avoided

- Modifications of published adjusted α

type	method	T/R	$lpha_{adj}$	max. TIE	$lpha_{adj}^{*}$	max. TIE*
1	В	0.05	0.95 0.0294	0.0485	0.0304	0.0501
2	С	0.95		0.0510	0.0282	0.0500
2	D	0.90	0.0280	0.0518	0.0270	0.0500

* Schütz H, Labes D, and A Fuglsang Modifications of 'Sequential design approaches for bioequivalence studies with crossover designs' in preparation (2015)



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Montague et al. (Method D – type 2) 1150·10⁶ Sim's (Method D) å 100% 0.05100 0.05000 0.04865 90% 0.04653 80% 0.04442 70% 0.035960 0.04230 0.040190 0.033845 60% 0.044420 0.03807 0.04019 0.029615 20 0.042305 50% 0.046535 0.03807 0.031730 40% 0.03596 0.03385 30% 0.046535 0.03173 0500 20% 0.048650 0.050000 0.02961 0.050000 10% 6% 0.02750 24 12 36 48 60

 n_1





Parallel Groups (Type 1/2)

•Fuglsang (2014)

Based on Potvin's Methods B/C (α_{adi} 0.0294, 80% power)

Framework: *n*₁ 48–120, *CV* 10–100%

- equal allocation $(N_{Test} = N_{Reference})$
- equal and unequal variances of groups
- conventional *t*-test and Welch-Satterthwaite approximation

Results

- No significant inflation of the TIE
- Power ≥78.4%

A Fuglsang

Sequential Bioequivalence Approaches for Parallel Designs AAPS J 16(3), 373–8 (2014), DOI: 10.1208/s12248-014-9571-1





Futility Rules revised

• EMA GL Section 4.1.8 'Two-stage design'

"[...] the stopping criteria should be clearly defined prior to the study."

- What does that mean?
 - Failing in stage 1 or the pooled analysis according to the chosen method.

 \rightarrow Part of the validated frameworks.

Early stopping for futility (e.g., 'bad' ratio, extreme stage 2 sample size caused by high CV – better to opt for reference-scaling...).
 → Not validated. A misunderstanding by regulators (stopping criterion ≠ futility rule).





Futility Rules revised

 Introduction of a futility rule does not inflate the TIE, but power may drop substantially!

- State stopping criteria unambiguously in the protocol.
- If you want to introduce a futility rule, simulations are mandatory in order to maintain sufficient power.

"Introduction of [...] futility rules may severely impact power in trials with sequential designs and under some circumstances such trials might be unethical."

A Fuglsang Futility Rules in Bioequivalence Trials with Sequential Designs APPS J 16(19), 79–82 (2014) <u>DOI: 10.1208/s12248-013-9540-0</u>





Validation of Frameworks

Jones and Kenward concluded that

"[...] before using any of the methods [...], their operating characteristics should be evaluated for a range of values of n_1 , CV and true ratio of means that are of interest, in order to decide if the Type I error rate is controlled, the power is adequate and the potential maximum total sample size is not too great."

Uncomplicated with current software

Automatically finding a suitable α_{adj} and validating for TIE and power takes ~20 minutes.

Jones B and MG Kenward Design and Analysis of Cross-Over Trials Chapman & Hall/CRC, Boca Raton (3rd ed. 2014) D Labes Package 'Power2Stage', Version 0.2-2 (2014-12-08)





•Consider certain questions:

- Is it possible to assume a best/worst-case scenario?
- How large should the size of the first stage be?
- How large is the expected sample size in the second stage?
- Which power can one expect in both stages?
- Will intruction of a futility criterion substantially decrease power?
- Is there a sample size penalty compared to a fixed-sample design?





• Example

Expected CV 20%, desired power is 80% for a *T/R*-ratio of 0.95. Comparison of a type 1 TSD with a conventional fixed-sample design (*n* 20, 83.5% power).

n ₁	E [N]	Studies stopped in stage 1 (%)	Studies failed in stage 1 (%)	Power in stage 1 (%)	Studies in stage 2 (%)	Final power (%)	Costs (%)
12	20.6	43.6	2.3	41.3	56.4	84.2	+2.9
14	20.0	55.6	3.0	52.4	44.5	85.0	+0.2
16	20.1	65.9	3.9	61.9	34.1	85.2	+0.3
18	20.6	74.3	5.0	69.3	25.7	85.5	+3.1
20	21.7	81.2	6.3	74.9	18.8	86.2	+8.4
22	23.0	87.2	7.3	79.8	12.8	87.0	+15.0
24	24.6	91.5	7.9	83.6	8.5	88.0	+22.9





Example (cont'd)

- With 14 or 16 subjects in the first stage similar costs (*E*[*N*] ~20) are expected; with 16 one has a 66% chance to stop the study already in the first stage (62% chance to pass and 4% to fail).
- With n₁ equal to the fixed design's n costs are expected to be 8% higher but we have a 75% chance to pass in the first stage and 86% power overall.
- Power of the TSD is always larger than the one of the fixedsample design – regardless the initial sample size and even if the assumed CV turns out to be correct.





