

A comparison of two model-based approaches to investigating covariate effects on the dose-exposure relationship in a Phase III context

Justin J Wilkins and Michael Looby

Novartis Pharma AG, Basel, Switzerland



Background

- In Phase III, where large populations are often exposed to chronic treatment, it is typical to attempt to identify covariates which explain variability in the dose exposure relationship
- Typically, this type of analysis is performed using sparse data (3-4 observations per visit) over a relatively long period (6-12 months)
- Gold standard method is compartmental analysis using nonlinear mixed-effects (NLME)
- Searching for covariate relationships in this manner can be extremely time-consuming
- Investigation of methods which allow more rapid, but nevertheless adequately sensitive identification of covariate effects are warranted

Objective

- Compare the NLME method with a linear mixed-effects (LME) approach in terms of their respective abilities to adequately elucidate relevant covariate relationships on exposure in a subject-rich but observation-poor context
- Methods will be compared in terms of sensitivity, accuracy, precision, and analytical time

Methods

Design

- A design corresponding to a typical Phase III confirmatory trial was used for simulation
- PK samples were simulated at 4 visits over a period of 6 months – on Day 1, Day 14, Day 84 and Day 182
- PK samples were simulated at nominal times of 1.5 h, 4 h and 23.5 h
- 400 patients were included
- Dosing was 600 µg once daily

