







Study Designs

The more 'sophisticated' a design is, the more information can be extracted.

Hierarchy of designs:

```
Full replicate (RTRT | TRTR or RTR | TRT) →
Partial replicate (RRT | RTR | TRR) →
2×2×2 crossover (RT | TR) →
Parallel (R | T)
```

Variances which can be estimated:

Parallel: total variance (pooled of between + within subjects)

2×2×2 crossover: + between, within subjects *∲*

Partial replicate: + within subjects (of R)

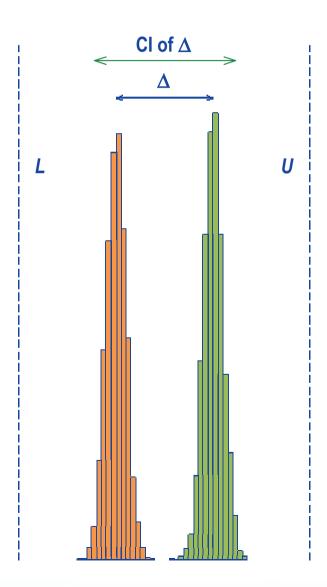
→

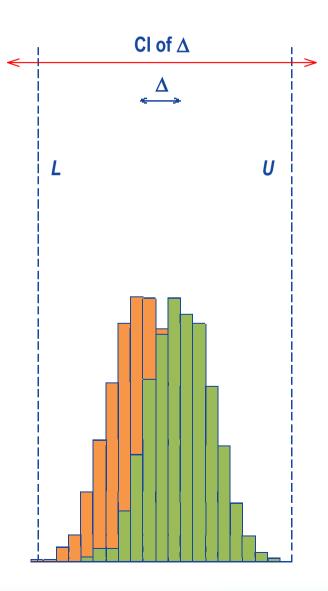
Full replicate: + within subjects (of R and T) *→*





Highly Variable Drugs / Drug Products





Counterintuitive concept of BE:

Two formulations with a large difference in means are declared bioequivalent if variances are low, but not BE – even if the difference is quite small – due to high variability.

Modified from Tothfálusi et al. (2009), Fig. 1



It may be almost impossible to demonstrate ABE with a reasonable sample size.

- Reference-scaling (*i.e.*, widening the acceptance range based of the variability of the reference) in 2010 introduced by the FDA and EMA and in 2016 by Health Canada.
 - Requires a replicate design, where at least the reference product is administered twice.
 - Smaller sample sizes compared to a standard 2×2×2 design but outweighed by increased number of periods.
 - Similar total number of individual treatments.
 - Any replicate design can be evaluated for 'classical' (unscaled) Average Bioequivalence (ABE) as well. Switching CV_{wR} 30%:
 - FDA: AUC and C_{max}
 - EMA: C_{max} ; MR products additionally: $C_{ss,min}$, $C_{ss,r}$, partial AUCs
 - Health Canada: AUC





Models (in log-scale).

- ABE Model:
 - A difference \triangle of ≤20% is considered to be clinically not relevant.
 - The limits [L, U] of the acceptance range are fixed to $log(1-\Delta) = log((1-\Delta)^{-1})$ or $L \sim -0.2231$ and $U \sim +0.2231$.
 - The consumer risk (α) is fixed with 0.05. BE is concluded if the 100(1 2α) confidence interval lies entirely within the acceptance range.

$$-\theta_{A} \leq \mu_{T} - \mu_{R} \leq +\theta_{A}$$

- SABEL Model:
 - Switching condition $\theta_{\rm S}$ is derived from the regulatory standardized variation $\sigma_{\rm 0}$ (proportionality between acceptance limits in log-scale and $\sigma_{\rm WR}$ in the highly variable region).

$$-\theta_{S} \leq \frac{\mu_{T} - \mu_{R}}{\sigma_{WR}} \leq +\theta_{S}$$

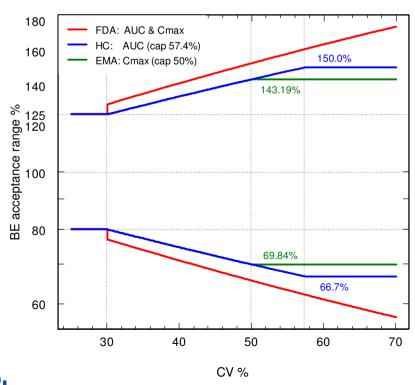


Regulatory Approaches.

