

*Non-linear mixed-effects models for tests of interaction  
or of lack of interaction in cross-over and parallel  
pharmacokinetic studies: application to the test of  
interaction between protease inhibitors and nucleoside  
analogs in HIV patients*

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## Context (1) : tests in PK cross-over trials

Tests comparing PK parameters between the 2 treatment arms

- interaction trials → comparison test
- bioequivalence trials → equivalence test

**Standard approach** (FDA 1999, EMEA 1998 and 2000)

Estimate the *AUC* by NC approach

- comparison: paired Student or Wilcoxon test on  $\log(AUC)$
- equivalence (Average Bioequivalence, FDA 2000)
  - Schuirmann's two one-sided test (TOST) on  $\log(AUC)$
  - 1 - estimate the  $CI_{90\%}$  of  $\mu_{AUC}^{(T)} - \mu_{AUC}^{(R)}$
  - 2 - compare this  $CI_{90\%}$  to  $[-0.2;0.2]$

Drawback:

- large number of samples per subject (between 10 and 20)
- underlying PK model not taken into account



## Context (2): anti-retroviral drugs of HIV infection

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### Anti-retrovirals prescribed in HIV infection

- combination of molecules with numerous interactions
- very high inter and intra-patient variability



# Plan

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- 1- evaluation of the type I error and power without modelling IOV
- 2- evaluation of the impact of modelling IOV
- 3- application to the Puzzle II study
- 4- evaluation of the randomization test on the interaction of ZDV on NFV/M8



## Methods (1) - Model and notations

Subject  $i$  ( $i = 1, \dots, N$ ), sampling times  $t_j$  ( $j = 1, \dots, n$ )

$y_{i,j}^{(k)}$  : observation of sujet  $i$  at time  $t_j$  for treatment  $k$  ( $k = R, T$ )

$$y_{i,j}^{(k)} = f(t_j, \phi_i^{(k)}) + \varepsilon_{i,j}^{(k)}$$

$\varepsilon$  : measurement error, gaussian with null mean and variance:

$$\sigma_{i,j}^{(k)2} = \sigma^2(a + f(t_j, \phi_i^{(k)}))^2$$

Individual parameters

$$\phi_i^{(k)} = \mu + \beta \mathbb{1}_{k=T} + \eta_i + c_i^{(k)}$$

$$\eta_i \rightsquigarrow \mathcal{N}(0, \Gamma), c_i^{(k)} \rightsquigarrow \mathcal{N}(0, \Psi)$$

$\Gamma$  and  $\Psi$  often supposed diagonal

$\log(AUC)$  is a component of  $\phi \rightarrow \phi_{AUC}$



## Methods (2) - Comparison tests

$$H_0 : \{ \mu_{AUC}^{(T)} - \mu_{AUC}^{(R)} = 0 \} \Leftrightarrow \{ \beta_{AUC} = 0 \}$$

### Standard tests (guidelines)

- separate analysis of each PK profile
- estimation of the individual NC  $AUC$
- paired Student and Wilcoxon comparing NC  $\log(AUC)_i^{(k)}$

### Development of 4 tests based on NLMEM

- Global tests
  - joint analysis of the two treatment groups
    - 1 - LRT, comparing model with  $\beta_{AUC} = 0$  and model with  $\beta_{AUC}$  estimated
    - 2 - extension of the Wald test comparing  $\beta_{AUC}$  to 0
- EB tests
  - separate analysis of each treatment group
  - estimation of the EB  $\theta_i^{(k)}$
  - Student and Wilcoxon comparing the EB  $\theta_{AUC, i}^{(k)}$



## Methods (3) - Equivalence tests

$$H_0 : \{ \mu_{\text{AUC}}^{(T)} - \mu_{\text{AUC}}^{(R)} \leq -\delta \text{ or } \mu_{\text{AUC}}^{(T)} - \mu_{\text{AUC}}^{(R)} \geq \delta \} \Leftrightarrow \{ \beta_{\text{AUC}} \leq -\delta \text{ or } \geq \delta \}$$

Typically  $\delta = 0.2$  ( $e^{-\delta} = 0.8$  and  $e^{\delta} = 1.25$ )

### Standard tests (guidelines)

- Student TOST on NC  $\log(\text{AUC})_i^{(k)}$
- adaptation of TOST to the Wilcoxon test on NC  $\log(\text{AUC})_i^{(k)}$   
(Chow and Liu, 1999)

### Global tests

- no simple extension of the LRT for equivalence
- Wald: same principle as for the TOST using  $\text{CI}_{90\%}$  of  $\beta_{\text{AUC}}$   
(SE of  $\beta_{\text{AUC}}$  is estimated by nlme)

### EB tests

- adaptation of the standard tests to the  $\theta_{\text{AUC}, i}^{(k)}$



## Evaluation by simulation - Theophylline

### Population PK on the original dataset

- $N=12$  subjects,  $n=10$  samples per subject
- one compartment, 1<sup>st</sup> order absorption and elimination
- parametrization on  $\log(k_a)$ ,  $\log(AUC)$ ,  $\log(V)$

### Estimated values used to simulate concentration data

- fixed effects:  $\hat{\mu}_{k_a} = 0.39$ ,  $\hat{\mu}_{AUC} = 4.61$  and  $\hat{\mu}_V = -0.73$
- combined error model:  $a=1$ ,  $\sigma=0.1$
- SD of the random effects for  $\log(k_a)$ ,  $\log(AUC)$  and  $\log(V)$  resp.
  - IIV: 0.10, 0.20 and 0.20
  - IOV: 0.05, 0.10 and 0.10

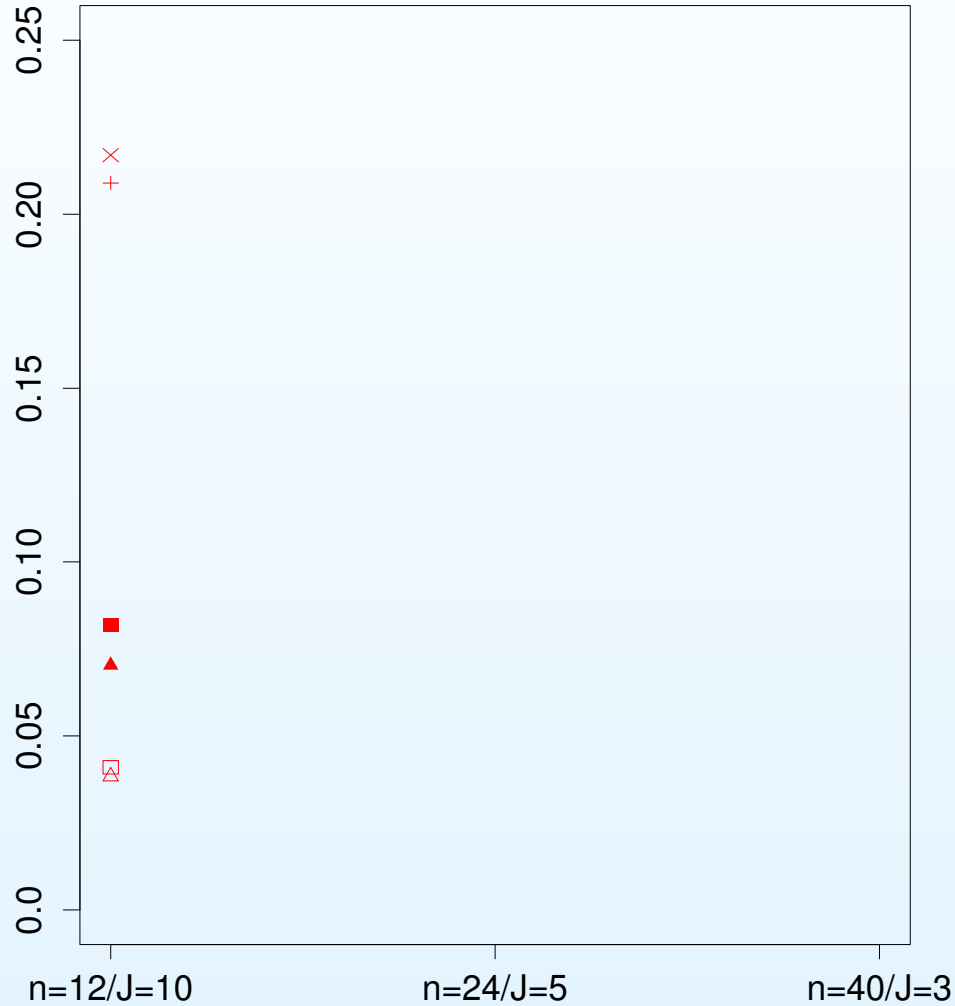
Simulated designs: combinations of  $N=12, 24$  and  $40$ ,  $n=10, 5$  and  $3$





