

Population Pharmacokinetics and Pharmacodynamics as a Tool in Drug Development

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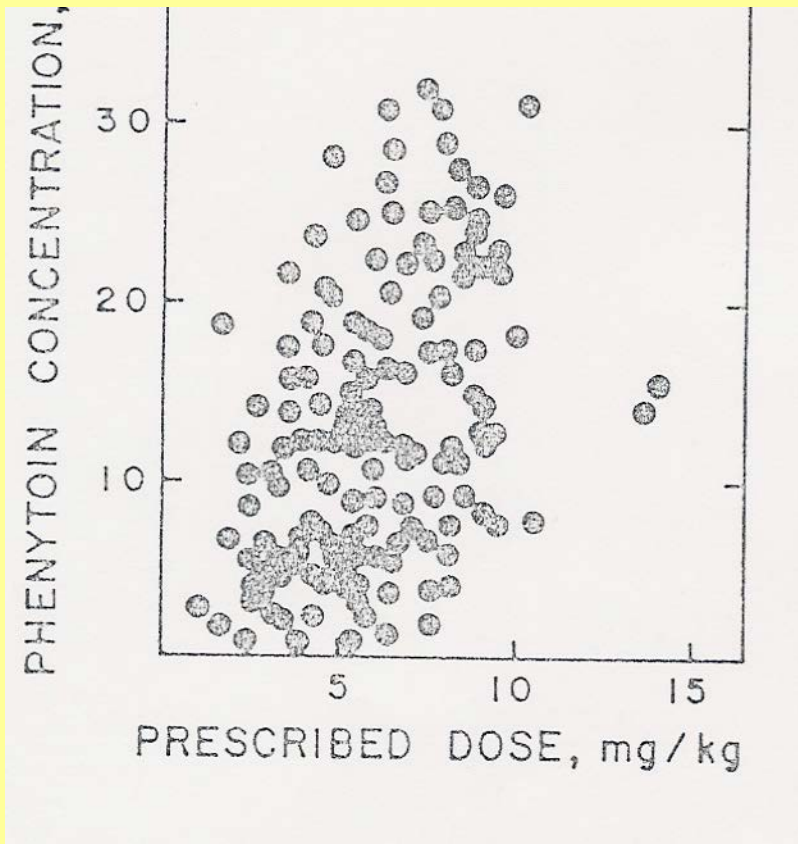
**Pharmacokinetics
and
Pharmacodynamics**

Clinical Pharmacokinetics

Pharmacokinetics

Pharmacodynamics





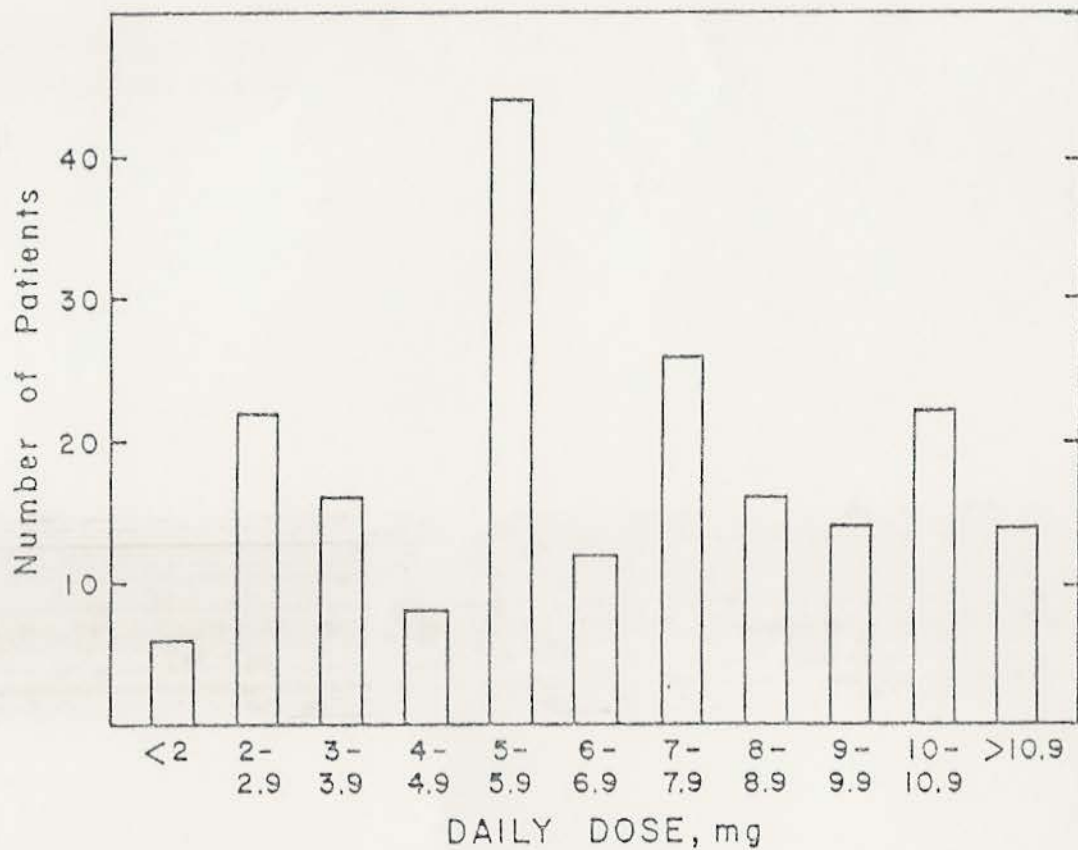


Fig. 8-1. Distribution of doses of warfarin in 200 patients during chronic therapy. (Data from Koch-Weser.¹)

