

Statistical aspects of reference-scaled studies

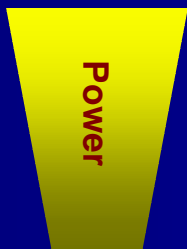
Helmut Schütz
BEBAC

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Hierarchy of Designs

- The more ‘sophisticated’ a design is , the more information (in terms of σ^2) we may obtain.

- Hierarchy of designs:



- Full replicate (TRTR | RTRT) ↗
- Partial replicate (TRR | RTR | RRT) ↗
- Standard 2x2 cross-over (RT | RT) ↗
- Parallel (R | T)

- Variances which can be estimated:

Parallel: total variance (between + within)

2x2 Xover: + between, within subjects ↗

Partial replicate: + within subjects (reference) ↗

Full replicate: + within subjects (reference, test) ↗

Variations

- For Highly Variable Drugs / Drug Products (HVDs/HVDPs) it may be almost impossible to show BE with a reasonable sample size.
- The common 2x2 Xover assumes Independent Identically Distributions (IID), which may not hold. If e.g., the variability of the reference is higher than the one of the test, one obtains a high common (pooled) variance and the test will be penalized for the 'bad' reference.

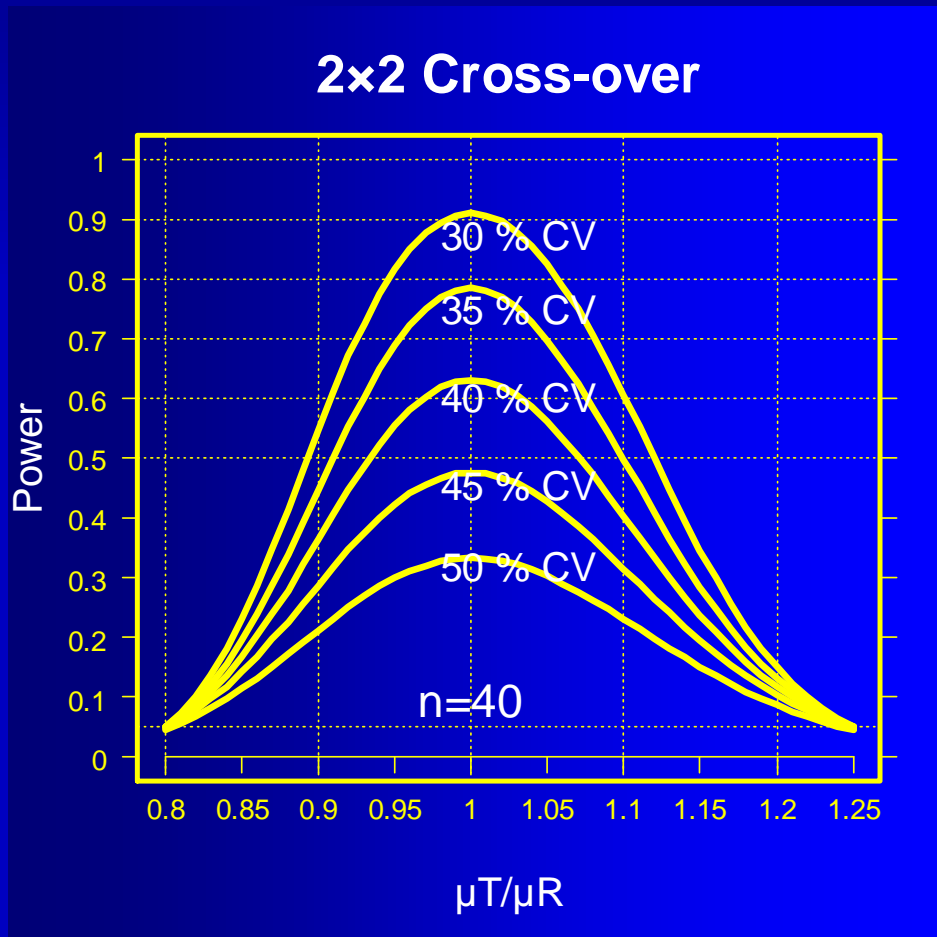
Variances

Power to show BE
with 40 subjects for
 $CV_{intra} = 30-50\%$

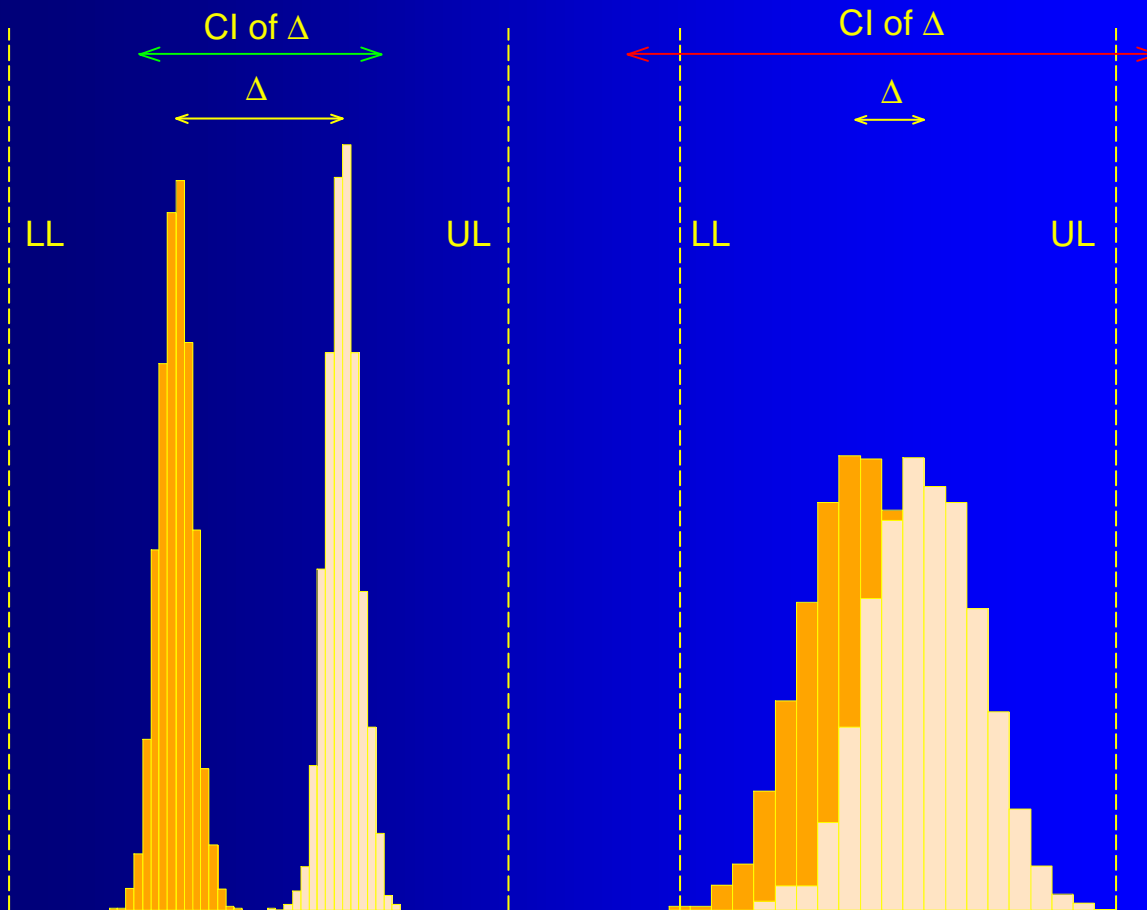
μ_T/μ_R 0.95, CV_{intra} 30%
→ power 0.816

μ_T/μ_R 1.00, CV_{intra} 45%
→ power 0.476 <
Roulette 0.486 (!)

μ_T/μ_R 0.95, CV_{intra} 50%
→ n=98 (power 0.803)



Variations