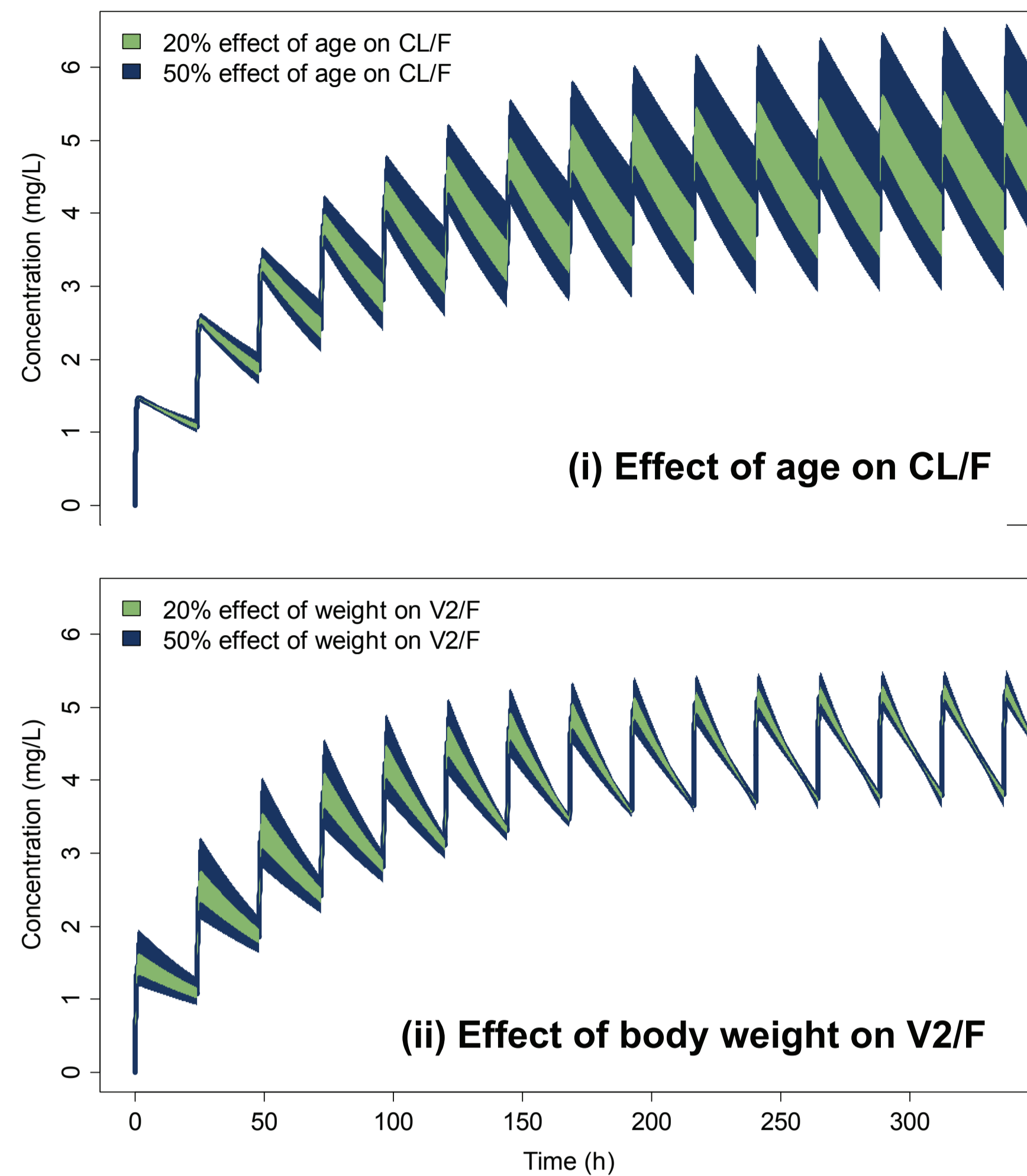
Simulation Model

- Fictitious compound with low clearance, high volume and consequently high accumulation (approximately five-fold)
- One-compartment model with linear absorption and elimination (Table 1)
- Covariate effects on CL/F and V2/F (between 5% and 50% on model parameters, over the range of values studied (20-80 y for age, and 40-120 kg for body weight)) – Figure 1 (5% effect not shown)

Table 1. Base simulation model

Parameter	Typical value	Variability
Apparent clearance (CL/F, L/h)	5.50	40%
Apparent central volume of distribution (V2/F, L)	400	40%
Absorption rate constant (ka, /h)	4.00	50%
Proportional residual error (σ^2 , variance)	0.09	–

Figure 1. Covariate effects



Simulation and analysis

- Each scenario simulated 1000 times using SSE tool in Perl-speaks-NONMEM (PsN)
- Each replicate was analyzed using two discrete approaches
 - NLME analysis of all the simulated data in a given scenario using a compartmental model implemented in NONMEM VI 2.0 (4800 observations per model fit)
 - LME analysis of observed steady-state peaks and pre-dose troughs only, as implemented in nlme package of R 2.8.1 (1200 observations per model fit)
- Results assessed in terms of peak and trough concentrations (observed or predicted)

Results

Table 2. Power to detect ALL covariate effects simultaneously (%) using both approaches. For LME, expressed as troughs [peaks]

Weight on V2/F	Age on CL/F					
	0%		20%		50%	
0%	LME	–	LME	30.2 [30.0]	LME	96.7 [94.3]
	Log LME	–	Log LME	29.1 [26.6]	Log LME	94.2 [90.5]
	NLME 5	–	NLME 5	72.8	NLME 5	99.8
	NLME 0.1	–	NLME 0.1	46.9	NLME 0.1	97.6
20%	LME	5.50 [5.90]	LME	2.20 [1.80]	LME	5.40 [6.10]
	Log LME	6.70 [6.50]	Log LME	2.20 [2.00]	Log LME	6.10 [6.10]
	NLME 5	47.0	NLME 5	28.5	NLME 5	61.0
	NLME 0.1	13.5	NLME 0.1	5.10	NLME 0.1	31.2
50%	LME	6.20 [12.0]	LME	2.40 [4.30]	LME	5.40 [6.10]
	Log LME	8.60 [11.7]	Log LME	3.30 [4.20]	Log LME	6.10 [6.10]
	NLME 5	94.6	NLME 5	41.4	NLME 5	85.2
	NLME 0.1	73.7	NLME 0.1	13.0	NLME 0.1	65.6

LME=linear mixed effects, ordinal; log LME=LME, log-transformed both sides; NLME 5=NLME, p < 0.05; NLME 0.01=NLME, p < 0.01. For LME: troughs [peaks]