# Results (1) - Type I error

## Comparison



+ : LRT

■ : Student EB

□ : Student NC

× : Wald

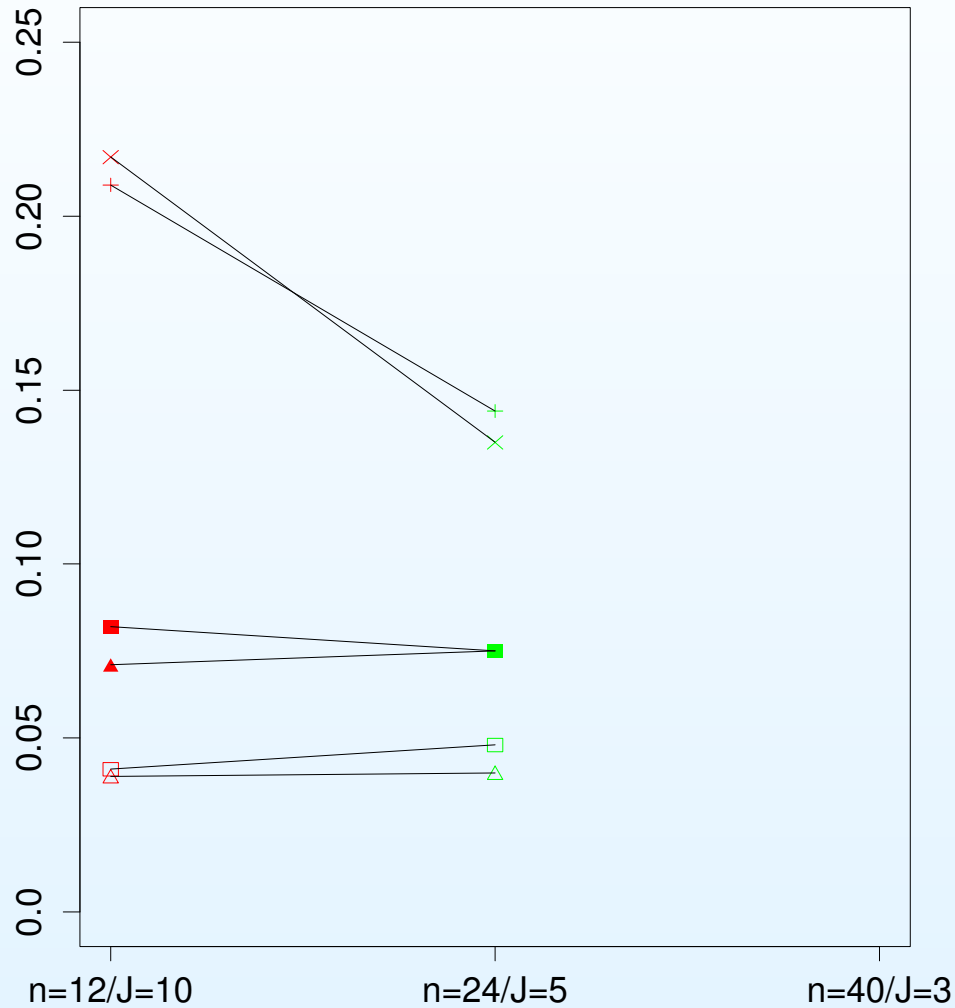
▲ : Wilcoxon EB

△ : Wilcoxon NC



# Results (1) - Type I error

## Comparison



+ : LRT

■ : Student EB

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× : Wald

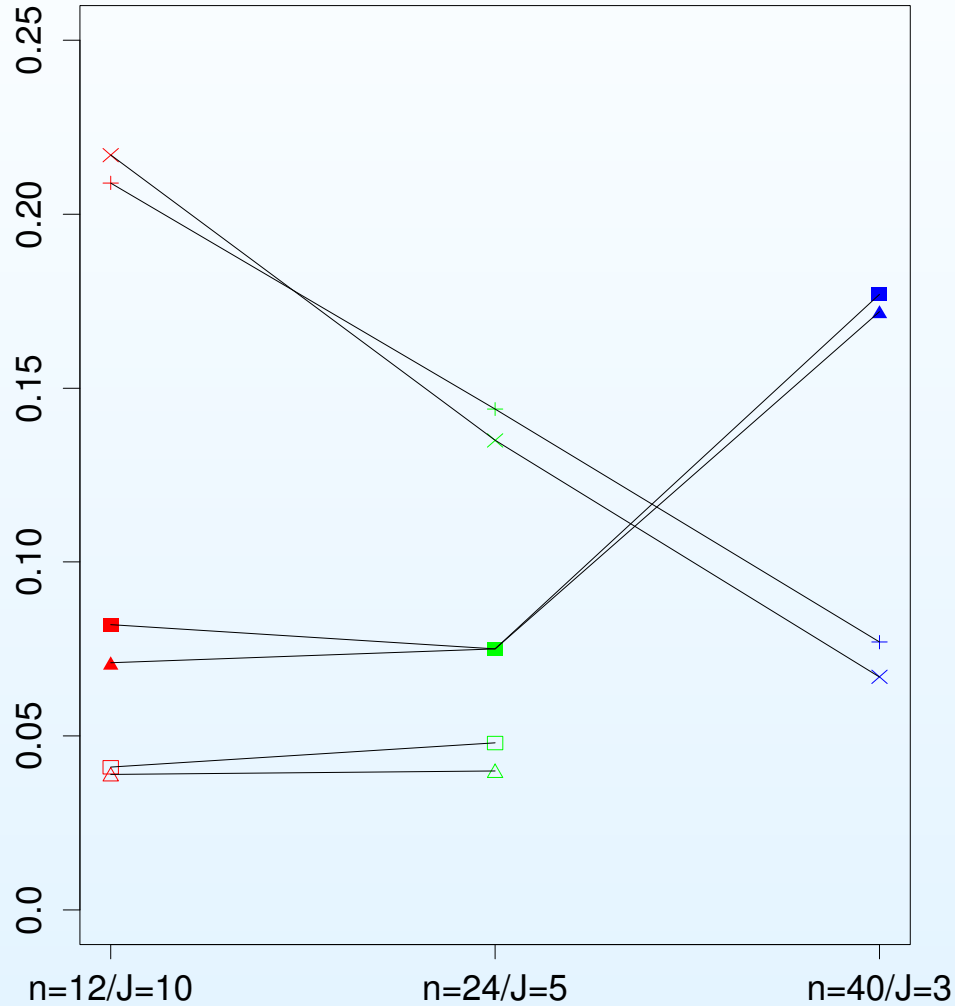
▲ : Wilcoxon EB

△ : Wilcoxon NC