Modified from Fig. 1
 L Tóthfalusi, L Endrenyi and
 A García Arieta
 Evaluation of Bioequivalence
 for Highly Variable Drugs
 with Scaled Average
 Bioequivalence
 Clin Pharmacokinet 48,
 725–743 (2009)

Counterintuitive
 concept of BE:
 Two formulations with
 a large difference in
 means are declared
 bioequivalent if vari-
 ances are low, but not
 bioequivalent – even
 if the difference is
 quite small – due to
 high variability.

Replicate designs

- Each subject is randomly assigned to sequences, where *at least one* of the treatments is administered *at least twice*.
 - Not only the *global within-subject variability*, but also the *within-subject variability per treatment* may be estimated.
 - Smaller subject numbers compared to a standard 2x2x2 design – but outweighed by an increased number of periods. Note: Same overall number of individual treatments!

Replicate designs

- Required if reference-scaled average bioequivalence (RSABE) is targeted or widening of the AR for C_{\max} (for countries following the 'old' EU guideline).
- Advantages
 - Some experience from FDA's initiative on Population Bioequivalence (PBE) and Individual Bioequivalence (IBE).
 - Mentioned in RSA's GL; FDA's API GLs and EMA.
 - RSABE of different metrics acceptable in some countries (FDA, RSA AUC/ C_{\max} , EMA C_{\max} , TGD AUC).
 - Handling of outliers (Subject-by-Formulation Interaction may be ruled out).

