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Adaptive Treatment Arm Selection in Multivariate Bioequivalence Trials

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## **Acknowledgments / References**



#### Presented theory based on methodological work regarding...

- Intersection-union-tests
  - Berger, R.L. (1982). Multiparameter hypothesis testing and acceptance sampling. *Technometrics* **24**: 295-300
- Inverse normal p-value combination
  - ADDPLAN, Inc., an Aptiv Solutions Company (2014). ADDPLAN Base version 6.1 User Manual. Aptiv Solutions, Cologne Germany.
- Adaptive treatment selection
  - Bretz, F., König, F., Brannath, W., Glimm, E., Posch, M. (2009). Tutorial in Biostatistics: Adaptive designs for confirmatory clinical trials. *Statistics in Medicine* **28**: 1181-1217
- ... and discussions with/within
  - Gernot Wassmer
  - Silke Jörgens
  - Vladimir Dragalin
  - BMBF project MÄQNU (03MS642G).

## **Motivation**





"Two pharmaceutical products are bioequivalent, if they are pharmaceutically equivalent and their bio-availabilities are similar to such a degree that their effects can be expected to be essentially the same"

- Measures of bio-availability:
  - Area under the time concentration curve
  - Maximum drug concentration C<sub>max</sub>

## **Motivation**



### **Establishing Bio-equivalence:**

- 90% confidence intervals on ratios:
  - A and B are called bioequivalent, if 90% CI for
    - $AUC_A/AUC_B$  and
    - $C_{max;A}/C_{max;B}$

are located within the equivalence range [0.8, 1.25]

... alternatively we consider 90% confidence intervals on the logarithms:

- A and B are called bioequivalent, if 90% CI for
  - $\theta_1 := \log(AUC_A) \log(AUC_B)$  and
  - $\theta_2 := \log(C_{max;A}) \log(C_{max;B})$

are located within the equivalence range [log(0.8), log(1.25)]

## **Motivation**



#### **Considered problem:**

• Unknown optimal formulation / dose level.

### Solution:

• Inclusion of several test arms.

### New problem:

• Multiplicty issues: Increase in sample size required?

### **Proposed solution:**

• Interim Analysis with selection of a test arm.

### How to implement this?

Transform the problem to well-known settings.

## 1. Proof of bioequivalence



### What is the "Proof of bioequivalence"?

- A and B are called bioequivalent, if both 90% CIs for
  - $\theta_1 := \log(AUC_A) \log(AUC_B)$  and
  - $\theta_2 := \log(C_{max;A}) \log(C_{max;B})$

are located within the equivalence range [ log(0.8) , log(1.25) ]

With : 
$$\hat{\theta}_i \sim \mathcal{N}\left(\theta_i, \frac{1}{n}\sigma_i^2\right)$$
 confidence intervals are given as:  
 $CI_{0.9} = \left[\widehat{\theta}_i - z_{0.95}\frac{\sigma_i}{\sqrt{n}}, \widehat{\theta}_i + z_{0.95}\frac{\sigma_i}{\sqrt{n}}\right]$ 

Bioequivalence is established, if for both endpoints *i* holds:

$$\log(0.8) \le \widehat{\theta}_i - z_{0.95} \frac{\sigma_i}{\sqrt{n}} \qquad \cap \qquad \widehat{\theta}_i + z_{0.95} \frac{\sigma_i}{\sqrt{n}} \le \log(1.25)$$

# 1. Proof of bioequivalence



What is the "Proof of bioequivalence"? (one endpoint) Bioequivalence is established, if:

$$\log(0.8) \le \widehat{\theta}_i - z_{0.95} \frac{\sigma_i}{\sqrt{n}} \qquad \cap \qquad \widehat{\theta}_i + z_{0.95} \frac{\sigma_i}{\sqrt{n}} \le \log(1.25)$$

The proof of bioequivalence coincides with the rejection of:

 $H_{01}: \theta_i \le \log(0.8)$  and  $H_{02}: \theta_i \ge \log(1.25)$ 

at level  $\alpha = 0.05$ .

The proof of bioequivalence is a test at lpha=0.05 for  $H_0:\ H_{01}\cup H_{02}$  .