# Results (1) - Type I error

## Comparison



+ : LRT

■ : Student EB

□ : Student NC

x : Wald

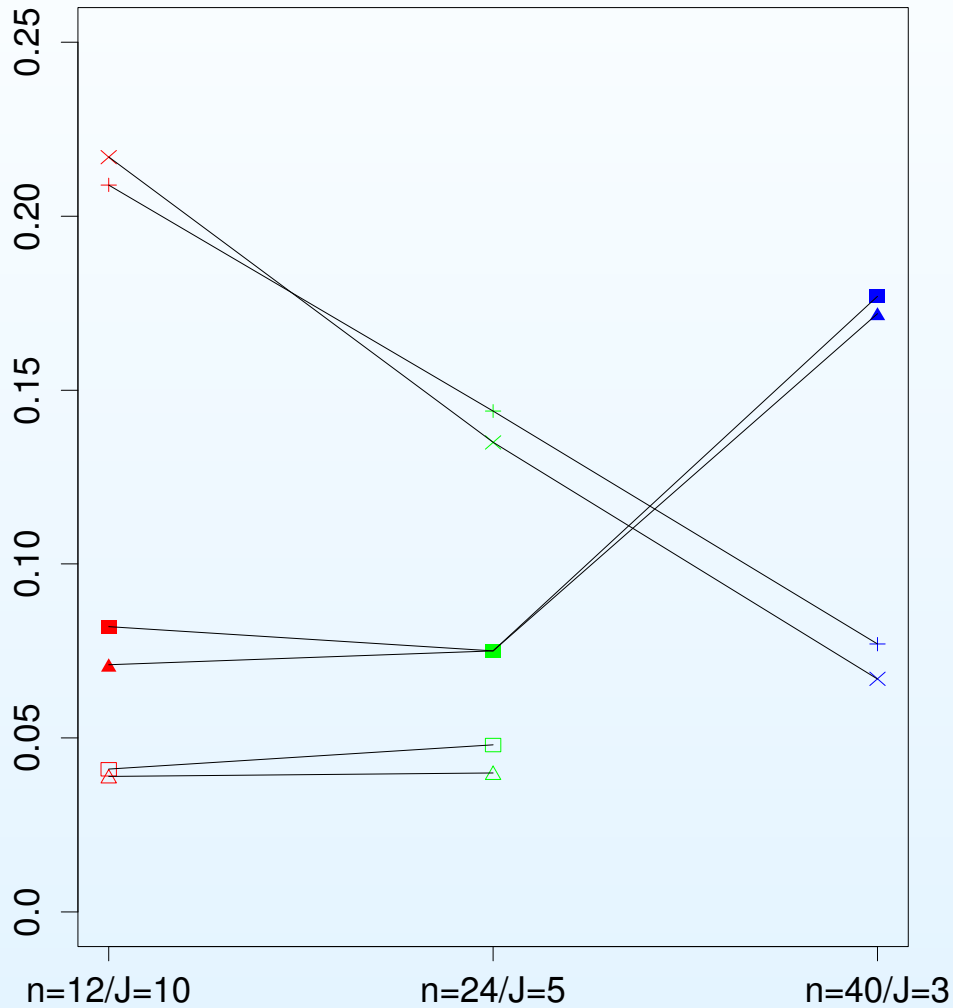
▲ : Wilcoxon EB

△ : Wilcoxon NC

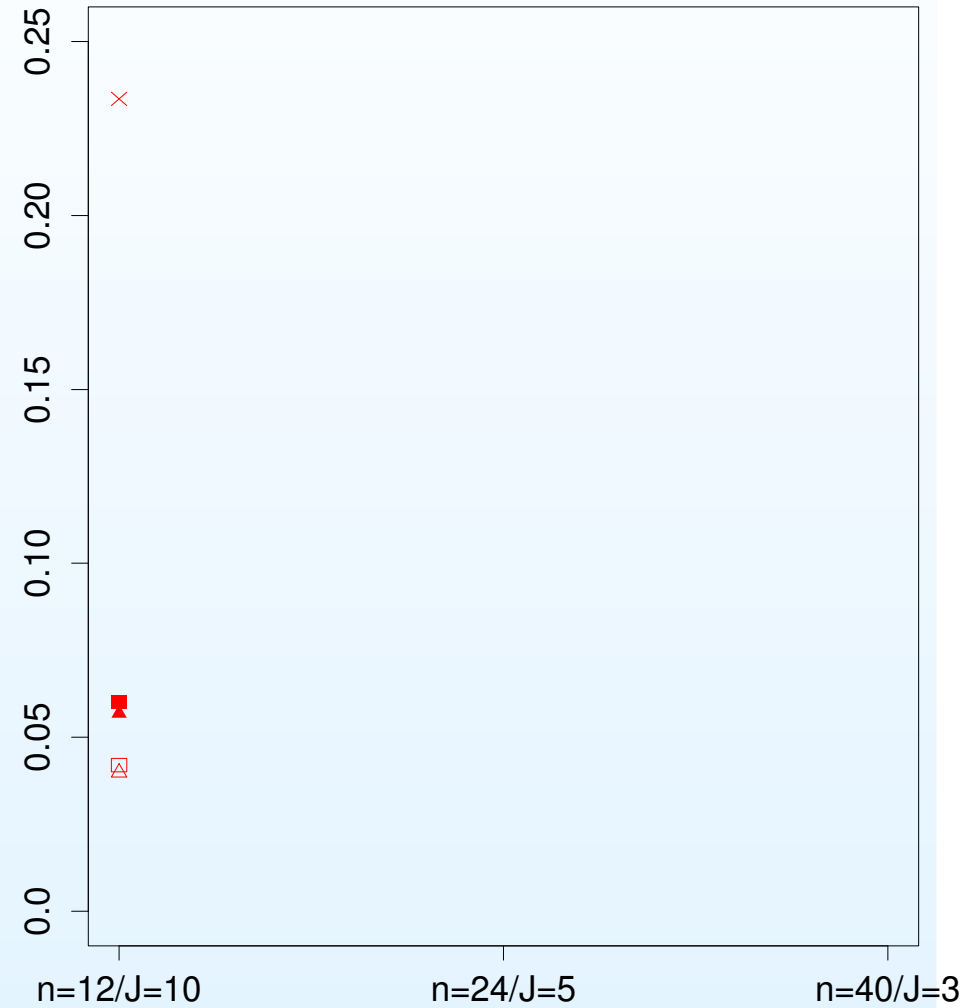


# Results (1) - Type I error

## Comparison



## Equivalence

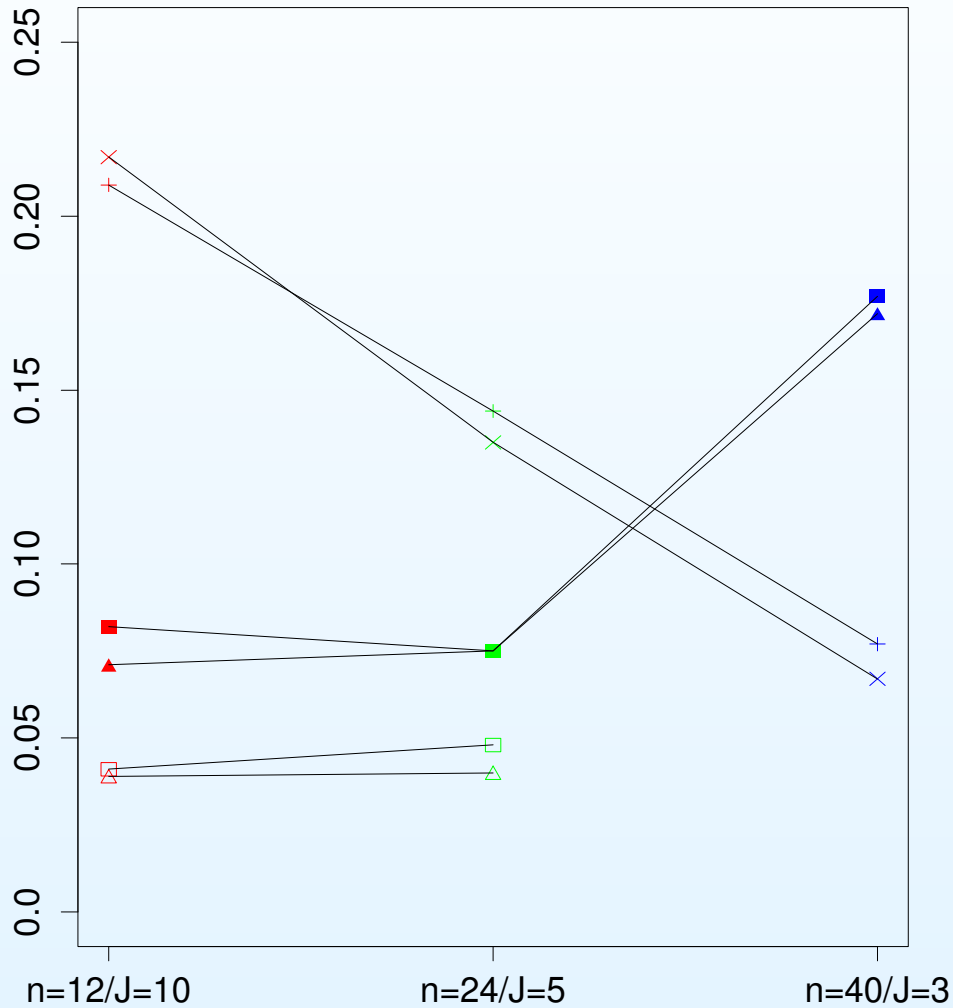


+ : LRT   ■ : Student EB   □ : Student NC   × : Wald   ▲ : Wilcoxon EB   △ : Wilcoxon NC