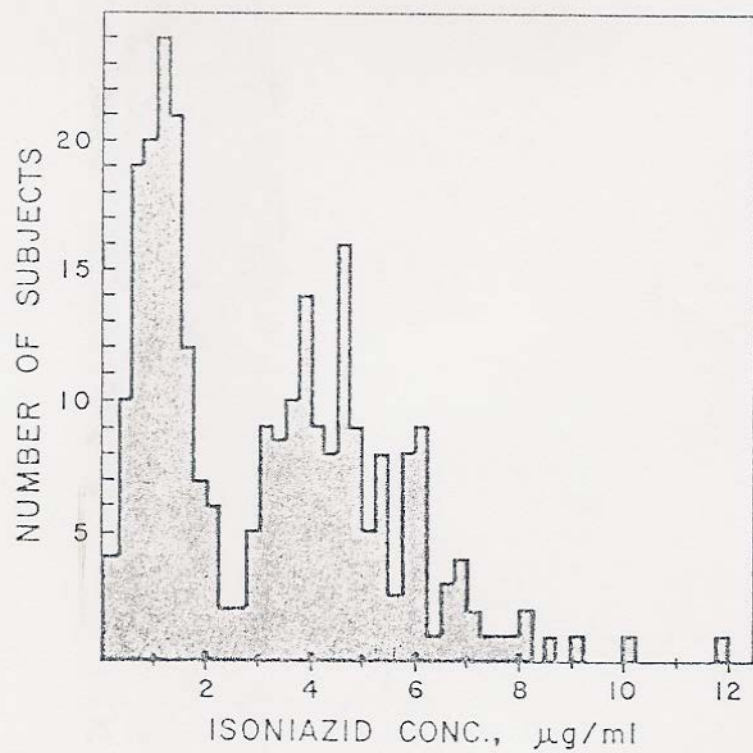
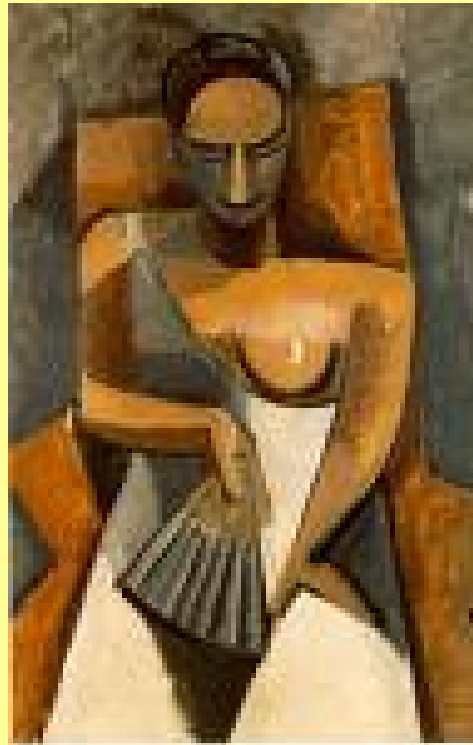
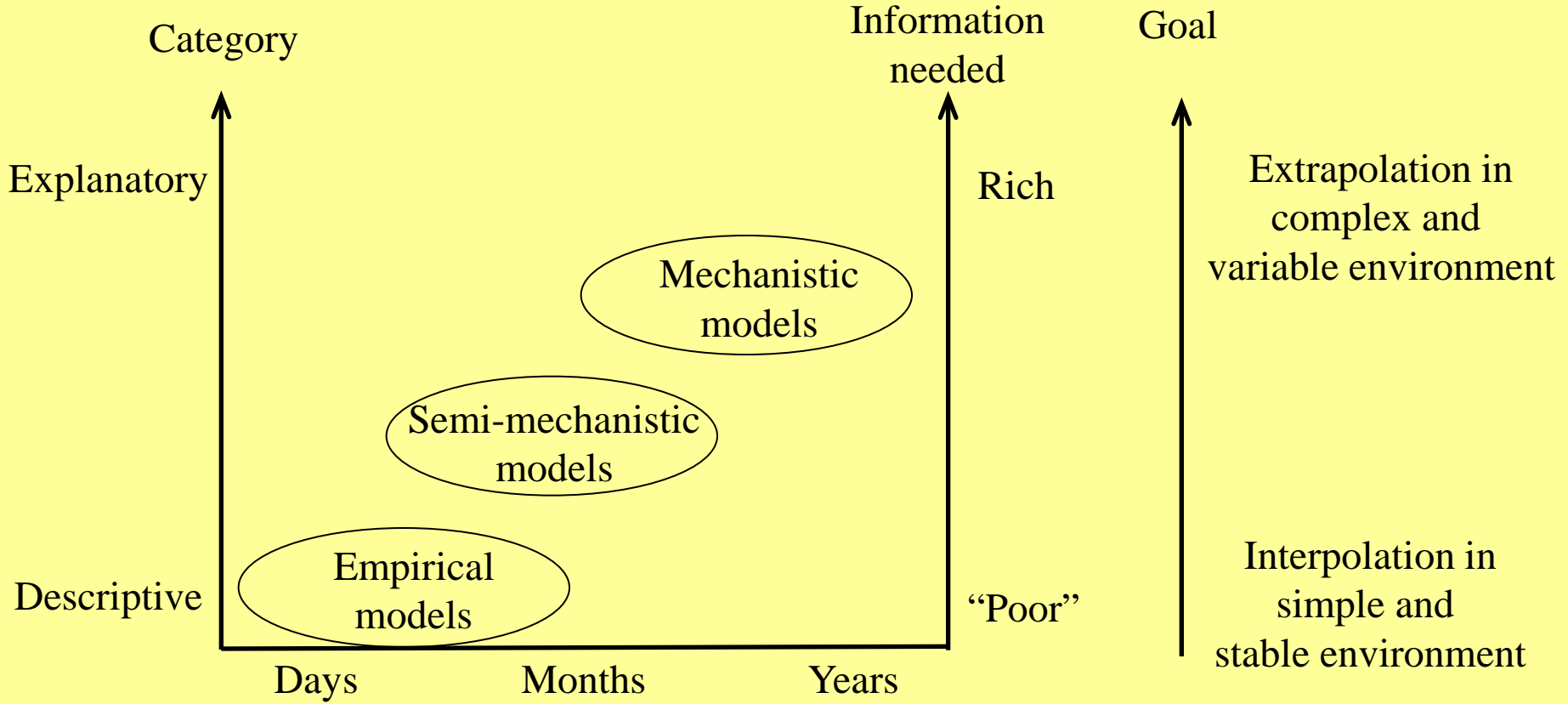