• Bioequivalence limits derived from $\sigma_{\!_{0}}$ and $\sigma_{\!_{wR}}$

$$\theta_{S} = \frac{\log(1.25)}{\sigma_{0}}, [L,U] = e^{\pm\theta_{S}\cdot\sigma_{WR}}$$

- FDA
 - Scaling σ_{wR} 0.25 (θ_{S} 0.893) but applicable at $CV_{wR} \ge 30\%$.
 - Discontinuity at CV_{wR} 30%.
- EMA
 - Scaling σ_0 0.2936 (θ_S 0.760).
 - Upper cap at CV_{wR} 50%.
- Health Canada
 - Like EMA but upper cap at CV_{wR} 57.4%.





The EMA's Approach.

- Average Bioequivalence with Expanding Limits ABEL (crippled from Endrényi and Tóthfalusi 2009).
 - Justification that the widened acceptance range is clinically not relevant (important – different to the FDA).
 - Assumes identical variances of T and R [sic] like in a $2\times2\times2$.
 - All fixed effects model according to the Q&A-document preferred.
 - Mixed-effects model (allowing for unequival variances) is 'not compatible with CHMP guideline'...
 - Scaling limited at a maximum of CV_{wR} 50% (i.e., to 69.84 143.19%).
 - GMR within 0.8000 − 1.2500.
 - Demonstration that $CV_{wR} > 30\%$ is not caused by outliers (box plots of studentized intra-subject residuals?)...
 - — ≥12 subjects in sequence RTR of the 3-period full replicate design.

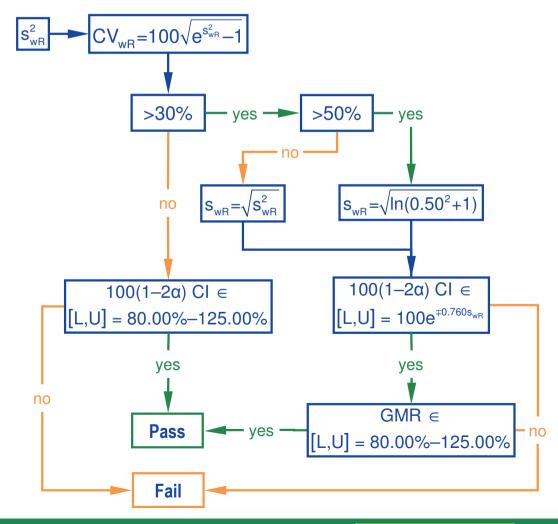




The EMA's Approach.

- Decision Scheme.
 - The Null Hypothesis is specified in the face of the data.
 - Acceptance limits themselves become random variables.
 - Type I Error (consumer risk) might be inflated.







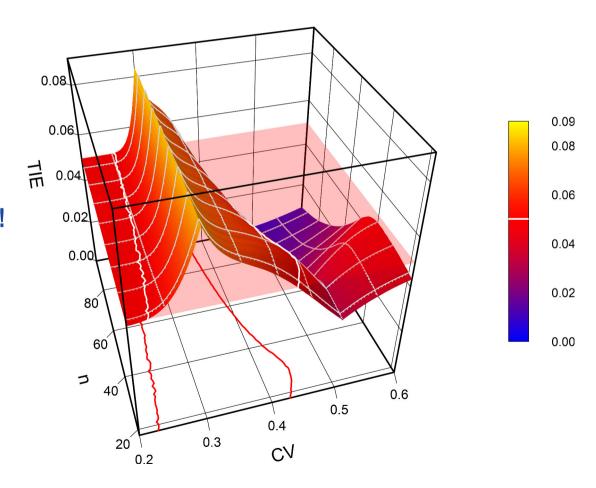
Assessing the Type I Error (TIE).

- TIE = falsely concluding BE at the limits of the acceptance range.
- Due to the decision scheme direct calculation of the TIE at the scaled limits is not possible;
 - \rightarrow extensive simulations required (10⁶ BE studies mandatory).
- Inflation of the TIE suspected. (Chow et al. 2002, Willavazie and Morgenthien 2006, Chow and Liu 2009, Patterson and Jones 2012).
- Confirmed,
 - EMA's ABEL
 (Tóthfalusi and Endrényi 2009, BEBA-Forum 2013, Wonnemann et al. 2015, Muñoz et al. 2016, Labes and Schütz 2016).
 - FDA's RSABE
 (Tóthfalusi and Endrényi 2009, BEBA-Forum 2013, Muñoz et al. 2016).



Example for ABEL

- RTRT | TRTR
 sample size 18 96
 CV_{wR} 20% 60%
 - TIE_{max} 0.0837.
 - Relative increase of the consumer risk 67%!







What is going on here?