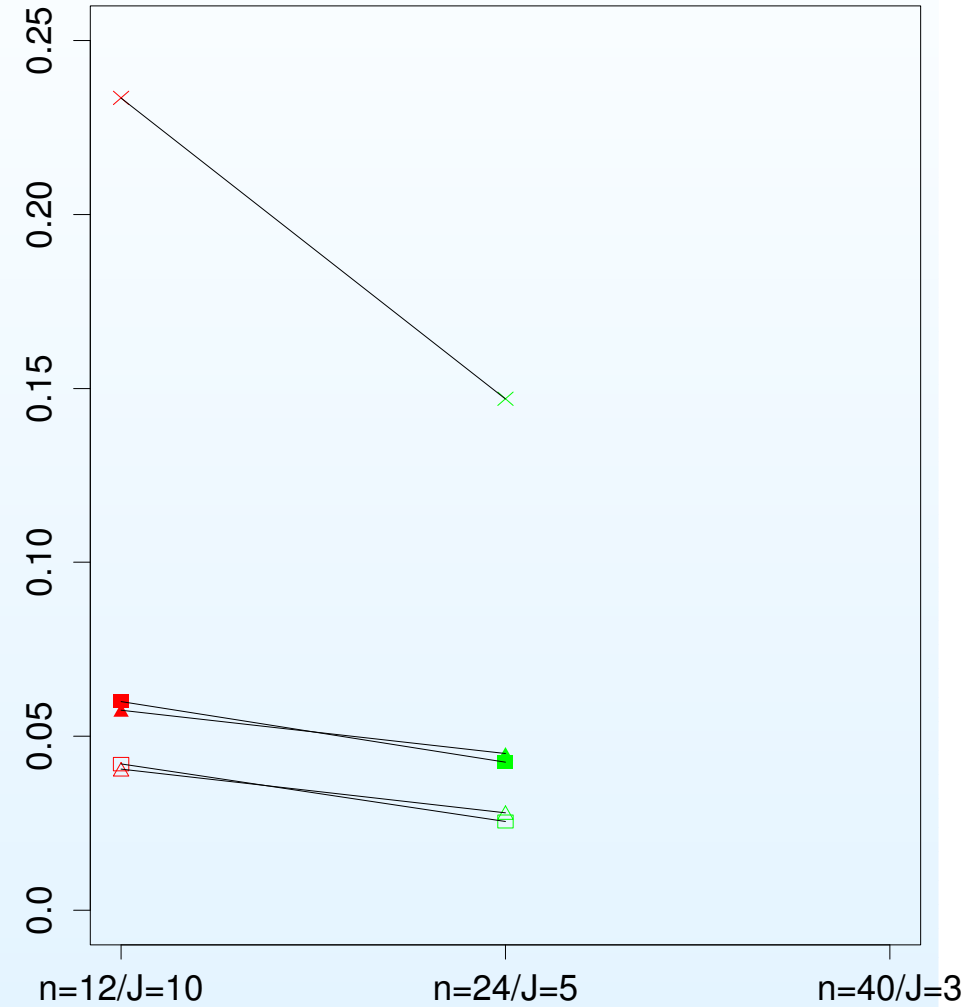


# Results (1) - Type I error

## Comparison



## Equivalence

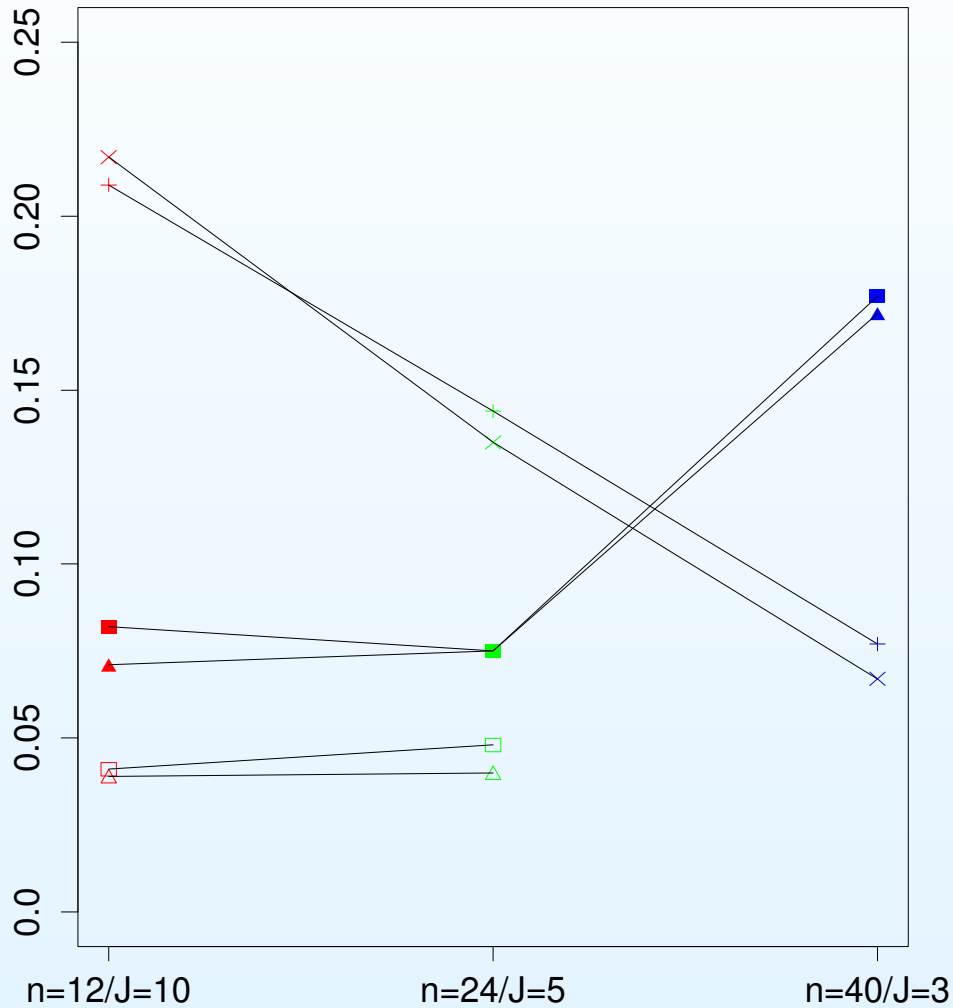


+ : LRT   ■ : Student EB   □ : Student NC   × : Wald   ▲ : Wilcoxon EB   △ : Wilcoxon NC

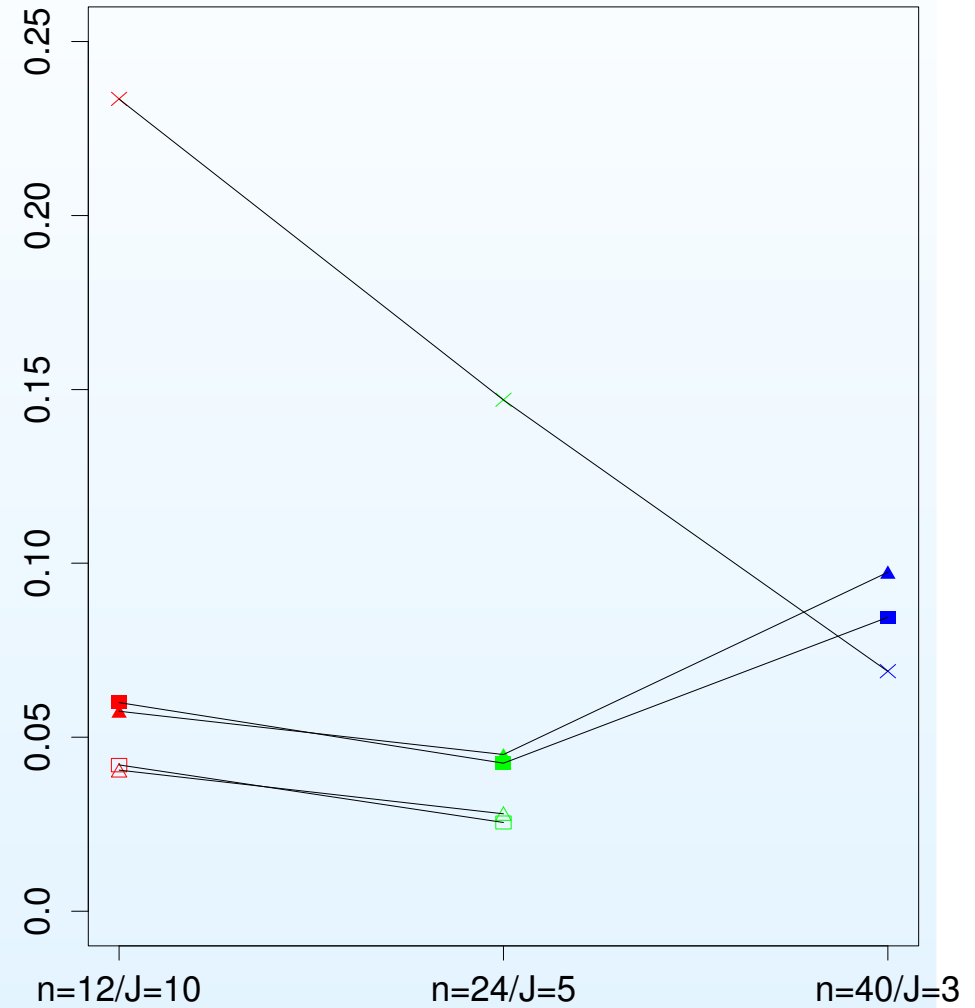


# Results (1) - Type I error

## Comparison



## Equivalence

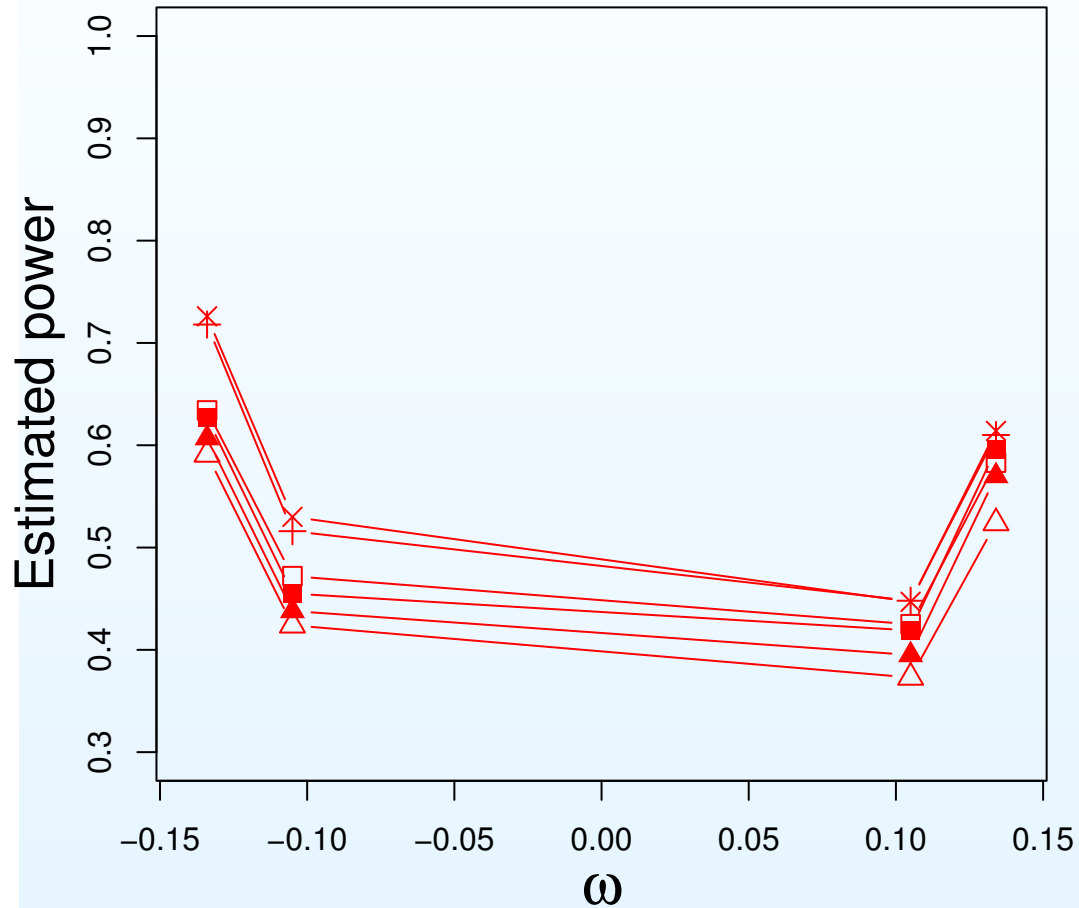


+ : LRT   ■ : Student EB   □ : Student NC   × : Wald   ▲ : Wilcoxon EB   △ : Wilcoxon NC