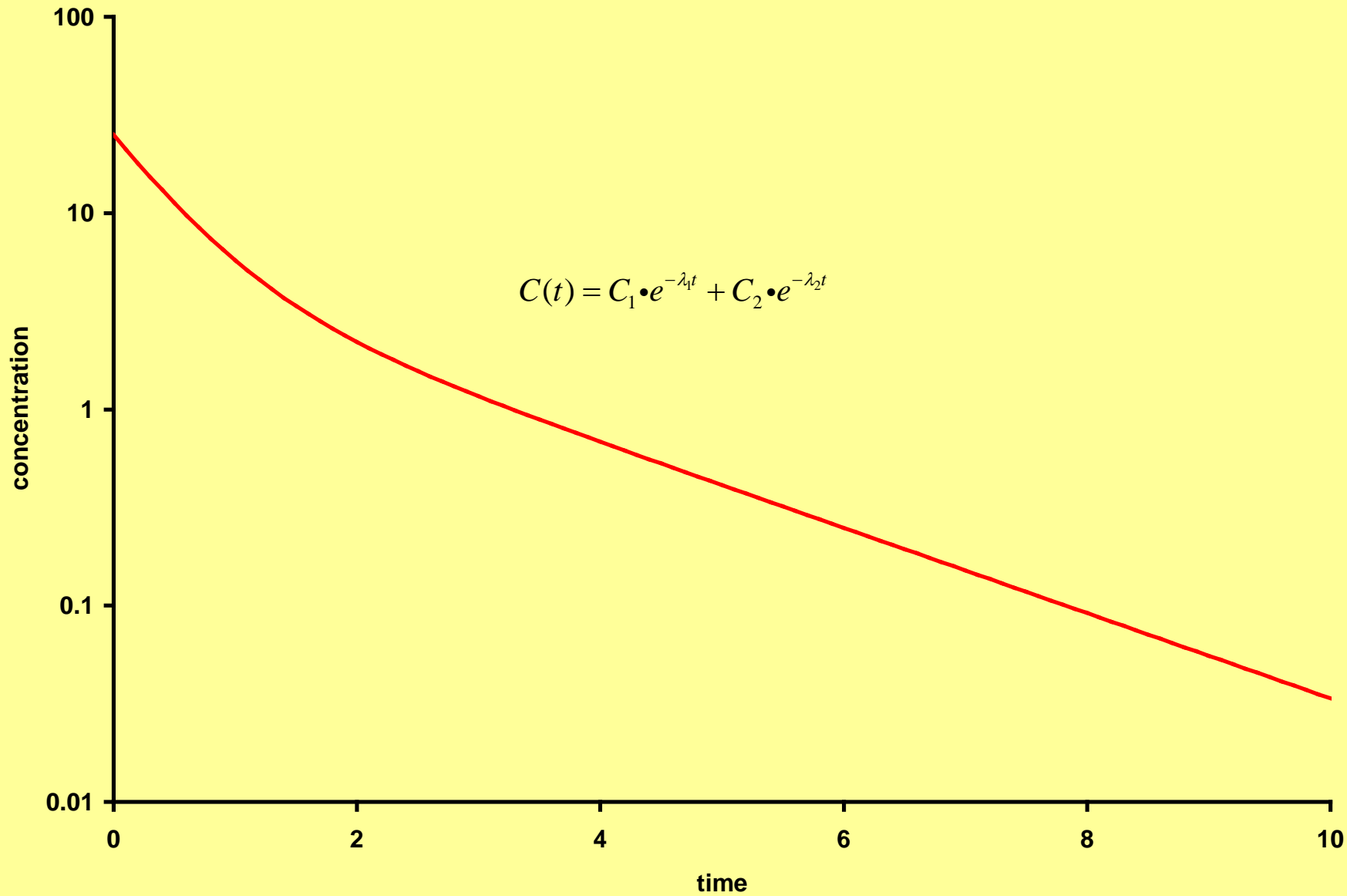
Fig. 7-6. Isoniazid concentration in plasma, 6 hr after oral administration of the same dose to 267 individuals. (Data from Evans, Manely and McKusick.⁵²)

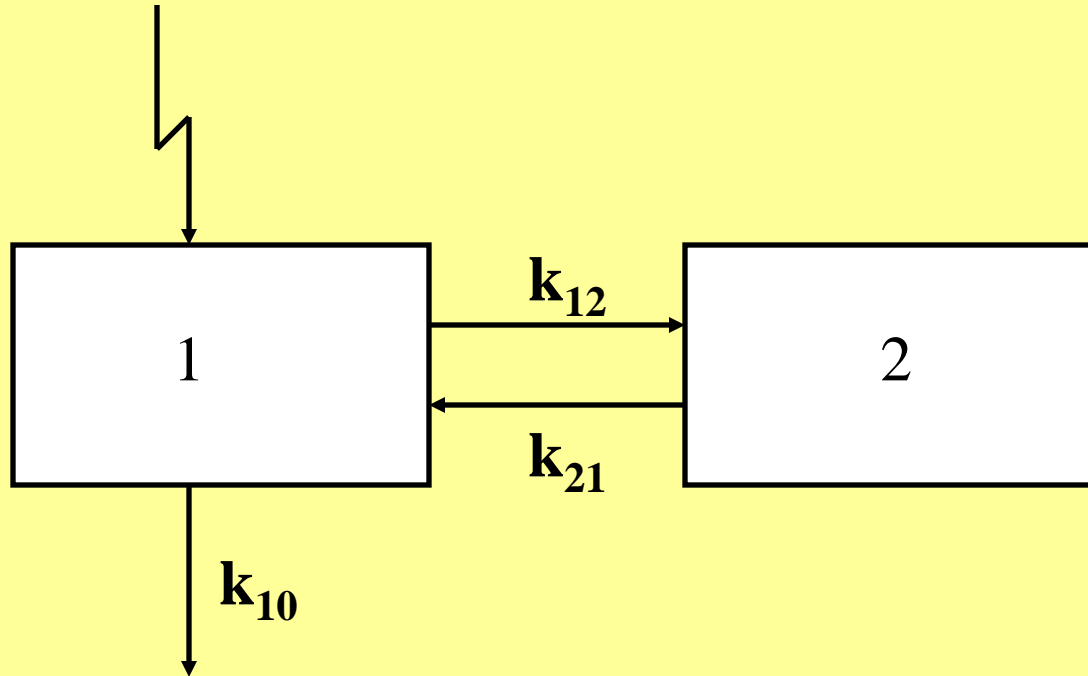
Models



The type of model to be developed should be driven by the available information and the goal of the simulations