SABE is stated in model parameters ...

$$-\theta_{S} \leq \frac{\mu_{T} - \mu_{R}}{\sigma_{WR}} \leq +\theta_{S}$$

- ... which are unknown.
- Only their estimates (GMR, s_{wR}) are accessible in the actual study.
- At CV_{wR} 30% the decision to scale will be wrong in ~50% of cases.
- If moving away from 30% the chances of a wrong decision decrease and hence, the TIE.
- At high CVs (>43%) both the scaling cap and the GMR-restriction help to maintain the TIE <0.05).



Outlook.

Utopia

— Agencies collect CV_{wR} from submitted studies. Pool them, adjust for designs / degrees of freedom. The EMA publishes a fixed acceptance range in the product-specific guidance. No need for replicate studies any more. $2\times2\times2$ crossovers evaluated by ABE would be sufficient.

Halfbaked

- Hope [sic] that e.g., Bonferroni preserves the consumer risk. Still apply ABEL, but with a 95% CI (α 0.025).
- Drawback: Loss of power, substantial increase in sample sizes.

Proposal

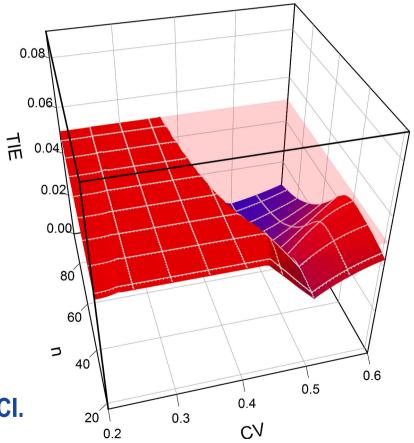
— Iteratively adjust α based on the study's CV_{wR} and sample size – in such a way that the consumer risk is preserved (Labes and Schütz 2016).





Previous example

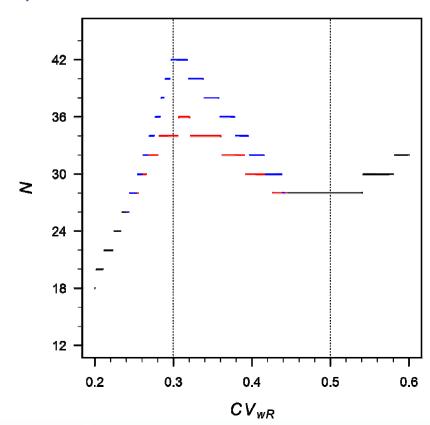
- Algorithm
 - Assess the TIE for the nominal α 0.05.
 - If the TIE \leq 0.05, stop.
 - Otherwise adjust α (downwards) until the TIE \cong 0.05.
 - At $\textit{CV}_\textit{wR}$ 30% (dependent on the sample size) $\alpha_\textit{adj}$ is 0.0273 0.0300;
 - → use a 94.00 94.54% CI.





Potential impact on the sample size.

- Example: RTRT | TRTR, θ_0 0.90, target power 0.80.
 - Moderate in the critical region (— —).
 - CV_{WR} 30%: 36 \rightarrow 42 (+17%);
 - CV_{WR} 35%: 34 \rightarrow 38 (+12%);
 - CV_{WR} 40%: 30 \rightarrow 32 (+7%).
 - None outside (—).





Example (RTRT | TRTR, expected CV_{wR} 35%, θ_0 0.90, target power 0.80); R package PowerTOST (\geq 1.3-3).

Estimate the sample size.

Estimate the empiric TIE for this study.

```
UL <- scabel(CV=0.35)[["upper"]] # scaled limit (1.2948 for CVwR 0.35) power.scabel(CV=0.35, theta0=UL, n=34, design="2x2x4", nsims=1e6) [1] 0.065566
```

Iteratively adjust α.

```
scabel.ad(CV=0.35, n=34, design="2x2x4")
++++++++ scaled (widened) ABEL ++++++++
        iteratively adjusted alpha
CVWR 0.35, n(i) 17|17 (N 34)
Nominal alpha
                            : 0.05
                     : 0.9000
Null (true) ratio
Regulatory settings : EMA (ABEL)
Empiric TIE for alpha 0.0500 : 0.06557
Power for theta0 0.900
                          : 0.812
Iteratively adjusted alpha : 0.03630
Empiric TIE for adjusted alpha: 0.05000
Power for theta0 0.900
                            : 0.773
```