# Results (2) - Power

## Comparison



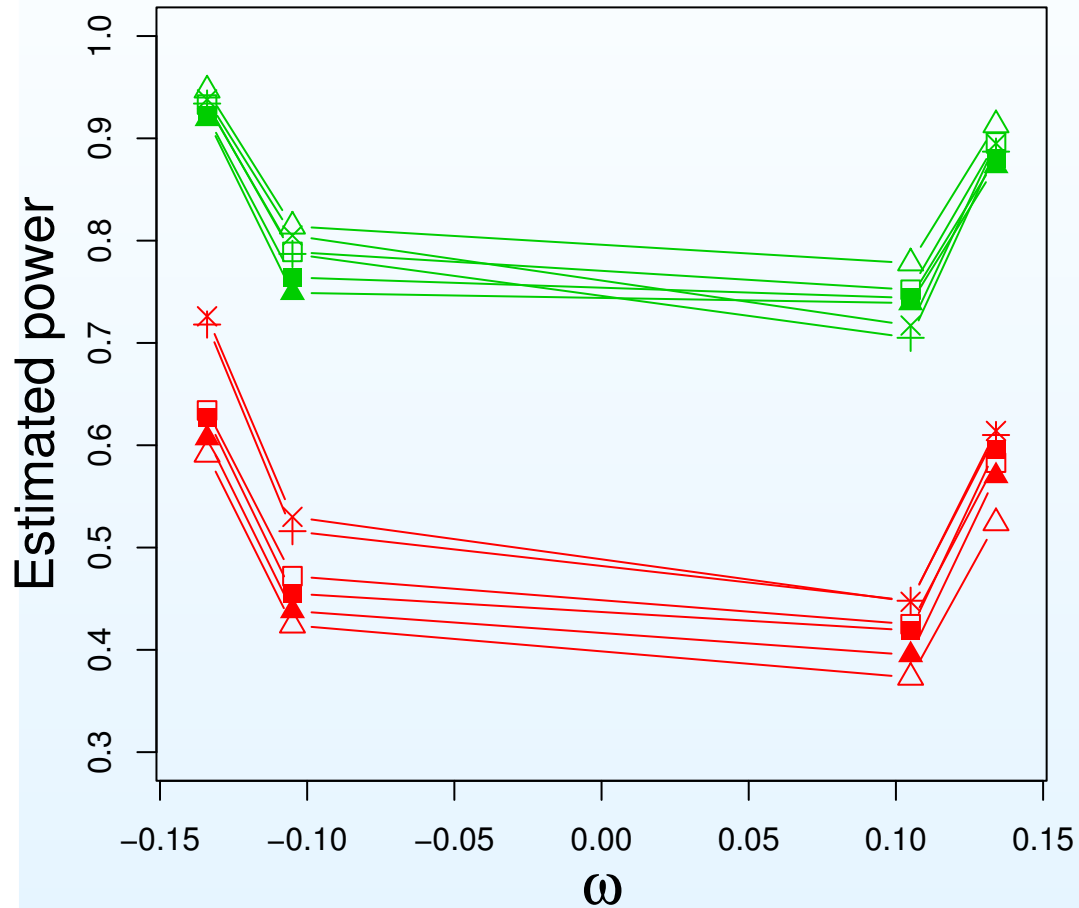
+ : LRT    ■ : Student EB    □ : Student NC  
 $\times$  : Wald    ▲ : Wilcoxon EB    △ : Wilcoxon NC

— :  $n = 12, J = 10$   
— :  $n = 24, J = 5$   
— :  $n = 40, J = 3$



# Results (2) - Power

## Comparison



+ : LRT    ■ : Student EB    □ : Student NC  
× : Wald    ▲ : Wilcoxon EB    △ : Wilcoxon NC

— :  $n = 12, J = 10$

— :  $n = 24, J = 5$

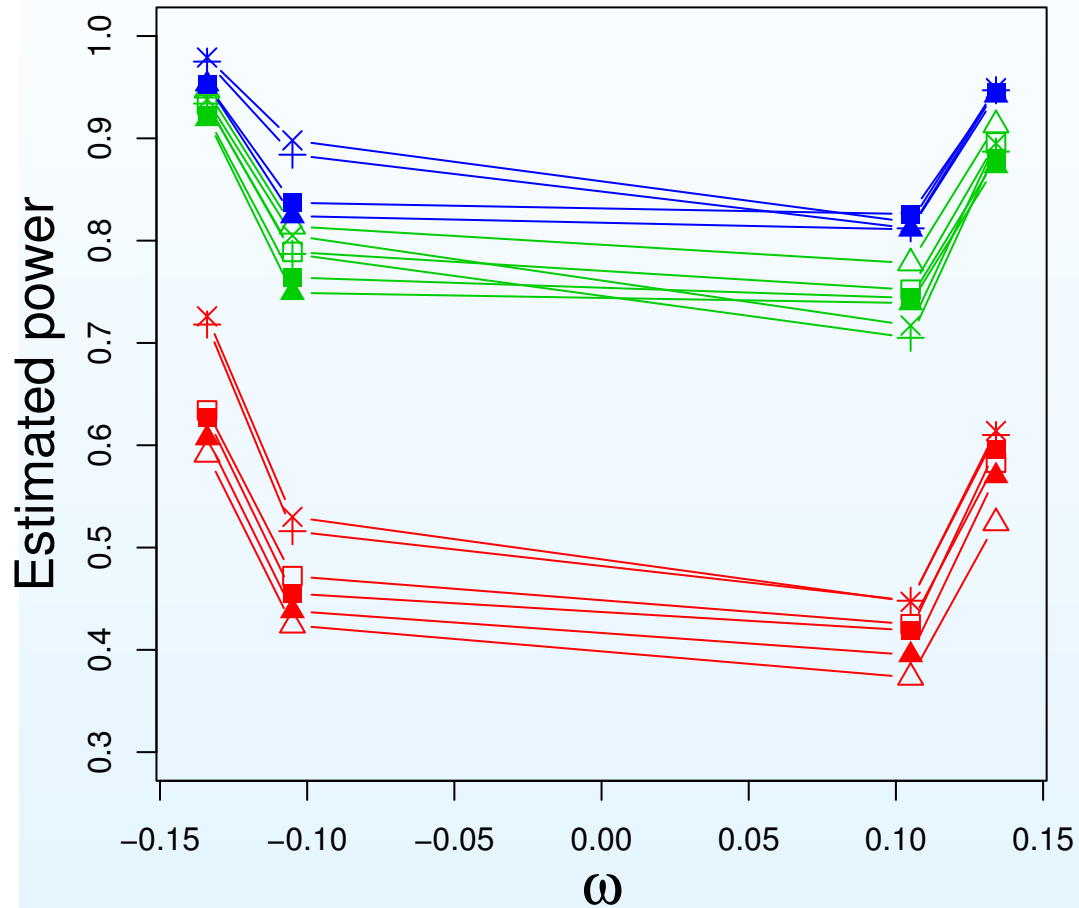
— :  $n = 40, J = 3$





# Results (2) - Power

## Comparison



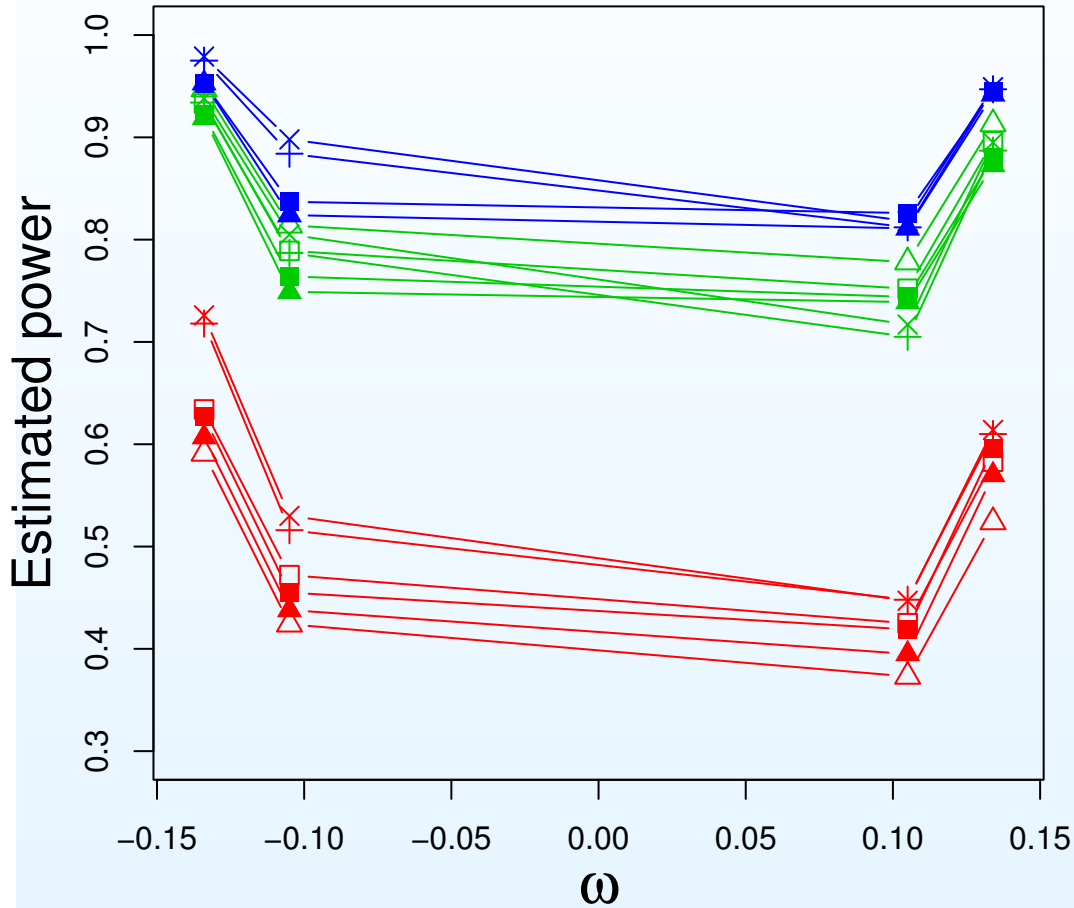
+ : LRT    ■ : Student EB    □ : Student NC  
 $\times$  : Wald     $\blacktriangle$  : Wilcoxon EB     $\triangle$  : Wilcoxon NC