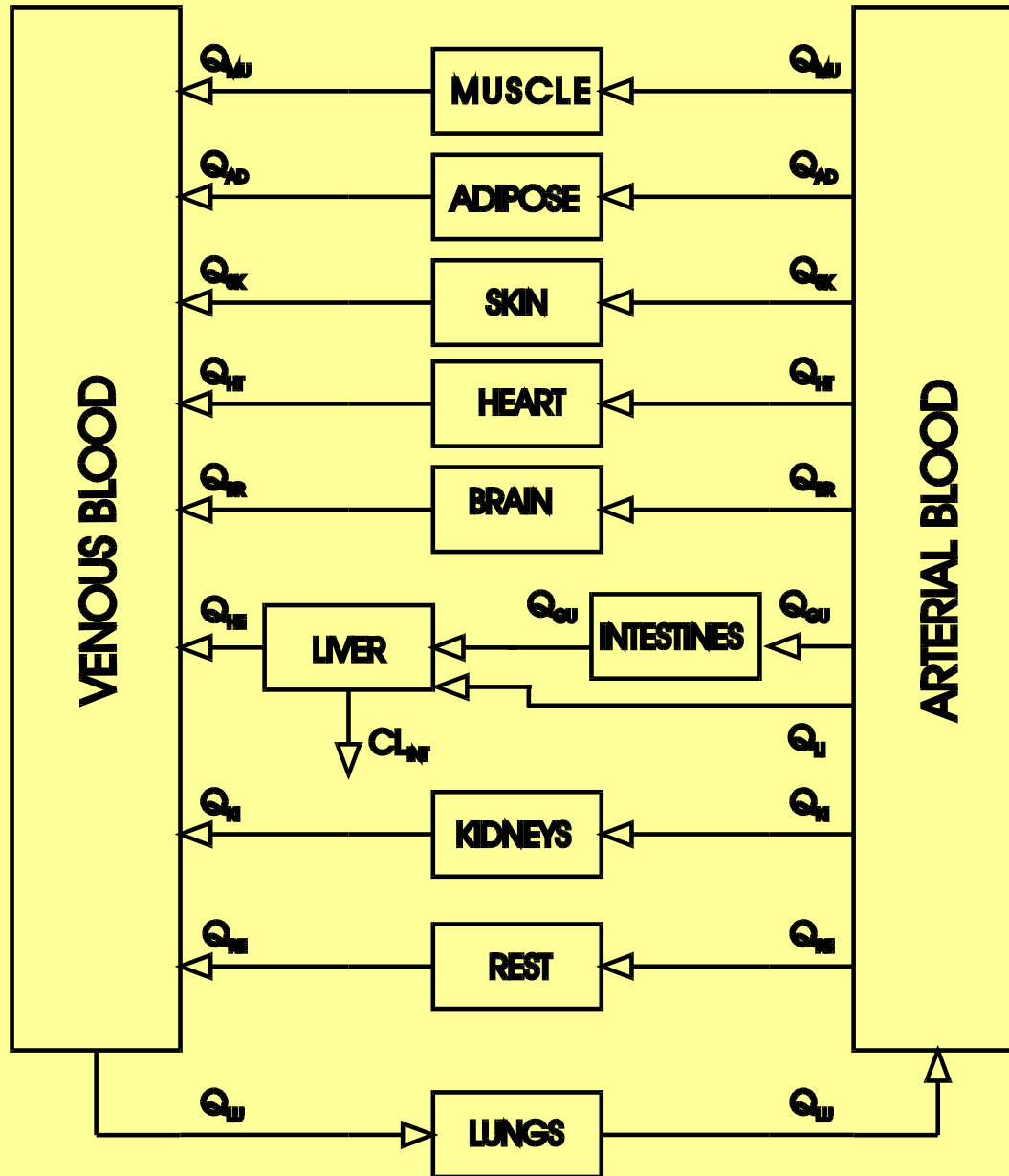




$$V_1 \cdot \frac{dC_1(t)}{dt} = -k_{12}V_1C_1(t) - k_{10}V_1C_1(t) + k_{21}V_2C_2(t)$$

$$V_2 \cdot \frac{dC_2(t)}{dt} = k_{12}V_1C_1(t) - k_{21}V_2C_2(t)$$

PBPK MODEL



$$Kp_H \cdot V_H \cdot \frac{dC_{out,H}(t)}{dt} = Q_H \cdot C_A(t) - Q_H \cdot C_{out,H}(t) - CL_{int} \cdot Kp_H \cdot C_{out,H}(t)$$

Pharmacokinetic Study Design

POPULATION PHARMACOKINETICS/ PHARMACODYNAMICS

Estimating the pharmacokinetic/pharmacodynamic similarity and differences between individuals from measurements of drug levels in biological fluids (often blood) and pharmacological effect of subjects or patients arising from some population of interest

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Classical pharmacokinetic/pharmacodynamic study

- .numbers are small
- .subjects are homogeneous - often volunteers
- .studies are well controlled - experimental
- .sufficient data per individual

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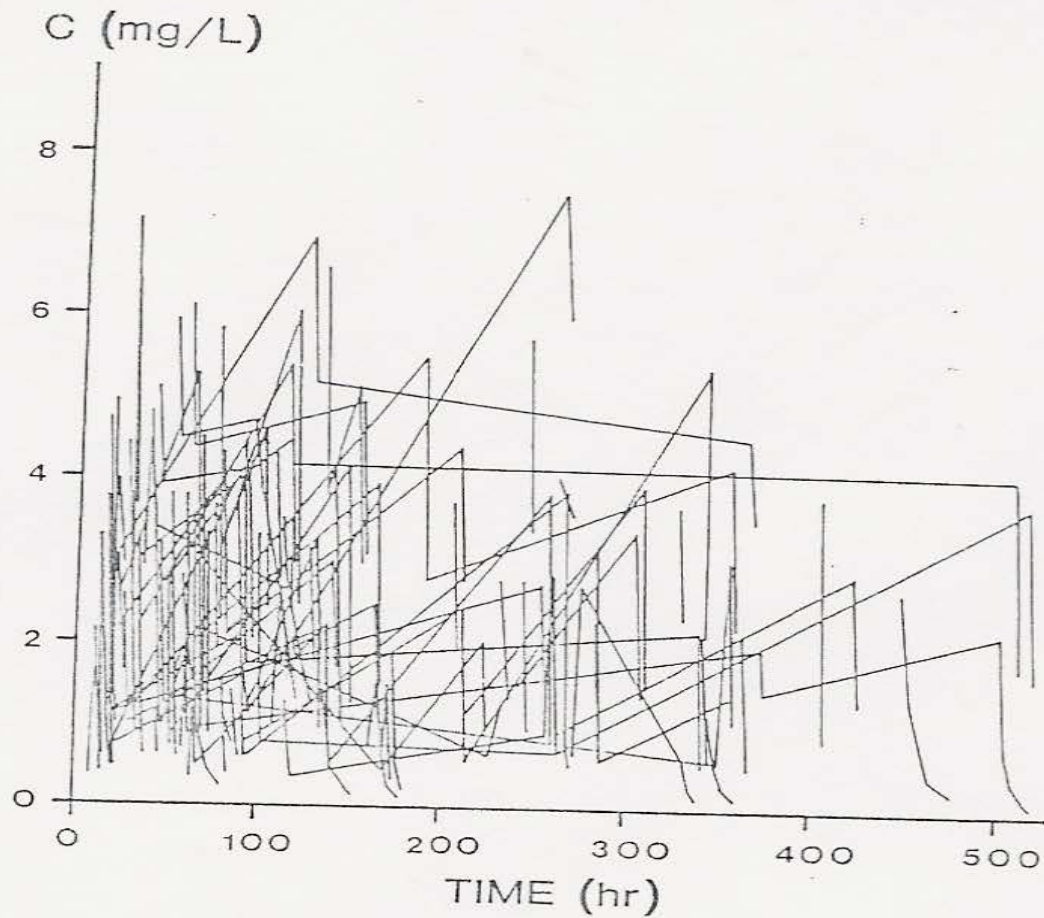
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Population study