Replicate designs

■ Disadvantages

- Statistical analysis quite complicated (especially in the case of drop-outs and if RSABE is the target) – not available in standard software.
- Many publications, but still no agreement on methodology (!)
- SAS-code published by the FDA in April 2010:
<http://www.fda.gov/downloads/Drugs/GuidanceComplianceRegulatoryInformation/Guidances/UCM209294.pdf>
- Handling of outliers. For the EMA it has to be shown that $CV_{WR} > 30\%$ is not caused by outliers. Method?
- SAS-code and two example datasets published by the EMA in March 2011:
http://www.ema.europa.eu/docs/en_GB/document_library/Scientific_guideline/2009/09/WC500002963.pdf

Replicate designs

■ Designs

- Two-sequence three-period

T R T

R T R

Sample size to obtain the same power as a 2x2x2 study: 75%

- Two-sequence four-period

T R T R

R T R T

Sample size to obtain the same power as a 2x2x2 study: 50%

- **and many others...** (FDA: TRR|RTR|RRT aka 'partial replicate')

- The statistical model is quite complicated – and dependent on the actual design!

$$X_{ijkl} = \mu \cdot \pi_k \cdot \Phi_l \cdot s_{ij} \cdot e_{ijkl}$$

Application: HVDs/HVDPs

- Highly Variable Drugs / Drug Products ($CV_{WR} > 30\%$)
 - ✓ USA Recommended in product specific guidances. GMR 0.80 – 1.25. Minimum sample size 24.
 - ✓ CAN 2010 draft GL. Scaling for AUC only. No restriction on GMR.
 - ± EU Widening of acceptance range (for C_{max} only: to maximum 69.84% – 143.19%), if CV_{WR} in the study $> 30\%$. GMR 0.80 – 1.25. Demonstration that $CV_{WR} > 30\%$ not caused by outliers.

Application: HVDs/HVDPs

- All (!) ANDAs submitted to FDA/OGD 2003 – 2005 (1010 studies, 180 drugs)
 - 31% (57/180) highly variable ($CV \geq 30\%$)
 - of these HVDs/HVDPs,
 - 60% due to PK (e.g., first pass metabol.)
 - 20% formulation performance
 - 20% unclear

Davit BM, Conner DP, Fabian-Fritsch B, Haidar SH, Jiang X, Patel DT, Seo PR, Suh K, Thompson CL, and LX Yu

Highly Variable Drugs: Observations from Bioequivalence Data Submitted to the FDA for New Generic Drug Applications

The AAPS Journal 10/1, 148–56 (2008)

<http://www.springerlink.com/content/51162107w327883r/fulltext.pdf>

HVDPs (US/EU)

- Advisory Committee for Pharmaceutical Sciences (ACPS) to FDA (10/2006) on HVDs
- Follow-up papers in 2008 (ref. in API-GLs)
 - Replicate study design [TRR–RTR–RRT]
 - Reference Scaled Average Bioequivalence (RSABE)
 - Minimum sample size 24 subjects
 - GMR restricted to [0.80,1.25]

Haidar SH, Davit B, Chen M-L, Conner D, Lee LM, Li QH, Lionberger R, Makhlouf F, Patel D, Schuirmann DJ, and LX Yu

Bioequivalence Approaches for Highly Variable Drugs and Drug Products

Pharmaceutical Research 25/1, 237-241 (2008)

<http://www.springerlink.com/content/u503p62056413677/fulltext.pdf>

Haidar SH, Makhlouf F, Schuirmann DJ, Hyslop T, Davit B, Conner D, and LX Yu

Evaluation of a Scaling Approach for the Bioequivalence of Highly Variable Drugs

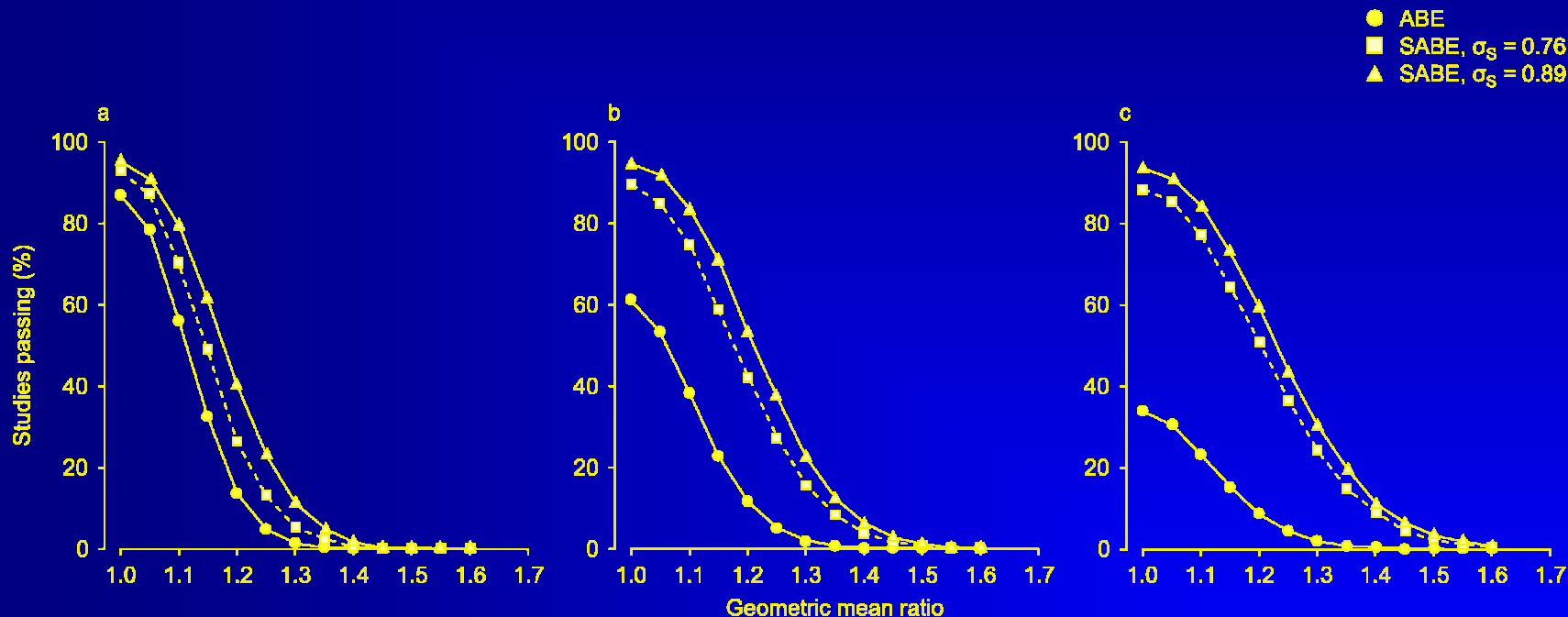
The AAPS Journal, 10/3, (2008) DOI: [10.1208/s12248-008-9053-4](https://doi.org/10.1208/s12248-008-9053-4)