## **Group-sequential test for bioequivalence (one endpoint)** Early rejection of "no-bioequivalence", if all tests can be rejected early

e.g. Using Inverse-normal p-value combination

• Stage-wise p-values  $p_{i;1}$  and  $p_{i;2}$  and information fraction  $\omega_{i;j}$ :

$$\min\left(\sum_{i=1}^{j} \omega_{i;j} \Phi^{-1}(1-p_{i;1}), \sum_{i=1}^{j} \omega_{i;j} \Phi^{-1}(1-p_{i;2})\right) > c_j$$

• Alternative combination:

$$\sum_{i=1}^{j} \omega_{i;j} \Phi^{-1}(1 - \max(p_{i;1}, p_{i;2})) > c_j$$

... controls  $\alpha$ , but is conservative.



### Usual multiplicity issues hold:

- Increased probability of having at least one false positive
  - Increased critical values -> Increased required sample size.
- Null hypothesis:
  - *p* equivalence endpoints (*2p* one-sided tests)
  - *k* treatment arms

 $H_0: \cap_{g=1}^k$  No equivalence for treatment  $k = \bigcap_{g=1}^k$ 

- Simplest adjustment: Bonferroni
  - Rejection of "no equivalence" for arm g, if

$$\max_{j=1,\dots,2p} \quad p_{1;j}^g \le \frac{\alpha}{k} \Leftrightarrow \max_{j=1,\dots,2p} \quad kp_{1;j}^g \le \alpha$$





### Decrease multiplicity penalty by dropping treatment arms:

- First stage still requires multiplicity penalty
- Second stage not adjusted for multiplicity.

Adaptive treatment selection (here not with focus on Bioequivalence):



## 2. Adaptive Treatment Arm Selection



### Similar strategy for equivalence hypothesis (one endpoint)

 Rejection of "no-equivalence", if every test for every one-sided hypotheses may be rejected:



 $(H_{01}^1 \cup H_{02}^1) \cap (H_{01}^2 \cup H_{02}^2) = (H_{01}^1 \cap (\cup_{j=1}^2 H_{0;j}^2)) \cup (H_{02}^1 \cap (\cup_{j=1}^2 H_{0;j}^2))$ 

## 2. Adaptive Treatment Arm Selection



#### **Decision rule:**

- General number of endpoints p
- Rejection of  $\cup_{i=1}^{2p} H_{0i}^s$ , if:

$$\min_{i=1,...,2p} \left\{ \omega_1 \Phi^{-1} (1-p_{1;i}^{\mathcal{J}}) + \omega_2 \Phi^{-1} (1-p_{2;i}^s) \right\} > c_{2;\alpha}$$
where  $p_{1;i}^{\mathcal{J}} = \max \left\{ p_{1;i}^s, 2\min\{p_{1;i}^s, \max_{j=1,...,2p}\{p_{1;j}^{\overline{s}}\}\} \right\}$ 

$$H_{0i}^s \cap \left( \cup_{j=1}^{2p} H_{0;j}^{\overline{s}} \right)$$

- Specially for "select the best\*":  $p_{1;i}^{\mathcal{J}} = 2p_{1;i}^{s}, \ i=1,...,2p$ 

\*Best defined as arm with minimum maximum p value



### Target:

- Evaluate A<sub>1</sub> and A<sub>2</sub> for Bioequivalence against reference C
- First stage: each individual receives all arms
- Each individual randomized to one sequence of study arms:



- After treatment period t<sub>1</sub> estimation of AUC and Cmax for the studied arm
- Assumption: No carry-over and sequence effects



#### Model of observations for one treatment arm on one individual:



• Assumption on unobservable effects:

$$\epsilon_{j;i} \sim \mathcal{N}(0, \Sigma) \quad \Sigma = \sigma_1^2 \begin{pmatrix} 1 & \rho \\ \rho & 1 \end{pmatrix}$$

• Assumption on individual effects:

$$b_i \sim \mathcal{N}(0, D) \quad D = \begin{pmatrix} d_1 & 0 \\ 0 & d_2 \end{pmatrix}$$