- .numbers are large
- .subjects are heterogeneous - often patients
- .study control is difficult - maybe multicentre
observational
- .sparse data

Sparse Data

1731 doses
322 concentrations

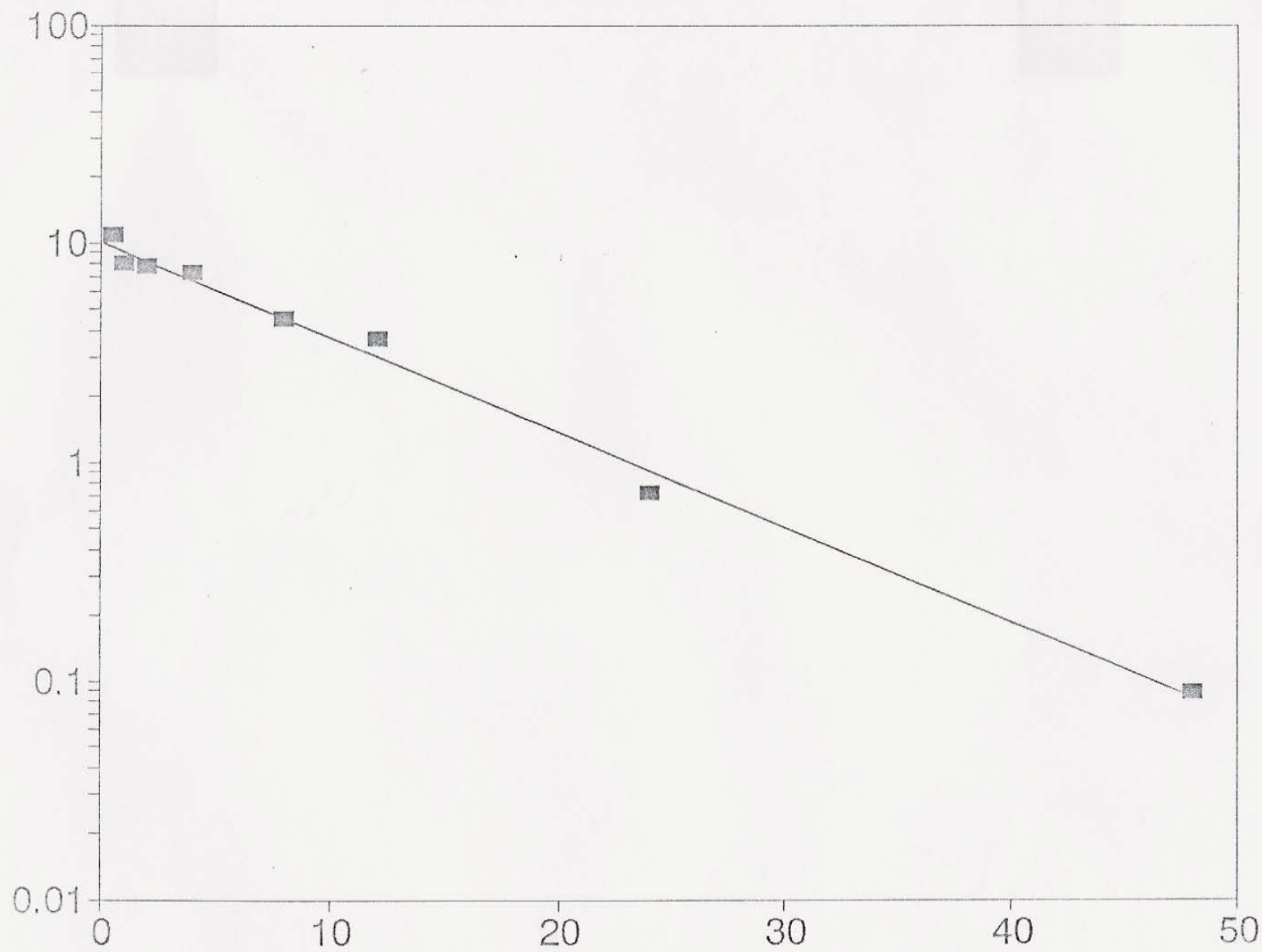


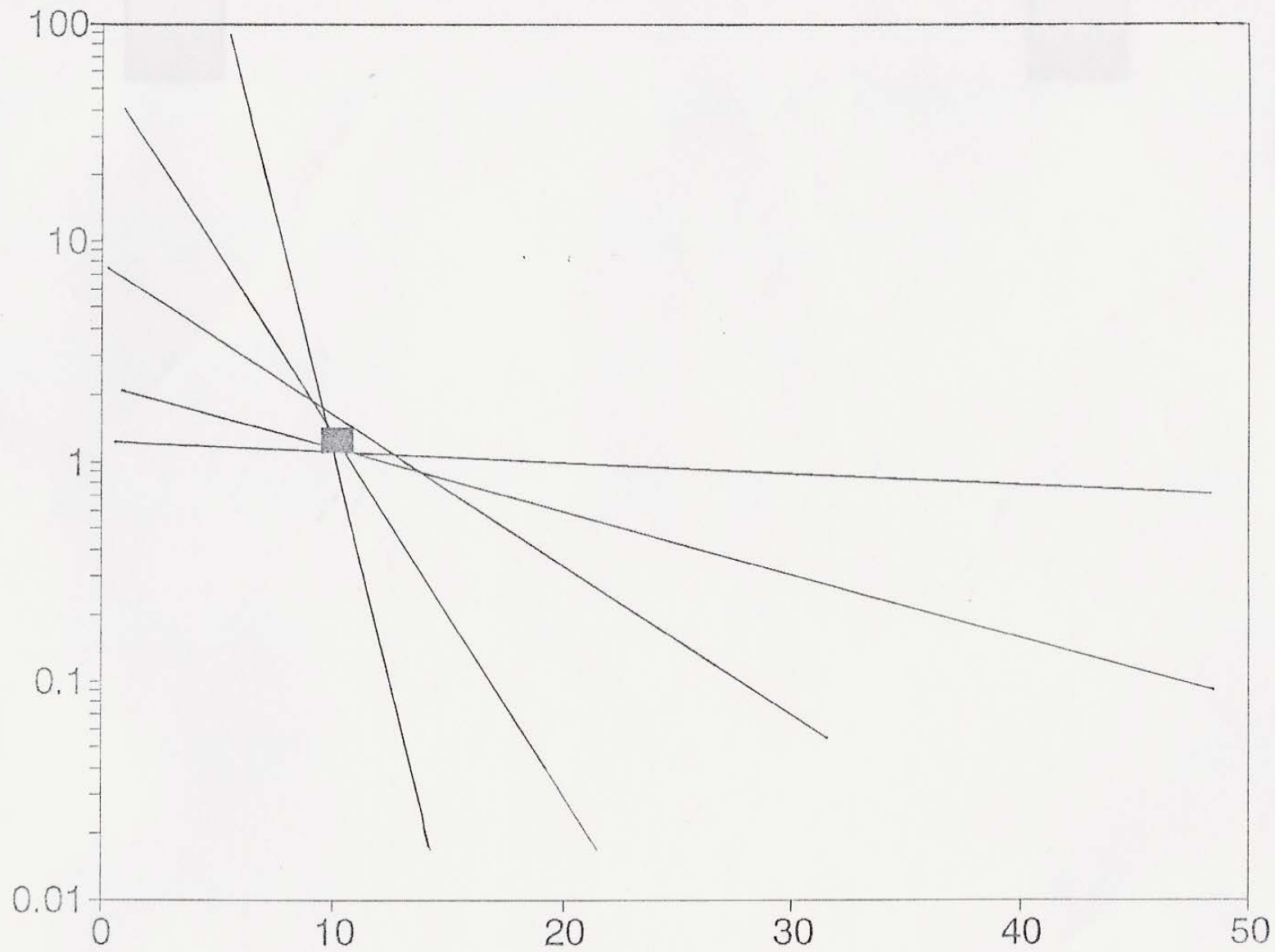
Tobramycin study

Objective: to establish dosage regimen guidelines to maintain maximum efficacy ($C_{max} > 6$ mg/L) and minimum toxicity ($C_{av} < 4$ mg/L) in a majority of patients