HVDs/HVDPs

- Replicate designs

- 4-period replicate designs:
sample size = $\frac{1}{2}$ of 2x2 study's sample size
- 3-period replicate designs:
sample size = $\frac{3}{4}$ of 2x2 study's sample size
- Reminder: number of treatments (and biosamples) identical to the conventional 2x2 cross-over.
- Allow for a safety margin – expect a higher number of drop-outs due to the additional period(s).
- Consider increased blood loss (ethics!)
Eventually bioanalytics has to be improved.

HVDPs (US/EU)



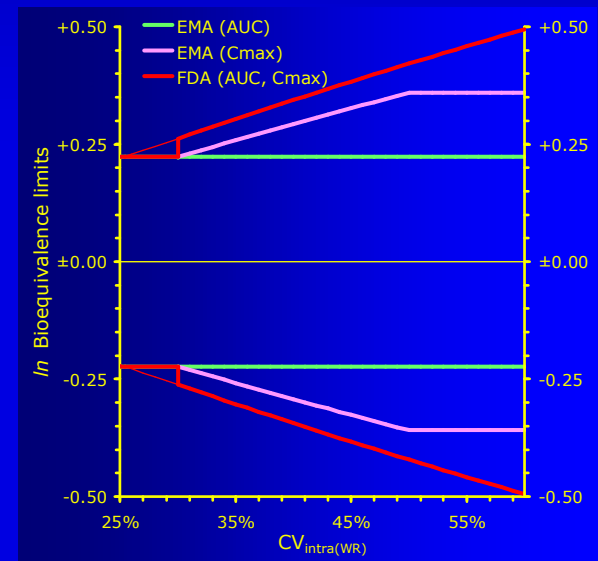
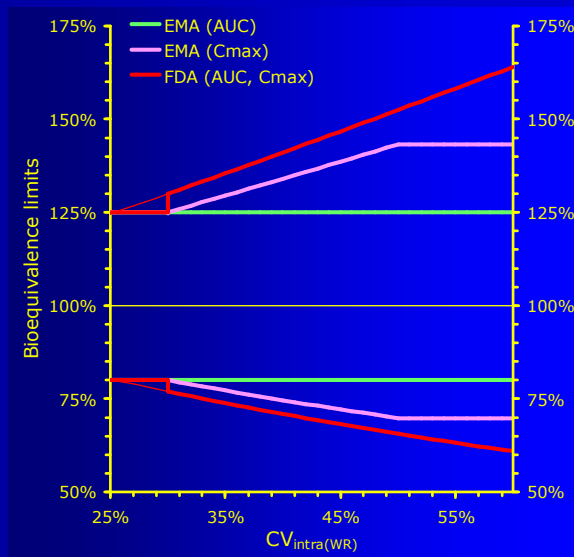
Tóthfalusi *et al.* (2009), Fig. 3

Simulated (n=10000) three-period replicate design studies (TRT-RTR) in 36 subjects; GMR restriction 0.80–1.25. (a) CV=35%, (b) CV=45%, (c) CV=55%.