— :  $n = 12, J = 10$   
— :  $n = 24, J = 5$   
— :  $n = 40, J = 3$

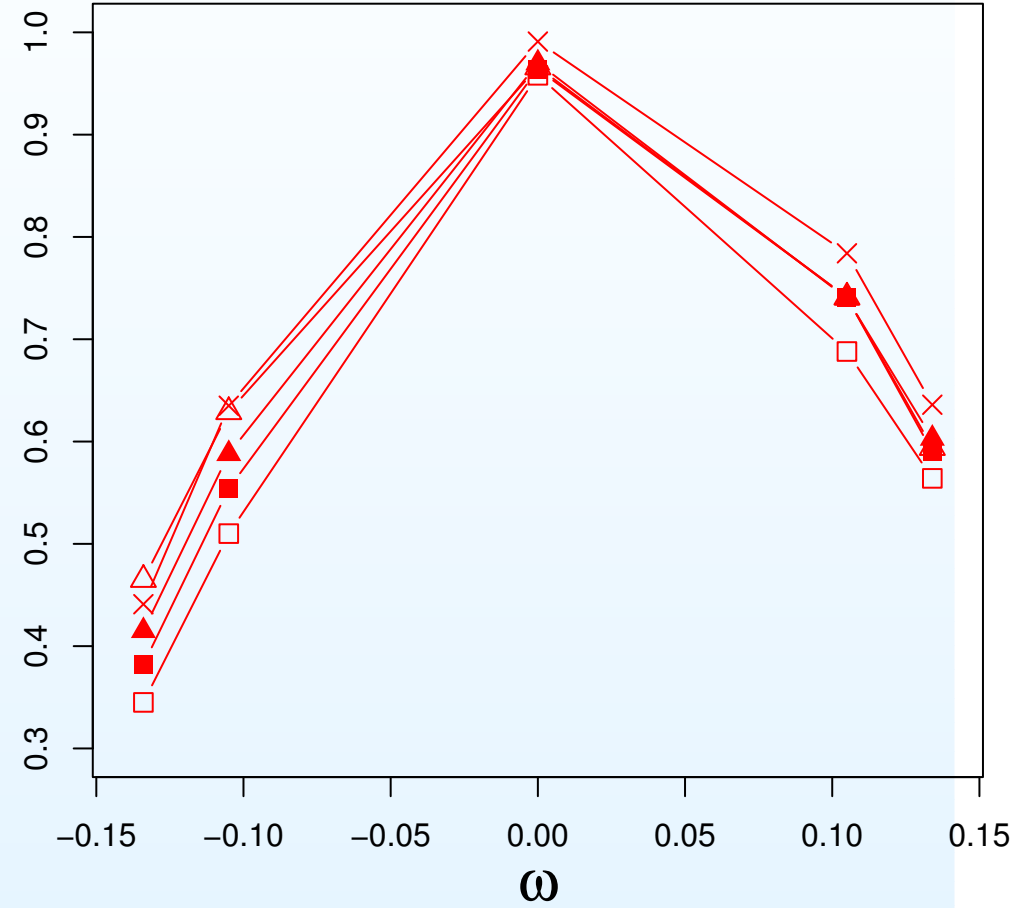


# Results (2) - Power

## Comparison



## Equivalence



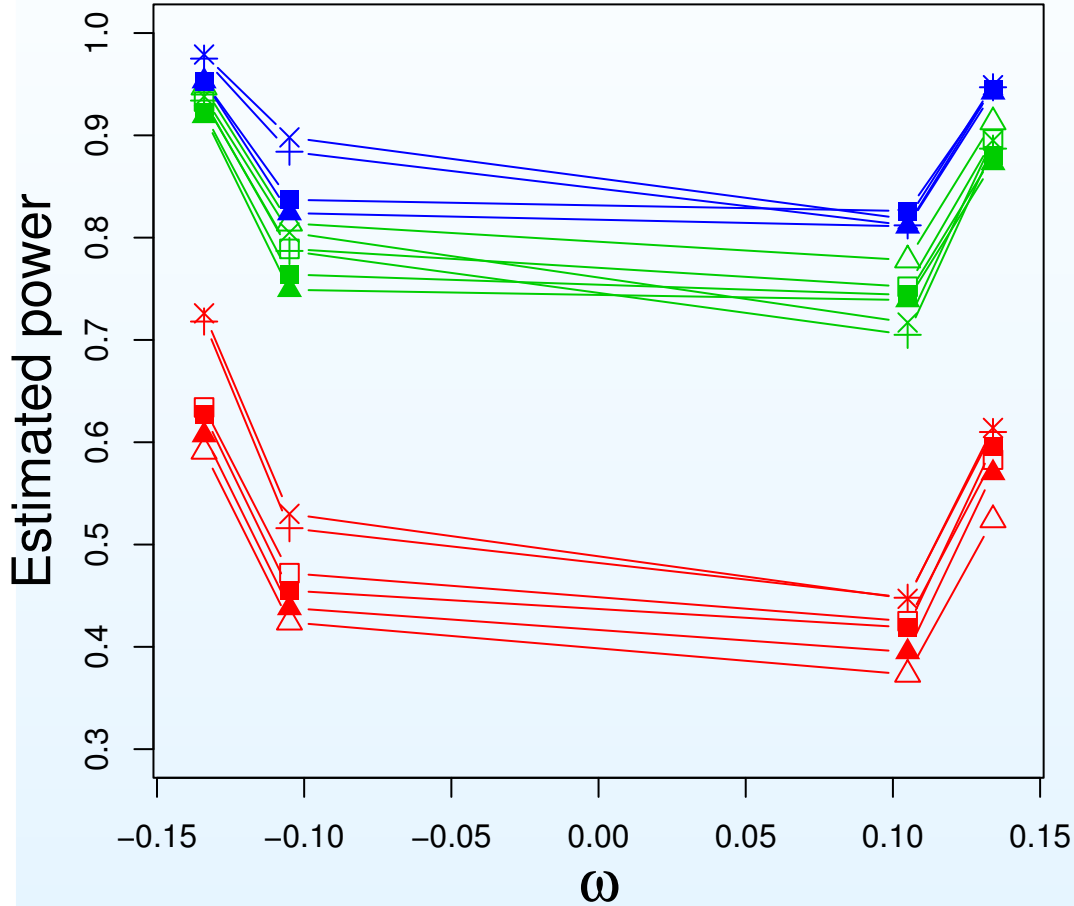
+ : LRT      ■ : Student EB      □ : Student NC  
 × : Wald    ▲ : Wilcoxon EB    △ : Wilcoxon NC

