Assessing systemic drug exposure in repeated dose toxicity studies in the case of complete and incomplete sampling

#### Martin J Wolfsegger<sup>1</sup> Thomas Jaki<sup>2</sup>

<sup>1</sup>Baxter Innovations GmbH, Vienna, Austria

<sup>2</sup>Department of Mathematics and Statistics, Lancaster University, Lancaster, United Kingdom

> NCS Conference September 28, 2010

A D N A P N A P N A P

### Outline





#### 3 Simulations



Martin J Wolfsegger Repeated dose toxicity studies in incomplete designs

Dac

э

→ E → < E →</p>

Toxicokinetics Sampling designs

## Introduction

Repeated dose toxicity studies

- contribute to the development of safe medicinal products for human use
- are performed prior application of a compound in clinical trials
- are essential for the design of subsequent studies and for safety assessment in humans

イロト イポト イヨト イヨト

Toxicokinetics Sampling designs

## **Toxicokinetics**

Objective: Does systemic exposure change?

- Increase: damage to the eliminating organs, accumulation, non-linear kinetics, ...
- Decrease: auto-induction of metabolizing enzymes, neutralizing antibody formation to a biological, ...

Commonly used parameter for assessing drug exposure is the area under the concentration versus time curve (AUC).

イロト イポト イヨト イヨト

Toxicokinetics Sampling designs

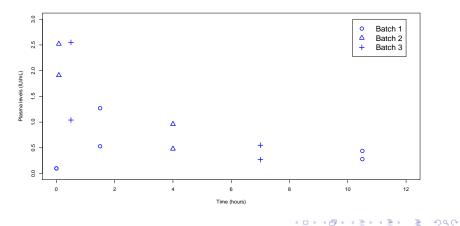
# Sampling designs

- Complete data design: blood samples are available for each animal at all time points
- Batch design: each animal contributes samples at some, but not all time points
- Serial sampling design: each animal contributes exactly one sample

イロト イポト イヨト イヨト

Toxicokinetics Sampling designs

Figure 1: Example of a batch design after first administration



Martin J Wolfsegger Repeated dose toxicity studies in incomplete designs

The model Estimating the AUC Difference of AUCs Ratio of AUCs

### The model

#### Concentration for animal i at time t after administration k:

$$Y_{itk} = \mu_{tk} + \epsilon_{itk},$$

where  $\epsilon_{itk} \sim G_{tk}$  with  $1 \leq j \leq J$ .

イロト イポト イヨト イヨト

3

DQC

The model Estimating the AUC Difference of AUCs Ratio of AUCs



We consider the case where:

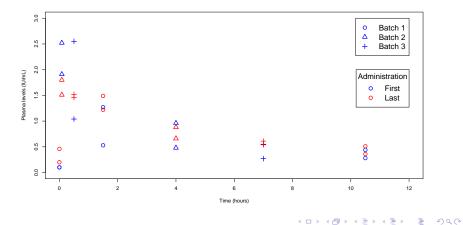
- A specific time point is used in exactly one batch
- Every animal belongs to exactly one batch and has measurements at all time points of this batch
- Animals, time points studied and batches are identical at first and repeated administration

A B A B A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A

. . . . . . .



Figure 2: Example of a batch design after first and last administration



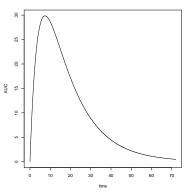
Martin J Wolfsegger Repeated dose toxicity studies in incomplete designs

The model Estimating the AUC Difference of AUCs Ratio of AUCs

### **Theoretical AUC**

The theoretical AUC from 0 to the last observed time point for treatment k is

$$AUC_k = \int_0^{t_{last}} \mu_{tk} dt.$$



A B + A B +
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A

→ Ξ → → Ξ

990

э

The model Estimating the AUC Difference of AUCs Ratio of AUCs

# Using the linear trapezoidal rule

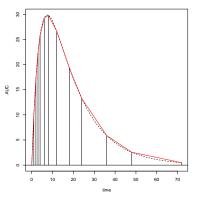
$$AUC_k = \sum_{j=1}^J w_j \mu_{t_j k}$$

The weights,  $w_j$ , equal

$$w_1 = \frac{1}{2} (t_2 - t_1)$$
  

$$w_j = \frac{1}{2} (t_{j+1} - t_{j-1})$$
  

$$w_J = \frac{1}{2} (t_J - t_{J-1})$$



The model Estimating the AUC Difference of AUCs Ratio of AUCs

## Using the linear trapezoidal rule

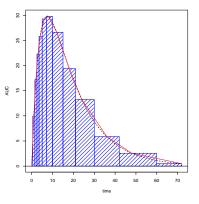
$$AUC_k = \sum_{j=1}^J w_j \mu_{t_j k}$$