Power to detect covariate effects

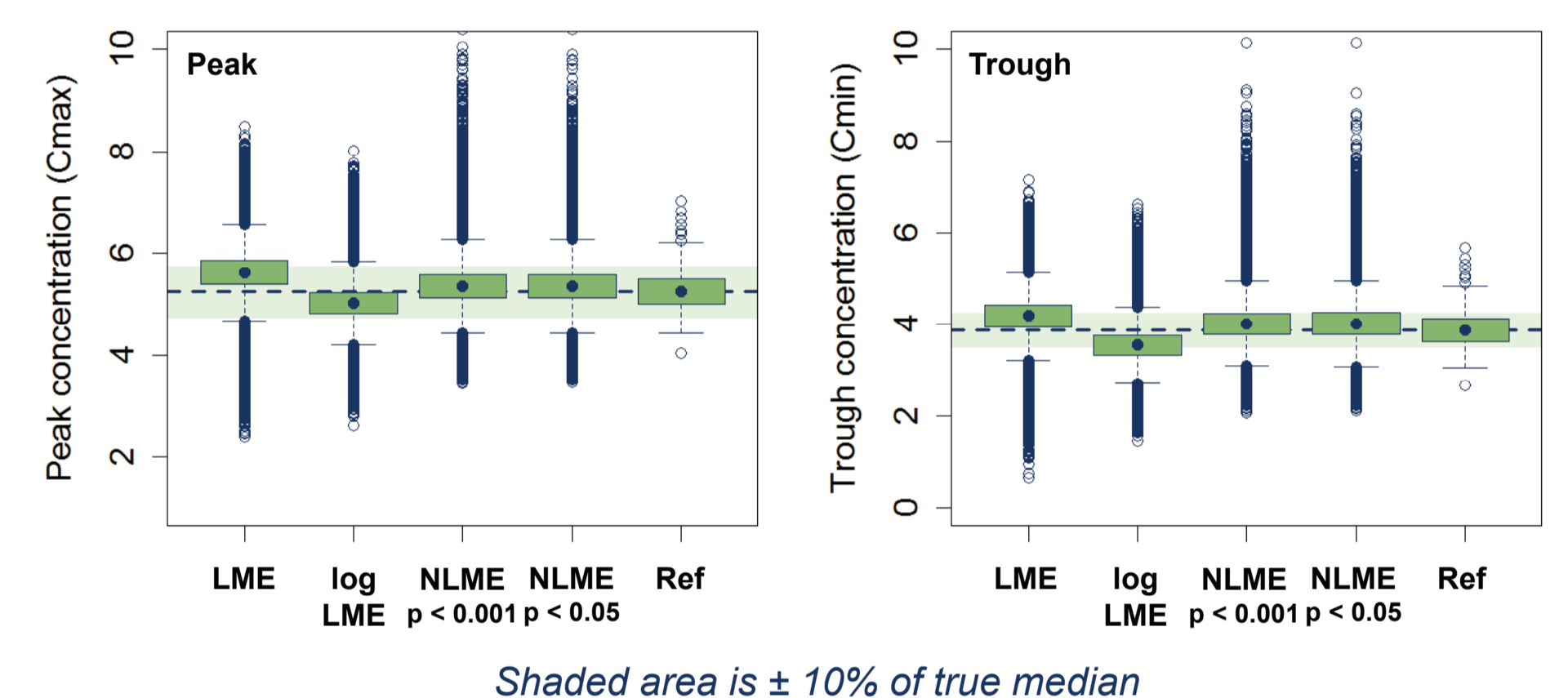
- NLME was most sensitive method, but also used much more data
- Clinically relevant effect (50%) on CL/F detected by LME with almost the same power
- Power of LME to detect effects on V2/F (at steady state) virtually zero in simulated scenario

Table 3. Bias (mean error, %) and precision (RMSE, %) of steady-state peak and trough concentration estimates produced by each method

Weight on V2/F	Age on CL/F					
	0%		20%		50%	
0%	Key	Bias (ME, %)	LME	7.13 [6.70]	LME	7.05 [6.70]
		Precision (RMSE, %)	LME	8.45 [7.56]	LME	8.24 [7.51]
			Log LME	-8.87 [-4.47]	Log LME	-8.92 [-4.51]
			NLME 5	3.55 [1.89]	NLME 5	3.52 [1.87]
20%	Key	Bias (ME, %)	LME	7.10 [6.68]	LME	7.14 [6.71]
		Precision (RMSE, %)	LME	7.71 [7.12]	LME	8.00 [7.67]
			Log LME	-8.92 [-4.49]	Log LME	-8.86 [-4.46]
			NLME 5	3.51 [1.86]	NLME 5	3.44 [1.79]
50%	Key	Bias (ME, %)	LME	7.09 [6.69]	LME	7.15 [6.72]
		Precision (RMSE, %)	LME	7.79 [7.23]	LME	8.69 [7.76]
			Log LME	-8.94 [-4.48]	Log LME	-8.88 [-4.45]
			NLME 5	3.56 [1.88]	NLME 5	3.58 [1.84]

LME=linear mixed effects, ordinal; log LME=LME, log-transformed both sides; NLME 5=NLME, p < 0.05; NLME 0.1=NLME, p < 0.001. Numbers expressed as troughs [peaks]

Figure 2. Illustrative example of results (50% age effect on CL/F)



Bias and precision in peaks, troughs

- Though the LME method was generally more biased and less precise, overall performance was comparable (Table 3, Figure 2)

Analytical time

- LME was > 15-fold faster in this example

Conclusions

- The LME method may be a more efficient method for elucidating PK covariate effects of clinical interest (i.e. influencing dosing recommendations) using sparse Phase III steady-state exposure data in large numbers of patients
- NLME remains the gold standard but LME is orders of magnitude faster and allows more efficient screening of the covariate space
- Simulation studies, such as the one carried out here are recommended prior to implementation on specific scenarios

Next Steps

- The usefulness of the LME approach will be examined in the context of
 - low- and moderately-accumulating drugs
 - more complex PK models
 - correlations between covariate effects
 - more recent LME implementations
 - alternative residual error models