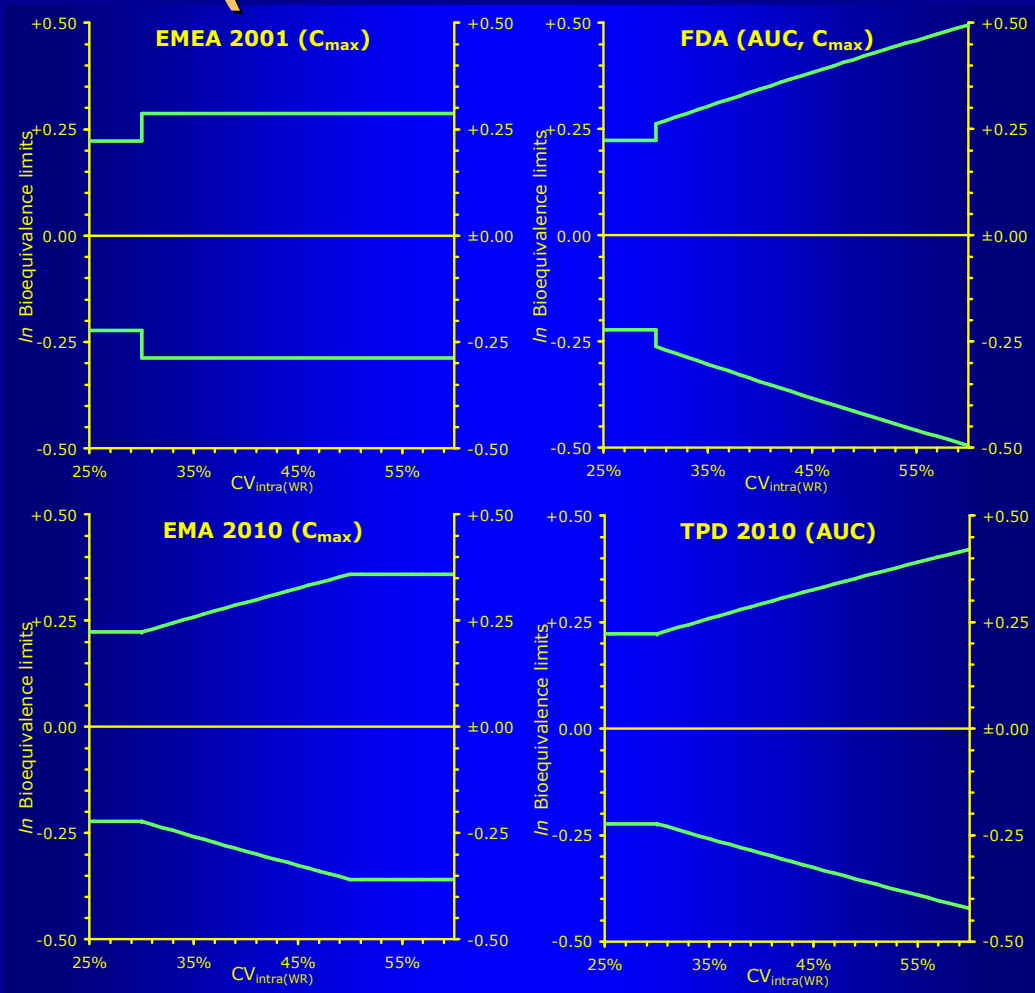
ABE: Conventional Average Bioequivalence, SABE: Scaled Average Bioequivalence, 0.76: EU criterion, 0.89: FDA criterion.

HVDPPs (US/EU)

- FDA's and EMA's approaches differ; FDA's leads to a discontinuity of the acceptance range at CV=30%, because FDA's scaling CV is 25.83% (σ_{WR} 0.294) – but to be *applied* at $CV \geq 30\%$.



HVDPs (No Global Harmonization!)



HVDs/HVDPs (Reg. models)

- Common to FDA and EMA

ABE model

$$-\theta_A \leq \mu_T - \mu_R \leq +\theta_A$$

SABE model

$$-\theta_S \leq \frac{\mu_T - \mu_R}{\sigma_W} \leq +\theta_S$$

Regulatory regulatory switching condition θ_S is derived from the regulatory standardized variation σ_0 (proportionality between acceptance limits in ln-scale and σ_W in the highly variable region).

Tóthfalusi *et al.* (2009)

HVDs/HVDPs (Reg. models)

- Differences between FDA and EMA

FDA: Regulatory regulatory switching condition θ_s is set to 0.893, which would translate into

$$CV_{WR} = 100 \sqrt{e^{\left(\frac{\ln(1.25)}{0.893}\right)^2} - 1} \approx 25.83\%$$

RSABE is allowed only if $CV_{WR} \geq 30\%$ ($s_{WR} \geq 0.294$), which explains to the discontinuity at 30%.