 Optionally compensate for the loss in power (0.812 → 0.773) by increasing the sample size:

```
sampleN.scABEL.ad(CV=0.35, theta0=0.90, targetpower=0.80, design="2x2x4")
  ++++++++ scaled (widened) ABEL ++++++++
              Sample size estimation
          for iteratively adjusted alpha
  Study design: 2x2x4 (RTRT|TRTR)
  Expected CVwR 0.35
  Nominal alpha : 0.05
  Null (true) ratio : 0.9000
  Target power : 0.8
  Regulatory settings: EMA (ABEL)
  Switching CVwR : 30%
  Regulatory constant: 0.760
  Expanded limits : 0.7723...1.2948
  Upper scaling cap : CVwR 0.5
  PE constraints : 0.8000...1.2500
  n 38, adj. alpha: 0.03610 (power 0.8100), TIE: 0.05000
- n 34 \rightarrow 38 (+12%), power 0.773 \rightarrow 0.810, \alpha_{adi} 0.0363 \rightarrow 0.0361.
```

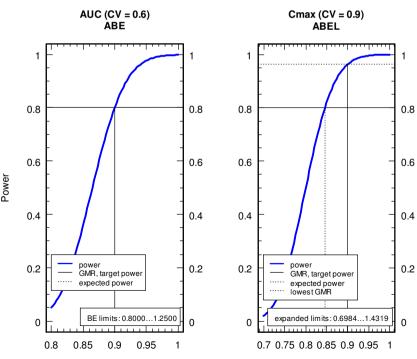


Side Effect

Allowing ABEL only for C_{max} .

- Some drugs show high variability in AUC as well.
 - Since in such a case the sample size is mandated by AUC, products with high deviations in C_{max} will be approved.
 - Example: CV_{wR} 90% (C_{max}), 60% (AUC), θ_0 0.90, target power 80% \rightarrow the study is 'overpowered' for C_{max} ; C_{max} -GMRs of [0.846 1.183] will pass BE. Really desirable?
 - With the FDA's RSABE the study could be performed in only 34 subjects...

ABEL (EMA): design RTRT|TRTR, target power = 0.8, n = 138 (sample size dependent on AUC)





GMR

GMR



Reference-scaled Average Bioequivalence

Thank You! Open Questions?



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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





100% of all disasters are failures of design, not analysis.

Ronald G. Marks

My definition of an expert in any field is a person who knows enough about what's really going on to be scared.



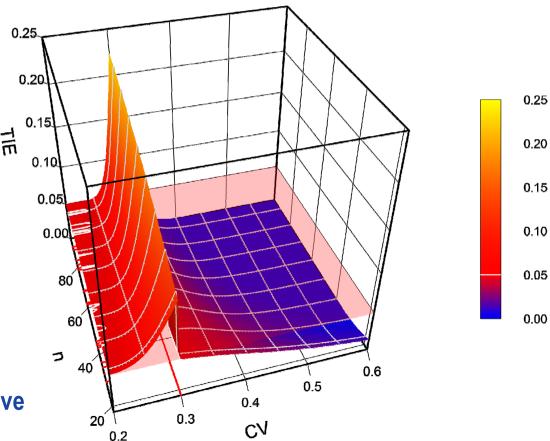
Phillip J. Plauger



Backup

Example for the FDA's RSABE

- RTRT | TRTR
 sample size 18 96
 CV_{wR} 20% 60%
 - TIE_{max} 0.2245.
 - Relative increase of the consumer risk 349%!
 - TIE more dependent on the sample size than in ABEL.
 - However, no inflation of the TIE for $CV_{wR} > 30\%$; RSABE is very conservative for 'true' HVD(P)s.



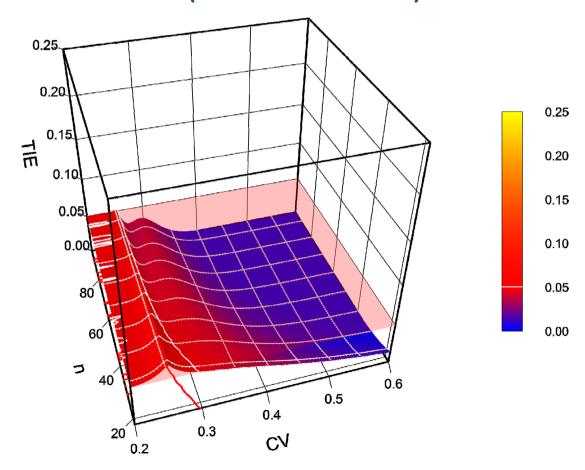




Backup

FDA's desired consumer risk model (Davit et al. 2012)

- Previous example
 - TIE assessed not at the scaled limits but
 - at 1.25 if CV_{wR} ≤25.4%
 - at $e^{0.893 \cdot \sigma_{WR}}$ otherwise.
 - TIE_{max} 0.0668.
 - Lászlo Endrényi: "Hocus pocus!"







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