Patients:	n	97 (after pruning)
	body weight (kg)	42-120
	age (yr)	16-85
	sex (M/F)	52/45
	creatinine clearance (ml/min)	10-166
	indication	variety of infection

Study design: no design - routine TDM
dosage - 20 to 140 mg every 8 to 24 hr
number of concentrations per individual 1-9 (median 2)
duration of therapy - 14 to 520 hr





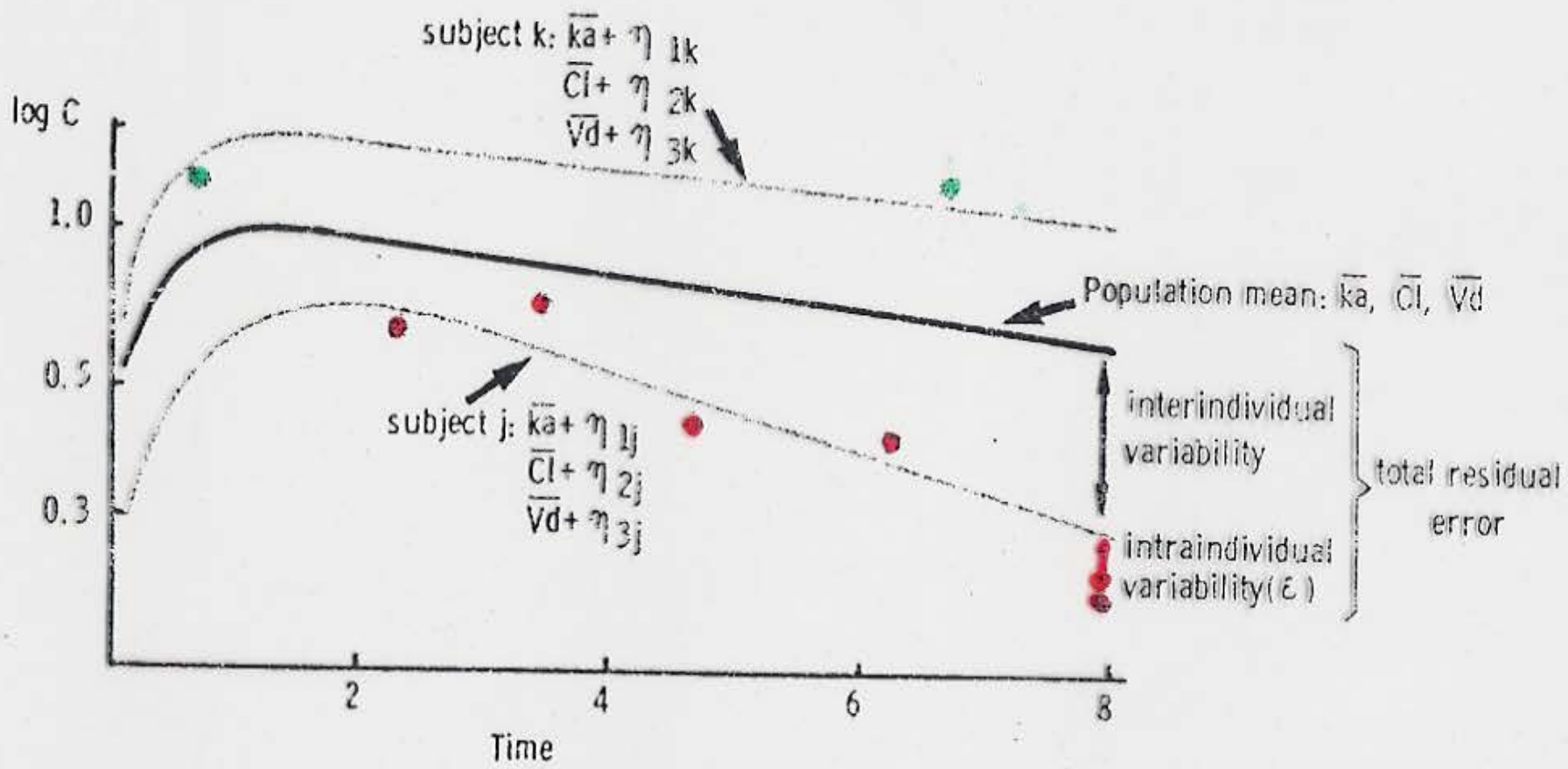
Intelligent Analysis of Unavailable Data

Sum Huan Else, J.Irreprod.Res. 28, 28-29 (1983).