HVDs/HVDPs (Reg. models)

- Differences between FDA and EMA

EMA: Regulatory regulatory switching condition θ_S avoids the discontinuity.

$$CV_W = 0.30$$

$$\sigma_0 = \sqrt{\ln(CV_W^2 + 1)} = 0.2935603792085 \dots$$

$$\theta_S = \frac{\ln(1.25)}{\sigma_0} = -\frac{\ln(0.80)}{\sigma_0} \approx 0.760$$

HVDs/HVDPs (FDA)

- Haidar *et al.* (2008), progesterone guid. (2010)

Starting from the SABE model

$$-\theta_S \leq \frac{\mu_T - \mu_R}{\sigma_W} \leq +\theta_S$$

Rearrangement leads to a linear form

$$(\mu_T - \mu_R)^2 - \theta_S^2 \cdot \sigma_W^2 \leq 0$$

Since we don't have the true parameters, we use estimates

$$E_m = (\mu_T - \mu_R)^2$$

$$E_s = \theta_S^2 \cdot \sigma_W^2$$

HVDs/HVDPs (FDA)

- Haidar *et al.* (2008), progesterone guid. (2010)
Distributions of E_m and E_s are known and their upper confidence limits can be calculated

$$C_m = \left(|m_T - m_R| + t_{\alpha, N-S} \cdot SE \right)^2$$

$$C_s = \frac{\theta_S^2 \cdot (N - S) \cdot s_W^2}{\chi_{\alpha, N-S}^2}$$

t and χ^2 are the inverse cumulative distribution functions at $\alpha 0.05$ and $N - S$ degrees of freedom (N subjects, S sequences). SE is the standard error of the difference between means.

HVDs/HVDPs (FDA)

- Haidar *et al.* (2008), progesterone guid. (2010)
Howe method gets the CI from individual CIs

$$L_m = (C_m - E_m)^2$$

$$L_s = (C_s - E_s)^2$$

$$CI = E_m - E_s + \sqrt{L_m + L_s}$$

The CI of the rearranged SABE criterion ([slide 20](#)) is evaluated at the 95% level. If the upper 95% is positive RSABE is rejected, and accepted otherwise.

HVDs/HVDPs (EMA)

- EU GL on BE (2010)
 - Average Bioequivalence (ABE) with Expanding Limits (ABEL)
 - The regulatory switching condition θ_S at CV_{WR} 30% would be 0.7601228297680...
 - According to the GL (2010) and the Q&A document (2011) use k ($\equiv \theta_S$) with 0.760 (not the exact value).

HVDs/HVDPs (EMA)

- EU GL on BE (2010)

- If you have σ_{WR} (the *intra*-subject standard deviation of the reference formulation) go to the next step; if not, calculate it from CV_{WR}

$$\sigma_{WR} = \sqrt{\ln(CV_{WR}^2 + 1)}$$

- Calculate the scaled acceptance range based on the regulatory constant k ($\theta_s=0.760$)

$$[L, U] = e^{\mp k \cdot \sigma_{WR}}$$

HVDs/HVDPs (EMA)

- Q&A document (March 2011)
 - Two methods proposed (Method A preferred)
 - **Method A:** All effects fixed; assumes equal variances of test and reference, and no subject-by-formulation interaction; only a common within (*intra-*) subject variance is estimated
 - **Method B:** Similar to A, but random effects for subjects. Common within (*intra-*) subject variance and between (*inter-*) subject variance are estimated.
 - **Outliers:** Boxplots (of model residuals?) suggested.

*Questions & Answers on the Revised EMA Bioequivalence Guideline
Summary of the discussions held at the 3rd EGA Symposium on Bioequivalence
June 2010, London
http://www.egagenerics.com/doc/EGA_BEQ_Q&A_WEB_QA_1_32.pdf*

HVDs/HVDPs (EMA)

- At higher CVs only the GMR is of importance!
- At CVs $> 50\%$ still large sample sizes required.
- No commercial software for sample size estimation can handle the GMR restriction.
- Recently sample size tables were published.
- Expect a solution from the  community soon...

L Tóthfalusi and L Endrenyi

Sample Sizes for Designing Bioequivalence Studies for Highly Variable Drugs

J Pharm Pharmaceut Sci 15(1), 73 – 84 (2011)

<http://ejournals.library.ualberta.ca/index.php/JPPS/article/download/11612/9489>

Example datasets (EMA)

- Q&A document (March 2011)
 - Data set I
 - RTRT | TRTR full replicate, 77 subjects, imbalanced, incomplete
 - FDA
 - $s_{WR} 0.446 \geq 0.294 \rightarrow$ apply RSABE ($CV_{WR} 46.96\%$)
 - a. critbound $-0.0921 \leq 0$ and
 - b. $80.00\% \leq$ pointest $115.46\% \leq 125.00\%$ ✓
 - EMA
 - $CV_{WR} 46.96\% \rightarrow$ apply RSABE ($> 30\%$)
 - Scaled Acceptance Range: $71.23\% - 140.40\%$
 - A: $71.23\% \leq 107.11\% - 124.89\% \leq 140.40\%$, PE 115.66% ✓
 - B: $71.23\% \leq 107.17\% - 124.97\% \leq 140.40\%$, PE 115.73% ✓

Example datasets (EMA)

- Q&A document (March 2011)

- Data set II

TRR | RTR | RRT partial replicate, 24 subjects, balanced, complete

- FDA

$s_{WR} 0.114 < 0.294 \rightarrow$ apply ABE ($CV_{WR} 11.43\%$)