The weights,  $w_j$ , equal

$$w_1 = \frac{1}{2} (t_2 - t_1)$$
  

$$w_j = \frac{1}{2} (t_{j+1} - t_{j-1})$$
  

$$w_J = \frac{1}{2} (t_J - t_{J-1})$$



< • • • • •

590

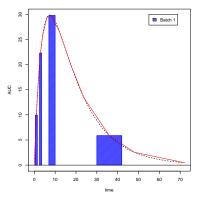
The model Estimating the AUC Difference of AUCs Ratio of AUCs

# Estimating the AUC

- *B* batches with *n<sub>b</sub>* animals
- *J<sub>b</sub>* are indices of time points investigated in batch *b*

Partial AUC of animal i of batch b:

$$\hat{A}_{b_ik} = \sum_{j \in J_b} w_j Y_{it_jk}$$



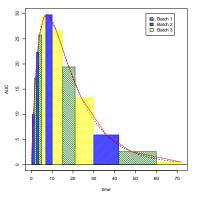
The model Estimating the AUC Difference of AUCs Ratio of AUCs

# Estimating the AUC

- *B* batches with *n<sub>b</sub>* animals
- *J<sub>b</sub>* are indices of time points investigated in batch *b*

The total AUC can be estimated:

$$\widehat{AUC}_k = \sum_{b=1}^B \frac{1}{n_b} \sum_{i=1}^{n_b} \hat{A}_{b_i k}$$



The model Estimating the AUC Difference of AUCs Ratio of AUCs

## Estimating the AUC

Jaki and Wolfsegger (2009) showed that for equal sample size in each batch,  $n_b = n$ ,  $\widehat{AUC}_k \xrightarrow{d} N\left(AUC_k, \xi_k^2\right)$  with

$$\xi_k^2 = \frac{1}{n} \sum_{b=1}^B \sum_{j \in J_b} \sum_{l \in J_b} w_j w_l \sigma_{t_j t_l k}$$

which can be estimated by (Yeh 1990; Holder et al. 1999)

$$\hat{\xi}_k^2 = \sum_{b=1}^B \frac{1}{n_b(n_b - 1)} \sum_{i=1}^{n_b} \left( \hat{A}_{b_i k} - \frac{1}{n_b} \sum_{l=1}^{n_b} \hat{A}_{b_i k} \right)^2$$

イロト 不得 とくほと くほとう

= 990

The model Estimating the AUC Difference of AUCs Ratio of AUCs

Confidence interval for the difference between AUCs

Using the paired structure of the experiment:

$$\hat{\delta} = \sum_{b=1}^{B} \frac{1}{n_b} \sum_{i=1}^{n_b} \hat{A}_{b_i k} - \sum_{b=1}^{B} \frac{1}{n_b} \sum_{i=1}^{n_b} \hat{A}_{b_i 1}$$
$$= \sum_{b=1}^{B} \frac{1}{n_b} \sum_{i=1}^{n_b} \hat{\delta}_{b_i}$$

where  $\hat{\delta}_{b_i} = \hat{A}_{b_ik} - \hat{A}_{b_i1}$ .

イロト イポト イヨト イヨト

DQC

1

The model Estimating the AUC Difference of AUCs Ratio of AUCs

Confidence interval for the difference between AUCs

$$\begin{split} E(\hat{\delta}) &= AUC_k - AUC_1 \\ V(\hat{\delta}) &= \vartheta^2 = \xi_k^2 + \xi_1^2 - 2\xi_{1k} \\ \widehat{V}(\hat{\delta}) &= \vartheta^2 = \sum_{b=1}^B \frac{1}{n_b(n_b - 1)} \sum_{i=1}^{n_b} \left( \hat{\delta}_{b_i} - \frac{1}{n_b} \sum_{l=1}^{n_b} \hat{\delta}_{b_l} \right)^2 \\ \widehat{\xi}_{1k} &= \frac{1}{2} (\hat{\xi}_k^2 + \hat{\xi}_1^2 - \vartheta^2) \\ \widehat{\delta} & \stackrel{d}{\to} N \left( AUC_k - AUC_1, \vartheta^2 \right) \end{split}$$

・ロト ・ 同ト ・ ヨト ・ ヨト … ヨ

Introduction The n Estimating exposure Estim Simulations Differ Conclusion Ratio

The model Estimating the AUC Difference of AUCs Ratio of AUCs

## Confidence interval for the ratio of AUCs

#### A $1 - \alpha$ Fieller-type confidence interval (CI) for the ratio is:

$$\begin{split} \left[ \hat{\Delta} + \left( \frac{c}{1-c} \right) \left( \hat{\Delta} - \frac{\xi_{k1}}{\xi_1^2} \right) \pm \\ & \frac{t_{1-\frac{\alpha}{2},\nu}}{\widehat{AUC}_1(1-c)} \sqrt{\xi_k^2 - 2\hat{\Delta}\xi_{k1} + \hat{\Delta}^2 \xi_1^2 - c\left(\xi_k^2 - \frac{\xi_{k1}^2}{\xi_1^2}\right)} \right] \\ \end{split}$$
where  $c = \xi_1^2 \widehat{AUC}_1^{-2} t_{1-\frac{\alpha}{2},\nu}^2$ 