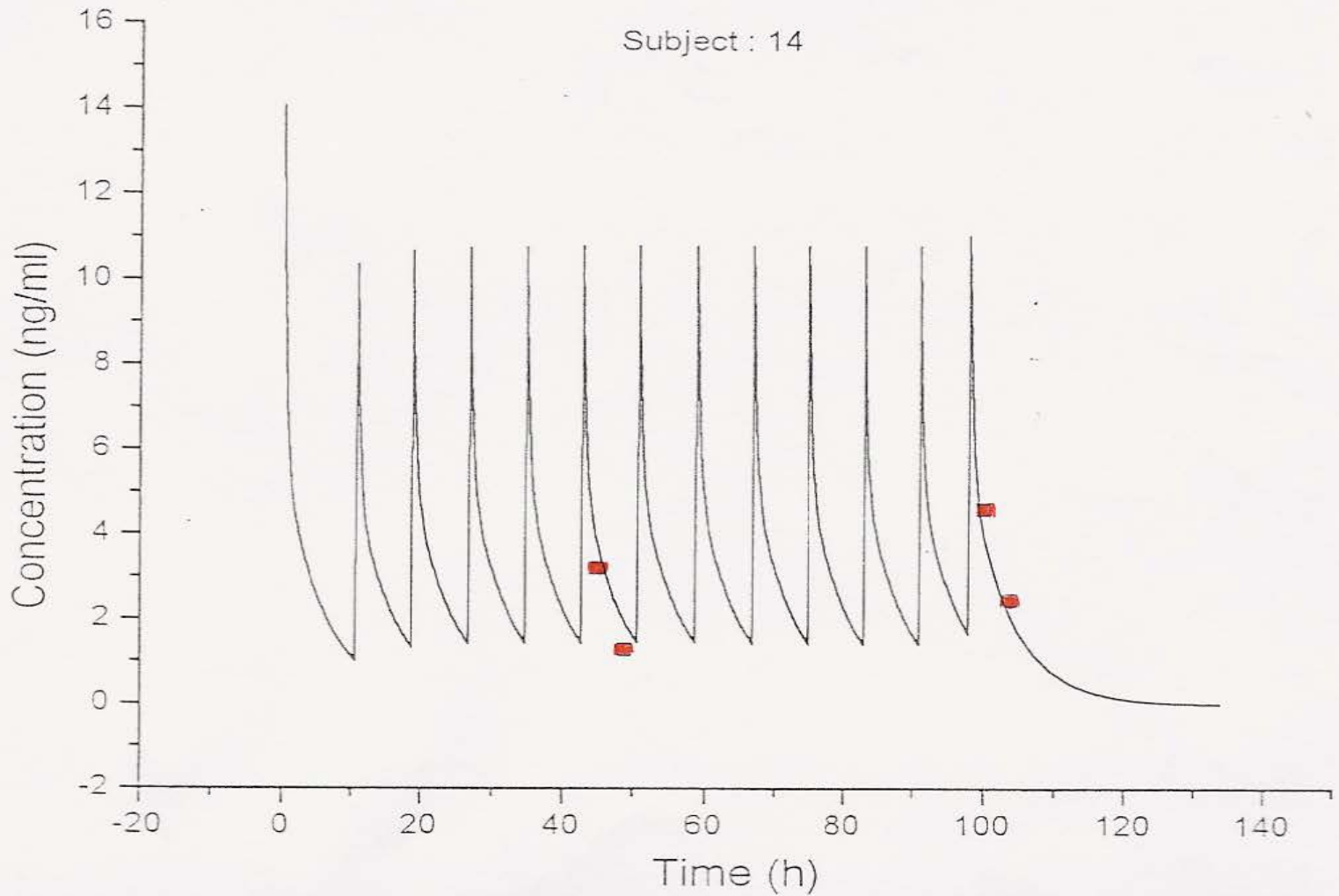
“The inexperienced or naïve analyst might perceive the lack of data to be a minor handicap. The fact of the matter is that a superfluity of data is extremely confining, imposing extreme constraints on the technique, imagination and creativity of the analyst. On the other hand, a lack of data permits full exercise of one’s talents and abilities. The ideal situation is to have absolutely no data available at all”.

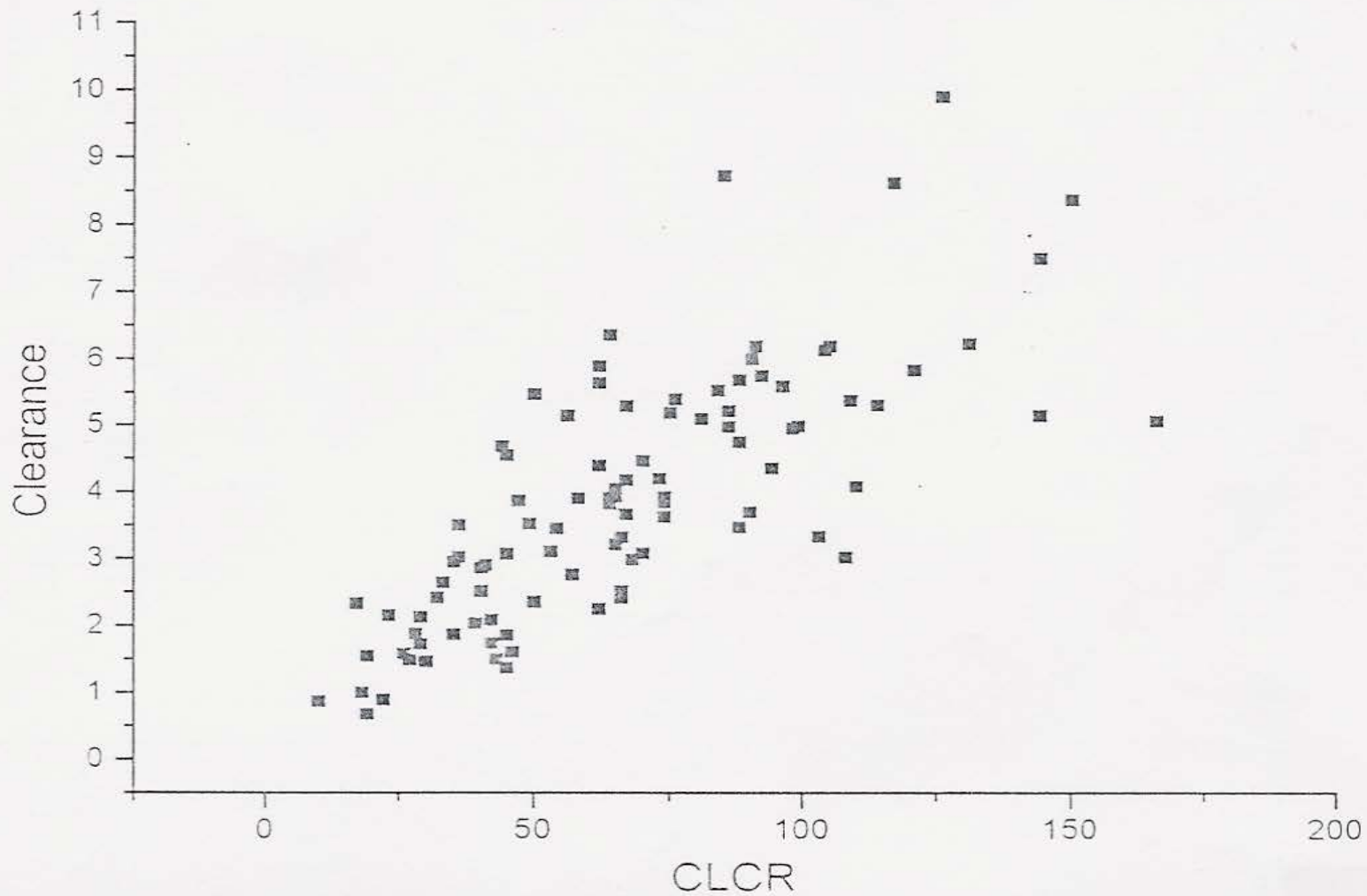
Why?