$80.00\% \leq 97.05 - 107.76 \leq 125.00\%$ ($CV_{intra} 11.55\%$) ✓

- EMA

- $CV_{WR} 11.17\% \rightarrow$ apply ABE ($\leq 30\%$)

- A: 90% CI 97.32% – 107.46%, PE 102.26% ✓

- B: 90% CI 97.32% – 107.46%, PE 102.26% ✓

- A/B: $CV_{intra} 11.86\%$

Outliers (EMA)

- EU GL on BE (2010), Section 4.1.10
 - The applicant should justify that the calculated intra-subject variability is a reliable estimate and that it is not the result of outliers.
- EGA/EMA Q&A (2010)
 - Q: How should a company proceed if outlier values are observed for the reference product in a replicate design study for a Highly Variable Drug Product (HVDP)?

Outliers (EMA)

- EGA/EMA Q&A (2010)
 - A: The outlier cannot be removed from evaluation [...] but should not be taken into account for calculation of within-subject variability and extension of the acceptance range.
An outlier test is not an expectation of the medicines agencies but outliers could be shown by a box plot. This would allow the medicines agencies to compare the data between them.

Outliers (EMA)

- Data set I (full replicate)

- CV_{WR} 46.96%

- ABEL 71.23% – 140.40%

- Method A: 107.11% – 124.89%

- Method B: 107.17% – 124.97%

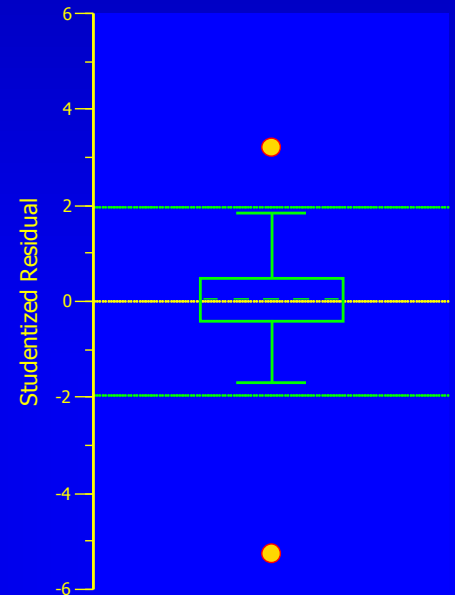
- But there are two outliers!

- Excluding subjects 45 and 52

- CV_{WR} drops to 32.16%.

- ABEL 78.79% – 126.93%

- Almost no more gain compared to conventional limits.



Thank You!

Statistical aspects of reference-scaled studies

Open Questions?

(EMA's and FDA's SAS code and EMA's datasets in the handouts)

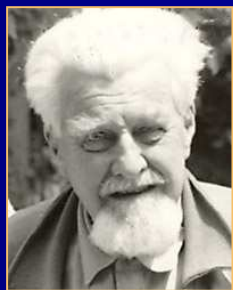


Helmut Schütz
BEBAC

Consultancy Services for
Bioequivalence and Bioavailability Studies
1070 Vienna, Austria
helmut.schuetz@bebac.at

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*



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

It keeps him young.

Konrad Lorenz

If you shut your door to all errors truth will be shut out.

Rabindranath Tagore



SAS code (EMA)

Method A

```
proc glm data=replicate;
  class formulation subject period sequence;
  model logDATA= sequence subject(sequence) period formulation;
  estimate "test-ref" formulation -1+1;
  test h=sequence e=subject(sequence);
  lsmeans formulation / adjust=t pdiff=control("R") CL alpha=0.10;
run;
```

Method B

```
proc mixed data=replicate;
  class formulation subject period sequence;
  model logDATA= sequence period formulation;
  random subject(sequence);
  estimate "test-ref" formulation -1 1 / CL alpha=0.10;
run;
```

CV_{WR} (both methods)

```
data var;
  set replicate;
  if formulation='R';
run;
proc glm data=var;
  class subject period sequence;
  model logDATA= sequence subject(sequence) period;
run;
```

SAS code (FDA)

Partial reference-replicated 3-way design

```
data test;
  set pk;
  if trt='T';
  latt=lauct;
run;

data ref1;
  set ref;
  if (seq=1 and per=2) or (seq=2 and per=1) or (seq=3 and per=1);
  lat1r=lauct;
run;

data ref2;
  set ref;
  if (seq=1 and per=3) or (seq=2 and per=3) or (seq=3 and per=2);
  lat2r=lauct;
run;

data ref2;
  set ref;
  if (seq=1 and per=3) or (seq=2 and per=3) or (seq=3 and per=2);
  lat2r=lauct;
run;
```

SAS code (FDA)