Example (cont'd)

If in a fixed-sample design the CV turns out to be higher than the assumed one, power will decrease, whereas in a TSD power is maintained.

Don't start the first stage always in a small group and hope for a smaller than expected CV – which would be substan-tially more economic than a fixed-sample design. This is not necessarily a good idea: With 12 subjects power in the first stage is only 41% and 56% of studies will proceed to the second stage.





Advanced Example

•'Must pass' BE in stage 1 (first to file)

- Fixed *T*/*R* 90% (pessimistic; very likely better).
- Expected CV 20% (pilot study with two references).
- **Expected dropout rate ~30%; start with 88 to have** $n_1 \ge 60$.
- Targets
 - >90% power for n_1 60 even for extreme *CV* of 45%.
 - **90%** power for $n_1 \ge 60$ (*CV* 20%) in stage 1.
 - Not <80% power for CV ≥25% in stage 1.
 - Low probability to proceed to stage 2.





Advanced Example

•'Must pass' BE in stage 1 (first to file)

- Sponsor prefered Method B (EU submission...).
- Fuglsang published α_{adj} 0.0269 for *T/R* 0.90 and 90% power – but only for Method C...
- Same α_{adj} applicable for Method B?
- Likely...
 - Potvin et al. showed less inflation of the TIE with Method B.
 - Fuglsang needed less adjustment in Method B.
 - But we have to justify that!
- 10⁶ simulations for the TIE and 10⁵ for power.





Advanced Example

•'Must pass' BE in stage 1 (first to file)

Targets met

- **93%** power for n_1 60 (CV 20%) and 90% for extreme CV of 45%.
- **90%** power for $n_1 \ge 60$ (CV 20%) in stage 1.
- Low chances to proceed to stage 2 with CV 20%: n_1 60: 6%, n_1 72: 1%
- \geq 80% power for *CV* \geq 20% even for a more extreme dropout rate.
- α_{adi} 0.0271 would work as well (with 0.0278 < 0.052).</p>
- Study passed in the first stage (February 2014)





Case Study 1

 Method C: Study passed BE in stage 1 (49 subjects, CV 30.65%, 90% CI)

- **UK**/Ireland: Unadjusted α in stage 1 not acceptable.
- Study passed BE with 94.12% CI as well (post hoc switch to Method B).
 - Austria: The Applicant should demonstrate that the type I error inflation, which can be expected from the chosen approach, did not impact on the decision of bioequivalence.
 - One million simulations based on the study's sample size and CV. TIE 0.0494 (95% CI: 0.0490 – 0.0498)





Case Study 2

- Method C: Study stopped in stage 1
 AUC power >80%: passed BE with 90% CI
 C_{max} power <80%: passed BE with 94.12% CI
 - The Netherlands: Adapting the confidence intervals based upon power is not acceptable and also not in accordance with the EMA guideline. Confidence intervals should be selected a priori, without evaluation of the power. Therefore, the applicant should submit the 94.12% confidence intervals for AUC.
 - AUC failed BE with 94.12% CI.
 - Sponsor repeated the study with a very (!) large sample size and failed on C_{max}. Project cancelled.





Case Study 3

- Method C: Two studies passed in stage 1 (SD n=15, MD n=16; C_{max} CV 17.9%, 8.54%; 90% CIs)
- •Would have passed with Method B as well; however, 94.12% CIs were *not* reported.
 - RMS Germany. Accepted by CMSs Austria, Denmark, Sweden, and The Netherlands.
 - Spain: Statistical analysis should be GLM. Please justify.
 - Evaluated with fixed-effects model.
 Both studies passed.
 Issue resolved (September 2013)





Conclusions

- Do not blindly follow guidelines. Some curent recommendations may lead to inflation of the patient's risk and/or deteriorate power.
- Validated frameworks can be applied without requiring the sponsor to perform own simulations – though they could further improve power based on additio-nal assumptions.
- Two-stage designs are both ethical and economical alternatives to fixed-sample designs.





Outlook

- •Feasibility / futility rules.
- Arbitrary expected T/R and/or power.
- Methods without interim power.
- Dropping a candidate formulation from a higher-order cross-over; continue with 2×2.
- •Continue a 2×2 in replicate design for scaling.
- •Fully adaptive methods.
- Exact method (not depending on simulations).
- Application to replicate designs / scaling.



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XII Congreso de la Sociedad Española de Farmacia Industrial y Galénica | Barcelona, 27 January 2015



¡Gracias! Two-Stage Sequential Designs in Bioequivalence Open Questions?



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XII Congreso de la Sociedad Española de Farmacia Industrial y Galénica | Barcelona, 27 January 2015



To bear in Remembrance...

The fundamental cause of trouble in the world today is that the
stupid are cocksure while
the intelligent are full of doubt.Bertrand Russell

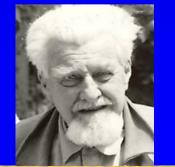




In bioequivalence we must not forget the only important – *the patient*! He/she is living person, not just α 0.05.

Dirk Marteen Barends

It is a good morning exercise for a research scientist to discard a pet hypothesis every day before breakfast. It keeps him young.







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