— :  $n = 12, J = 10$   
 — :  $n = 24, J = 5$   
 — :  $n = 40, J = 3$

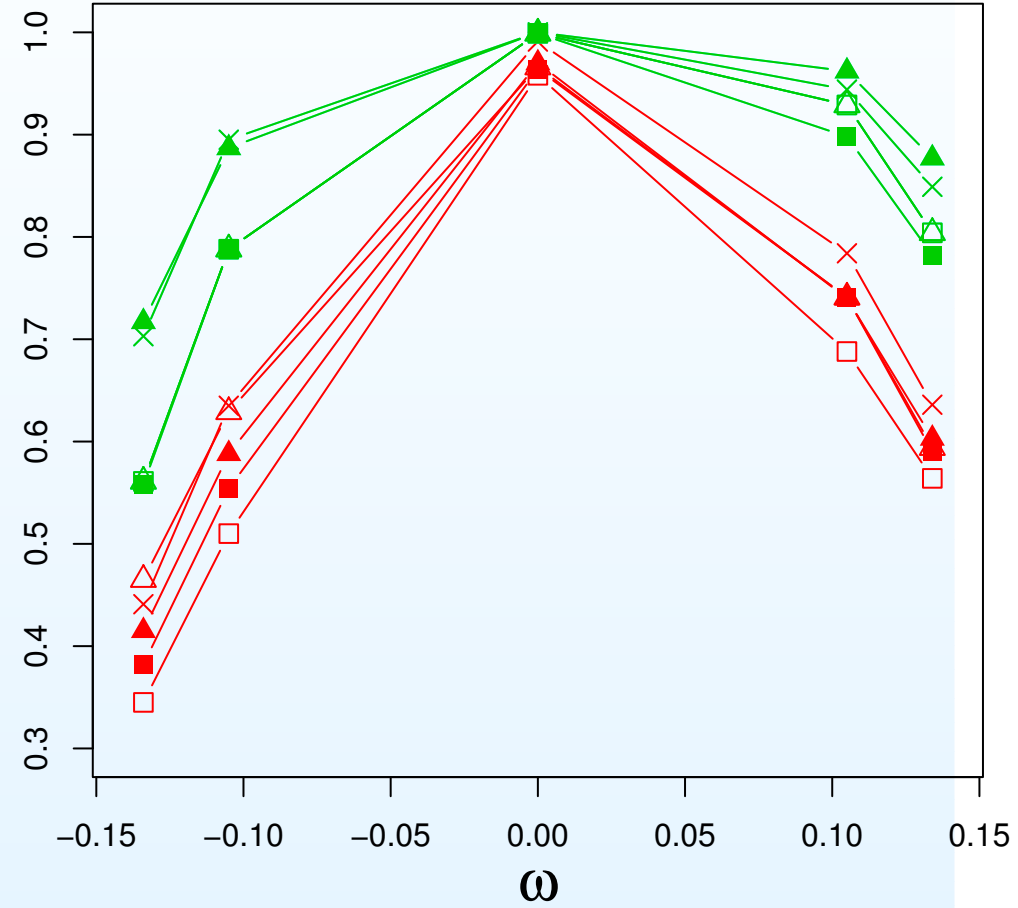


# Results (2) - Power

## Comparison



## Equivalence



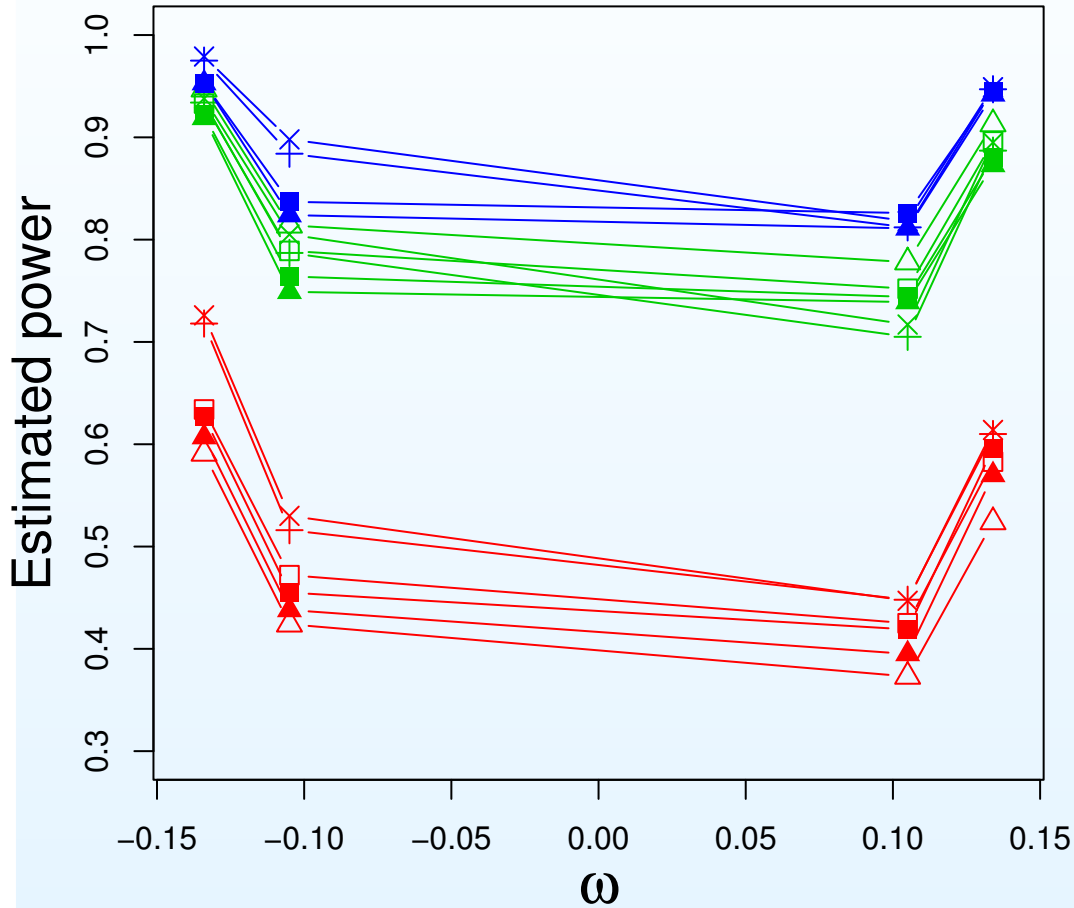
+ : LRT      ■ : Student EB      □ : Student NC  
 × : Wald     ▲ : Wilcoxon EB     △ : Wilcoxon NC

— :  $n = 12, J = 10$   
 — :  $n = 24, J = 5$   
 — :  $n = 40, J = 3$

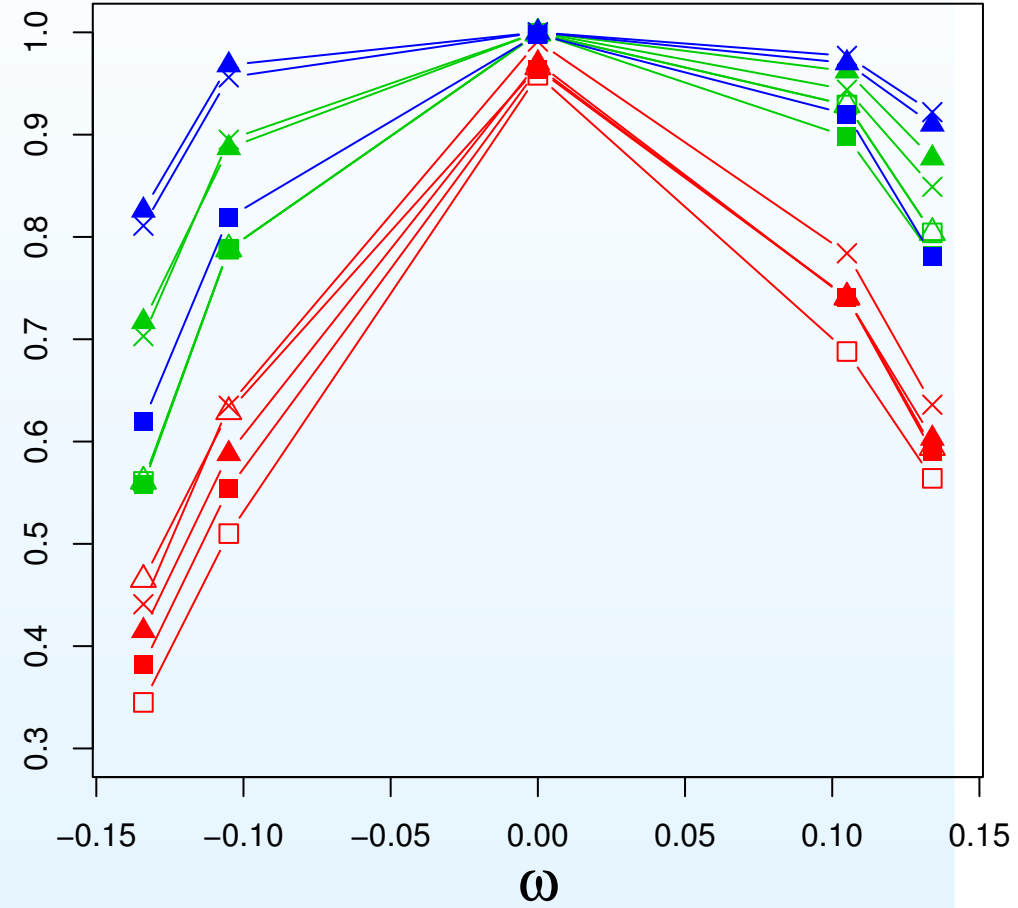


# Results (2) - Power

## Comparison



## Equivalence



+ : LRT      ■ : Student EB      □ : Student NC  
 × : Wald     ▲ : Wilcoxon EB     △ : Wilcoxon NC

— :  $n = 12, J = 10$   
 — :  $n = 24, J = 5$   
 — :  $n = 40, J = 3$



## Results (3)

- good performance of EB tests
- inflation of the type I error of global tests at finite distance
- the 2 global tests are the most powerful

→ Panhard X, Mentré F. *Evaluation by simulation of tests based on non-linear mixed-effects models in pharmacokinetic interaction and bioequivalence cross-over trials.* **Stat Med** 2005



## Impact of modelling IOV

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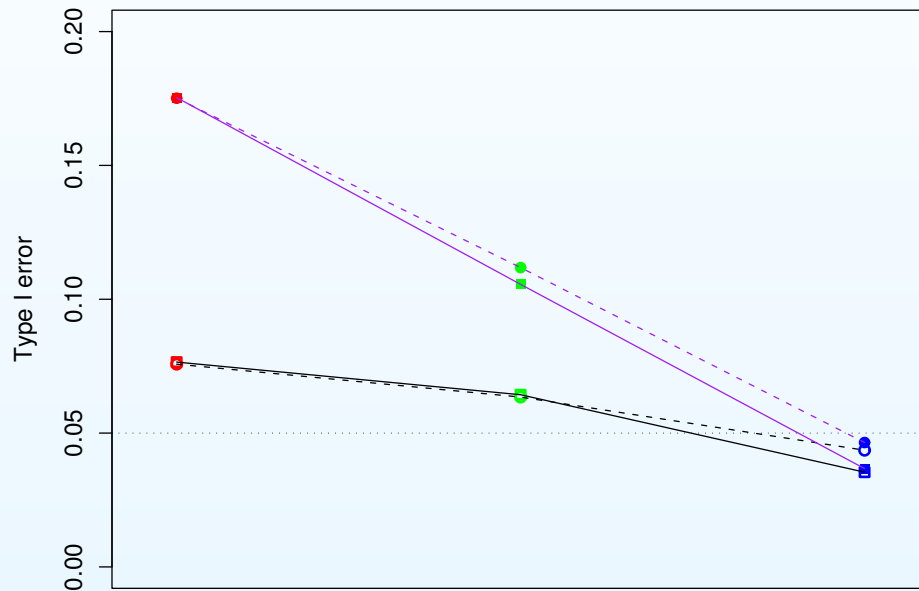
- same PK model (theophylline)
- only designs with 240 samples per subject
- for each parameter, estimation of :
  - IIV
  - IOV
  - an interaction effect
- evaluation of type I error and power (only for interaction)



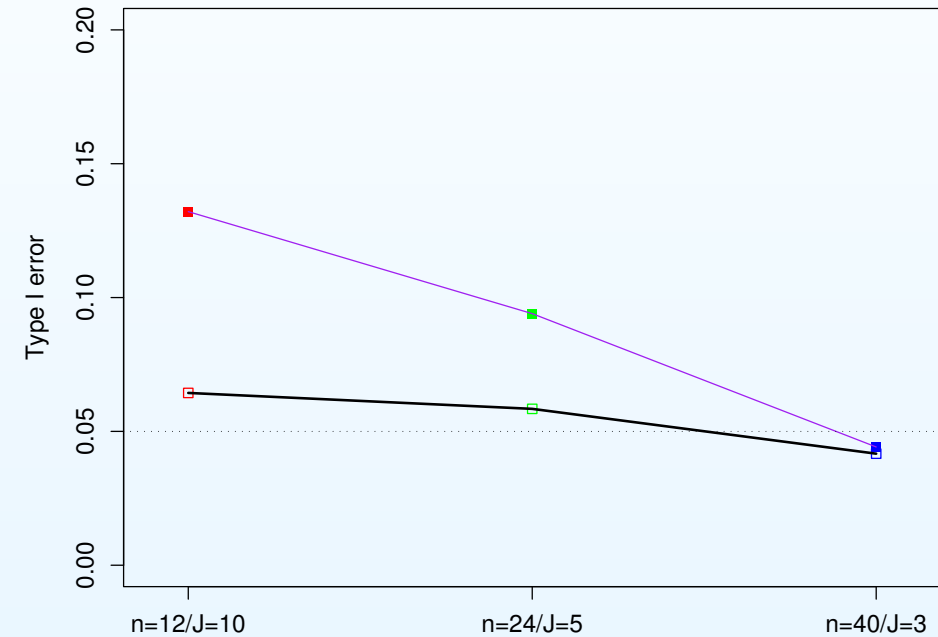
# Results

## Type I error

### Comparaison



### Equivalence



## Power

- evaluated for  $e^\delta = 0.8, 0.9, 1.1$  and  $1.25$
- good power for the 2 tests
- 1 to 2% inferior to that obtained without modelling IOV