Partial reference-replicated 3-way design (cont'd)

```
proc glm data=scavbe;
  class seq;
  model llat=seq/clparm alpha=0.1;
  estimate 'average' intercept 1 seq 0.3333333333 0.3333333333 0.3333333333;
  ods output overallanova=iglm1;
  ods output Estimates=iglm2;
  ods output NObs=iglm3;
  title1 'scaled average BE';
run;
```

```
pointest=exp(estimate);
x=estimate**2-stderr**2;
boundx=(max((abs(LowerCL)), (abs(UpperCL))))**2;
```

```
proc glm data=scavbe;
  class seq;
  model dlat=seq;
  ods output overallanova=dglm1;
  ods output NObs=dglm3;
  title1 'scaled average BE';
run;
```

```
dfd=df;
s2wr=ms/2;
```

SAS code (FDA)

Partial reference-replicated 3-way design (cont'd)

```

theta=((log(1.25))/0.25)**2;
y=-theta*s2wr;
→ boundy=y*dfd/cinv(0.95,dfd);
SWR=sqrt(s2wr);
critbound=(x+y)+sqrt(((boundx-x)**2)+((boundy-y)**2));

```

Apply RSABE if $SWR \geq 0.294$

RSABE if

a. $critbound \leq 0$ and

b. $0.8000 \leq pointest \leq 1.2500$

If $SWR < 0.294$, apply conventional (unscaled ABE), mixed effects model.

ABE if 90% CI within 0.8000 and 1.2500.

SAS code (FDA)

Fully replicated 4-way design

```
data test1;  
  set test;  
  if (seq=1 and per=1) or (seq=2 and per=2);  
  lat1t=lauct;  
run;
```

```
data test2;  
  set test;  
  if (seq=1 and per=3) or (seq=2 and per=4);  
  lat2t=lauct;  
run;
```

```
data ref1;  
  set ref;  
  if (seq=1 and per=2) or (seq=2 and per=1);  
  lat1r=lauct;  
run;
```

```
data ref2;  
  set ref;  
  if (seq=1 and per=4) or (seq=2 and per=3);  
  lat2r=lauct;  
run;
```

SAS code (FDA)

Fully replicated 4-way design (cont'd)

```
data scavbe;
  merge test1 test2 ref1 ref2;
  by seq subj;
  → ilat=0.5*(lat1t+lat2t-lat1r-lat2r);
  dlat=lat1r-lat2r;
run;

proc mixed data=scavbe;
  class seq;
  model ilat =seq/ddfm=satterth;
  estimate 'average' intercept 1 seq 0.5 0.5/e c1 alpha=0.1;
  ods output CovParms=iout1;
  ods output Estimates=iout2;
  ods output NObs=iout3;
  title1 'scaled average BE';
  title2 'intermediate analysis - ilat, mixed';
run;

pointest=exp(estimate);
x=estimate**2-stderr**2;
boundx=(max((abs(lower)),(abs(upper))))**2;
```

SAS code (FDA)

Fully replicated 4-way design (cont'd)

```
proc mixed data=scavbe;  
  class seq;  
  model dlat=seq/ddfm=satterth;  
  estimate 'average' intercept 1 seq 0.5 0.5/e c1 alpha=0.1;  
  ods output CovParms=dout1;  
  ods output Estimates=dout2;  
  ods output NObs=dout3;  
  title1 'scaled average BE';  
  title2 'intermediate analysis - dlat, mixed';
```

```
run;
```

```
s2wr=estimate/2;  
dfd=df;
```

```
theta=((log(1.25))/0.25)**2;  
y=-theta*s2wr;  
boundy=y*dfd/cinv(0.95,dfd);  
SWR=sqrt(s2wr);
```

```
→ critbound=(x+y)+sqrt(((boundx-x)**2)+((boundy-y)**2));
```

SAS code (FDA)

Unscaled 90% BE confidence intervals (applicable if critbound>0)

PROC MIXED

```
data=pk;  
CLASSES SEQ SUBJ PER TRT;  
MODEL LAUCT = SEQ PER TRT/ DDFM=SATTERTH;  
RANDOM TRT/TYPE=FA0(2) SUB=SUBJ G;  
REPEATED/GRP=TRT SUB=SUBJ;  
ESTIMATE 'T vs. R' TRT 1 -1/CL ALPHA=0.1;  
ods output Estimates=unsc1;  
title1 'unscaled BE 90% CI - guidance version';  
title2 'AUct';
```

run;

```
data unsc1;  
set unsc1;  
unscabe_lower=exp(lower);  
unscabe_upper=exp(upper);
```

run;

Note: Lines marked with an arrow are missing in FDA's code!

Example datasets (EMA)

- Q&A document (March 2011)
 - Data set I
4-period 2-sequence (RTRT | TRTR) full replicate, imbalanced (77 subjects), incomplete (missing periods: two periods in two cases, one period in six cases).
 - Data set II
3-period 3-sequence (TRR | RTR | RRT) partial replicate, balanced (24 subjects), complete (all periods).
 - Download in Excel 2000 format:
http://bebac.at/downloads/Validation_Replicate_Design_EMA.xls