#### Model of observations for all treatment arms on one individual:

$$\begin{array}{c} \text{AUC and } \mathbf{C}_{\max} \, \text{arm } \mathbf{C} \\ Y_i = \begin{pmatrix} Y_C; i \\ Y_{A_1}; i \\ Y_{A_2}; i \end{pmatrix} \sim \mathcal{N} \left( \begin{pmatrix} \mu_C \\ \mu_{A_1} \\ \mu_{A_2} \end{pmatrix}, S \right) \\ \text{AUC and } \mathbf{C}_{\max} \, \text{arm } \mathbf{A}_2 \end{array}$$

• Variance structure:

• Between treatments: 
$$Cov(Y_{j;i}, Y_{j',i}) = D$$

- Within treatments:  $Var(Y_{j;i}) = \Sigma + D$
- Maximum-Likelihood estimator for the mean parameters:

$$\widehat{\mu} = \frac{1}{n} \sum_{i=1}^{n} Y_i, \ Cov(\widehat{\mu}) = \frac{1}{n} S$$



To be studied: Bioequivalence with reference C:



- Variance structure:
  - Between treatments:
  - Within treatments:

$$Cov(\widehat{\theta}_{j},\widehat{\theta}_{j'}) = \frac{1}{n}\Sigma$$
$$Var(\widehat{\theta}_{j}) = \frac{2}{n}\Sigma$$

• Variance of estimator in dependence on treatment arms and sequence size:

$$Cov(\widehat{\theta}_1) = \frac{1}{n_1 \times 3!} V$$



#### **Treatment arm selection:**

• After interim analysis, only one arm vs. reference



- Possible advantages:
  - Time savings: Second stage savings:  $t_1+t_2$
  - Observation savings due to reduced number of treatment arms
  - Patient savings due to reduced multiplicity
- Variance of estimator in dependence on treatment arms and sequence size:

$$Cov(\widehat{\theta}_2) = \frac{1}{2n_2} 2\Sigma$$



**Considered situations:** 

## 1. Fixed max. number of subjects:

- 1. Dependence of power on stopping rule?
- 2. Dependence of power on standard deviation?
- 3. Dependence of power on interim timing?
- 2. Required patients for target power:
  - 1. Dependence of patient number on standard deviation?
  - 2. Dependence of patient number on interim timing?







#### **Considered situations:**

1. Fixed max. number of subjects: N=60

Dependence of power on the stopping rule:



- > Average across all standard deviation scenarios (0.1 0.5)
- O'Brien&Fleming: Power here similar to "No Stop"



#### **Considered situations:**

1. Fixed max. number of subjects: N=60

Dependence of power on the standard deviation



The later the interim, the higher the power (simple...)

Type-1 error controlled



#### **Considered situations:**

1. Fixed max. number of subjects: N=60

Average number of patients / Probability to select correct arm



- ➢ Given similar power: treatment selection after 24 / 36 promising
- Early selection: High probability to select wrong treatment.



#### **Considered situations:**

1. Fixed max. number of subjects: N=60

Average number of observations



Fixed designs needs much more observations



#### **Considered situations:**

- 1. Fixed max. number of subjects:
  - 1. Dependence of power on stopping rule?
  - 2. Dependence of power on standard deviation?
  - 3. Dependence of power on interim timing?

### 2. Required patients for target power:

- 1. Dependence of patient number on standard deviation?
- 2. Dependence of patient number on interim timing?



#### **Considered situations:**

1. Target power: 80%

#### Required Number of Patients



Number of patients at minimum with no selection

## 4. Summary



### Adaptive treatment arm selection in bioequivalence trials promising...

- Reduction of the required number of observations
- Reduction of study duration
- ... but not always the best option:
- > Number of required patients may be larger than in a fixed trial
- Complex trial design (Switch from 3-way to 2-way crossover)

### Need to take constraints into account, e.g.:

- $\succ$  High costs per patient, low costs per observation  $\longrightarrow$  fixed design
- $\succ$  High cost per observation, low costs per patient  $\longrightarrow$  adaptive design





### Thank you for your attention!