## Application: ANRS 107 - Puzzle II

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Prospective, open, multicenter trial in HIV+ patients:

- under stable treatment for at least 1 month
- with a viral load  $> 10\,000$  copies/mL

PK substudy: 2 period cross-over in 10 patients

- from inclusion to W2 : atazanavir (ATZ)
- from W3 to W26 : ATZ + tenofovir (TFV)

Objective: evaluate the interaction of TFV on ATZ PK

Samples taken at W2 and W6: 1, 2, 3, 5, 8, and 24 h after drug intake





# Population PK of atazanavir (1)

## Objectives

- build the PK model, taking the 2 periods into account
- estimate IIV and IOV
- test the interaction effect of TFV on the PK parameters

## PK model

- one compartment model with zero order absorption and 1<sup>st</sup> order elimination
- parametrized in  $\log(T_a)$ ,  $\log(Cl/F)$  and  $\log(V/F)$
- homoscedastic variance

$$f(\theta, t) = \frac{FD}{T_a Cl} \left( (1 - e^{-\frac{Cl}{V}t}) \mathbb{1}_{t < T_a} + \frac{e^{-\frac{Cl}{V}\tau} \mathbb{1}_{t < T_a} (1 - e^{-\frac{Cl}{V}T_a}) e^{-\frac{Cl}{V}(t-T_a)}}{(1 - e^{-\frac{Cl}{V}\tau})} \right)$$



## Resultats - population PK

	Mean	SE	IIV (%)	IOV(%)
$\log(T_a)$	1.32	0.10	21.7	0
$\beta_{T_a}$	0.306	0.10	—	—
$\log(AUC)$	10.7	0.17	49.2	0
$\beta_{AUC}$	-0.380	0.090	—	—
$\log(V/F)$	4.01	0.20	0	53.5
$\beta_{V/F}$	0.159	0.003	—	—

Significant interaction effect (Wald test) of TVF on:

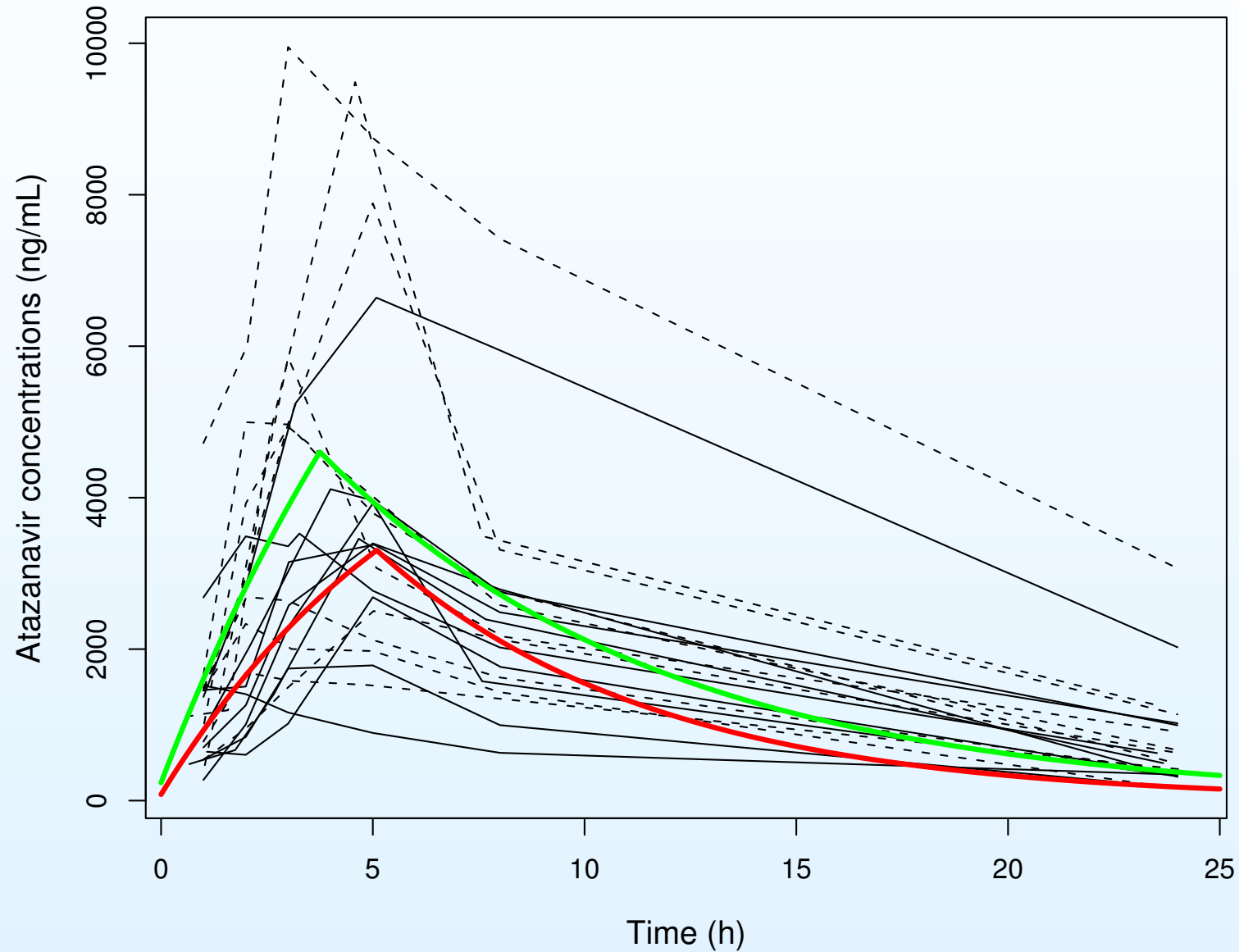
- $\log(AUC)$ :  $p < 10^{-4}$
- $\log(T_a)$ :  $p = 0.0019$

Equivalence between the 2 treatment groups for:

- $\log(V/F)$ : 90% CI for  $\beta_{V/F} = [-0.335 ; 0.652]$



# Concentrations and predicted curves



# Test of interaction in usual population PK analyses

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## Inflation of the type I error of the test of binary covariates

- already showed by several authors for comparison tests  
→ no reevaluation by simulation
- among correction methods: randomization tests
- possible extension to absence of interaction



## Interaction of ZDV on the PK of NFV/M8 (1)

- Cophar I - ANRS 102 study: prospective, open, multicenter trial in HIV+ patients
  - under stable treatment for at least 4 months
  - with a viral load <200 copies/mL for at least 4 months
- Nelfinavir (NFV) and M8 concentrations obtained in 46 patients
  - first visit: before and 0.5, 1, 3 and 6h after drug intake
  - second visit: before and 3h after drug intake

→ Panhard X et al. *Population pharmacokinetic analysis for nelfinavir and its metabolite M8 in virologically controlled HIV-infected patients on HAART*. **Brit J Clin Pharmacol** in press



## Simultaneous population PK of NFV/M8

- NFV: one compartment, 1<sup>st</sup> order absorption and elimination
- M8: one compartment, 1<sup>st</sup> order metabolization rate constant  $k_m$
- identifiable parameters:
  - NFV:  $V/F$ ,  $Cl/F$  and  $k_a$
  - M8:  $V_m/Fk_m$  and  $Cl_m/Fk_m$
- selection of random effects based on AIC:
  - IIV estimated on  $V/F$ ,  $Cl/F$  and  $Cl_m/Fk_m$
  - IOV estimated on  $Cl/F$
- combined error model



## Use of the randomization test

Significant interaction effect in the final model (LRT):

- $Cl/F$  increased by 1.2 fold ( $p_{LRT} < 10^{-4}$ ,  $p_{Wald} = 0.135$ )
- $Cl_m/Fk_m$  decreased by 1.8 fold ( $p_{LRT} = 0.020$ ,  $p_{Wald} = 0.011$ )

in the 27 patients receiving ZDV

Randomization test

- 1000 random permutations of comedication with ZDV
- pop PK analysis of the corresponding data sets
- evaluation of the significance of the interaction effect

Resulting corrected p-values

- $Cl/F$ :  $p_{LRT} = 0.030$ ,  $p_{Wald} = 0.170$
- $Cl_m/Fk_m$ :  $p_{LRT} = 0.052$ ,  $p_{Wald} = 0.016$



# Conclusion

- tests based on NLMEM allow
  - to test PK interaction or lack of interaction
  - to greatly decrease the number of samples per patient
    - great interest for trials performed
      - in patients, as HIV patients illustrated here
      - special populations (children, older patients)
- necessity of a correction method the type I error?
  - need of further evaluation
  - depends on the estimation method or algorithm
- next step: planification of PK interaction studies
  - estimation of the expected SE taking IOV into account using PFIM
  - estimation of the corresponding power or sample size