イロト イポト イヨト イヨト 一臣

DQC

Methods compared Data generation Results

## Methods compared

- Fieller approach using covariance between AUCs (Fieller-dep-t) based on t-distribution
- Fieller approach assuming independent AUCs based on a t-distribution (Fieller-ind-t)
- Asymptotic Fieller interval based on normal distribution (Fieller-dep-z)
- Asymptotic Wald-type interval based on the asymptotic normal distribution of the ratio (Asymptotic)

イロト イポト イヨト イヨト

SAR

Methods compared Data generation Results

# Data generation

- One-compartmental model after IV bolus administration
- 3 batches with 3 time points with n = 3 and 6
- Within-batch correlations: dependence of measurements between time points per animal
- Between-dosing correlations: dependence of measurements between first and last dosing per animal
- CV of Y<sub>itjk</sub> was set to 20% for all time points greater zero and follow a log-normal distribution

イロト イポト イヨト イヨ

Introduction Estimating exposure Simulations Conclusion	Methods compared Data generation Results
--	--

Table 1: Empirical coverage for log-normally distributed drug levels with identical coefficient of variation of 20% for  $\Delta=1$  and using a nominal coverage of 90%

Intra-animal correlation			Method			
within-batch	between-dosing	Ν	Fieller-dep-t	Fieller-ind-t	Fieller-dep-z	Asymptotic
0 0	0	3	0.9227	0.9159	0.8497	0.8479
		6	0.9051	0.9065	0.8794	0.8786
0.3	0	3	0.9233	0.9152	0.8489	0.8468
		6	0.9045	0.9057	0.8786	0.8775
	0.3	3	0.9215	0.9698	0.8473	0.8446
		6	0.9064	0.9716	0.8801	0.8783
0.6	0.3	3	0.9239	0.9617	0.8480	0.8445
		6	0.9048	0.9596	0.8791	0.8769
	0.6	3	0.9243	0.9936	0.8495	0.8457
		6	0.9050	0.9969	0.8790	0.8766
0.9	0.3	3	0.9226	0.9544	0.8483	0.8455
		6	0.9041	0.9530	0.8778	0.8756
	0.6	3	0.9212	0.9868	0.8458	0.8421
		6	0.9045	0.9903	0.8773	0.8751
	0.9	3	0.9241	0.9999	0.8501	0.8448
		6	0.9047	1.0000	0.8788	0.8761

N per batch

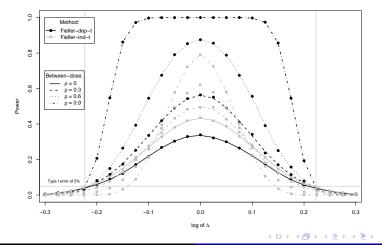
イロト イポト イヨト イヨト

э.

590



Figure 3: Empirical power curves for a within-batch  $\rho = 0.9$  and N=3 per batch



Martin J Wolfsegger Repeated dose toxicity studies in incomplete designs

990

э

Summary References

# Summary

- Maintains nominal coverage and was slightly conservative for small sample sizes per batch
- Uniformly superior in power to the corresponding interval not incorporating the dependence between the AUCs as long as a correlation exists
- Measure for variability while additionally allowing to formally test for difference as well as for equivalence
- Framework that can be applied to a wide range of investigations where changes between pre- and post values are of interest

A B A B A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A
 A

- A 🗉 🕨

Summary References

### References

HOLDER, DJ, HSUAN, F, DIXIT, R & SOPER, K (1999). A method for estimating and testing area under the curve in serial sacrifice, batch, and complete data designs. *Journal of Biopharmaceutical Statistics* **9**, 451–464.

JAKI, T & WOLFSEGGER, MJ (2009). A theoretical framework for estimation of AUCs in complete and incomplete sampling designs. *Statistics in Biopharmaceutical Research* 1(2), 176–184.

JAKI, T & WOLFSEGGER, MJ (2010). PK: Basic Non-Compartmental Pharmacokinetics. *R-package Version 1.2-3.* 

JAKI, T & WOLFSEGGER, MJ (2010). Estimation of pharmacokinetic parameters with the R package PK. *Pharmaceutical Statistics*, published online ahead of print.

WOLFSEGGER, MJ & JAKI, T. Assessing systemic drug exposure in repeated dose toxicity studies in the case of complete and incomplete sampling. *Biometrical Journal* **51**(6), 1017–1029.

YEH, C (1990). Estimation and significant tests of area under the curve derived from incomplete blood sampling. *American Statistical Association Proceedings of the Biopharmaceutical Section*, 74–81.

イロト イポト イヨト イヨト