- **It seeks to obtain relevant pharmacokinetic information in patients who are representative of the target population to be treated with the drug**
- **It recognizes variability as an important feature that should be identified and measured during drug development and evaluation**
- **It seeks to explain variability by identifying factors of demographic, pathophysiological, environmental or drug-related origin that may influence the pharmacokinetic behavior of a drug**
- **It seeks to quantitatively estimate the magnitude of the unexplained variability in the patient population**



Subject : 14





Warfarin study

Objective:

to predict warfarin maintenance dose requirements to achieve a desired degree of anticoagulation based on measurements obtained after a single dose of warfarin

Data:

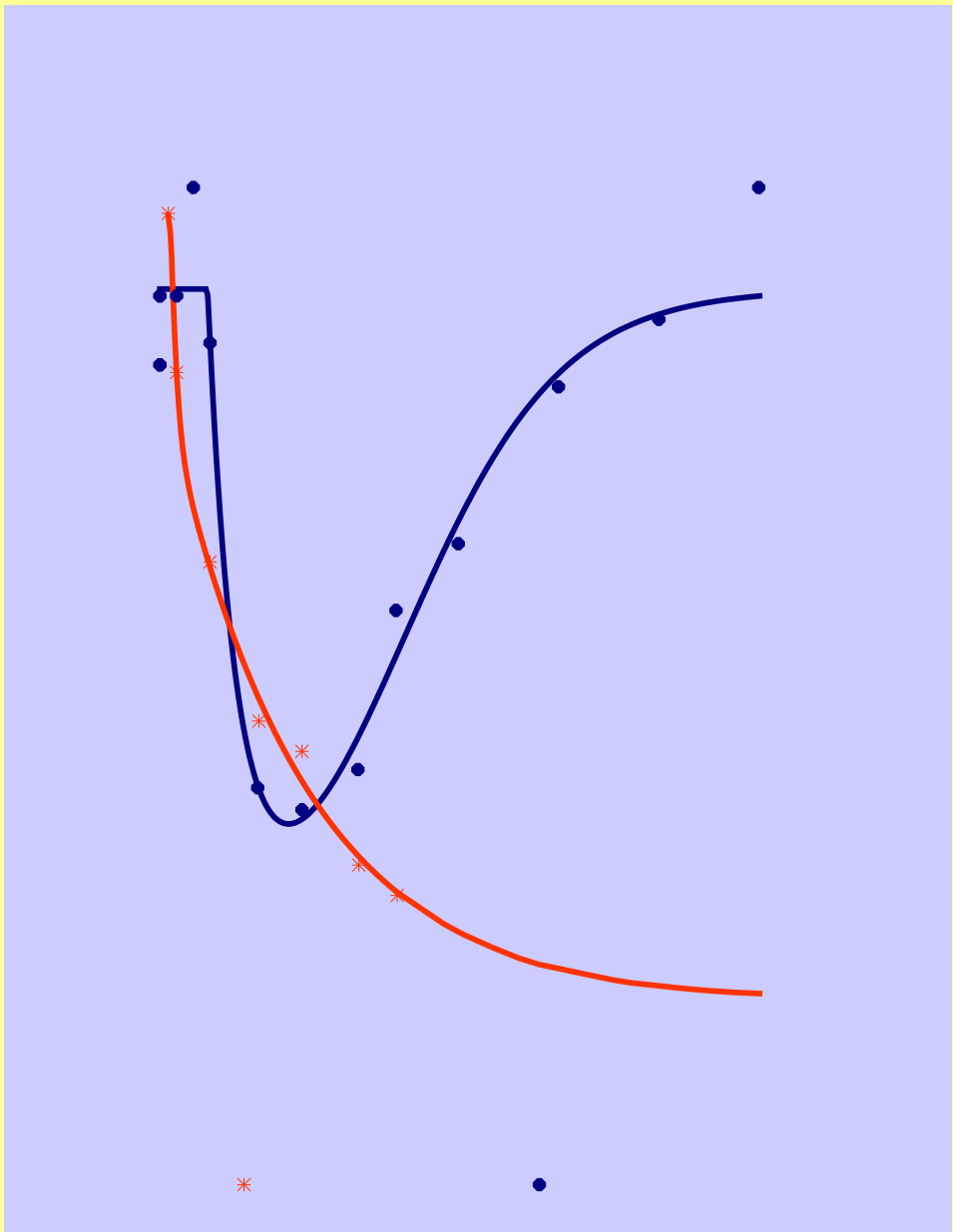
n	48 normal subjects
weight (kg)	66-75
age (yr)	20-27
sex	male

Study design:

25mg single dose of racemic warfarin
14 blood samples (0-168 hr)

Measurements:

R and S warfarin (HPLC)
Prothrombin time (Quick one stage)
Factor VII (chromogenic method)

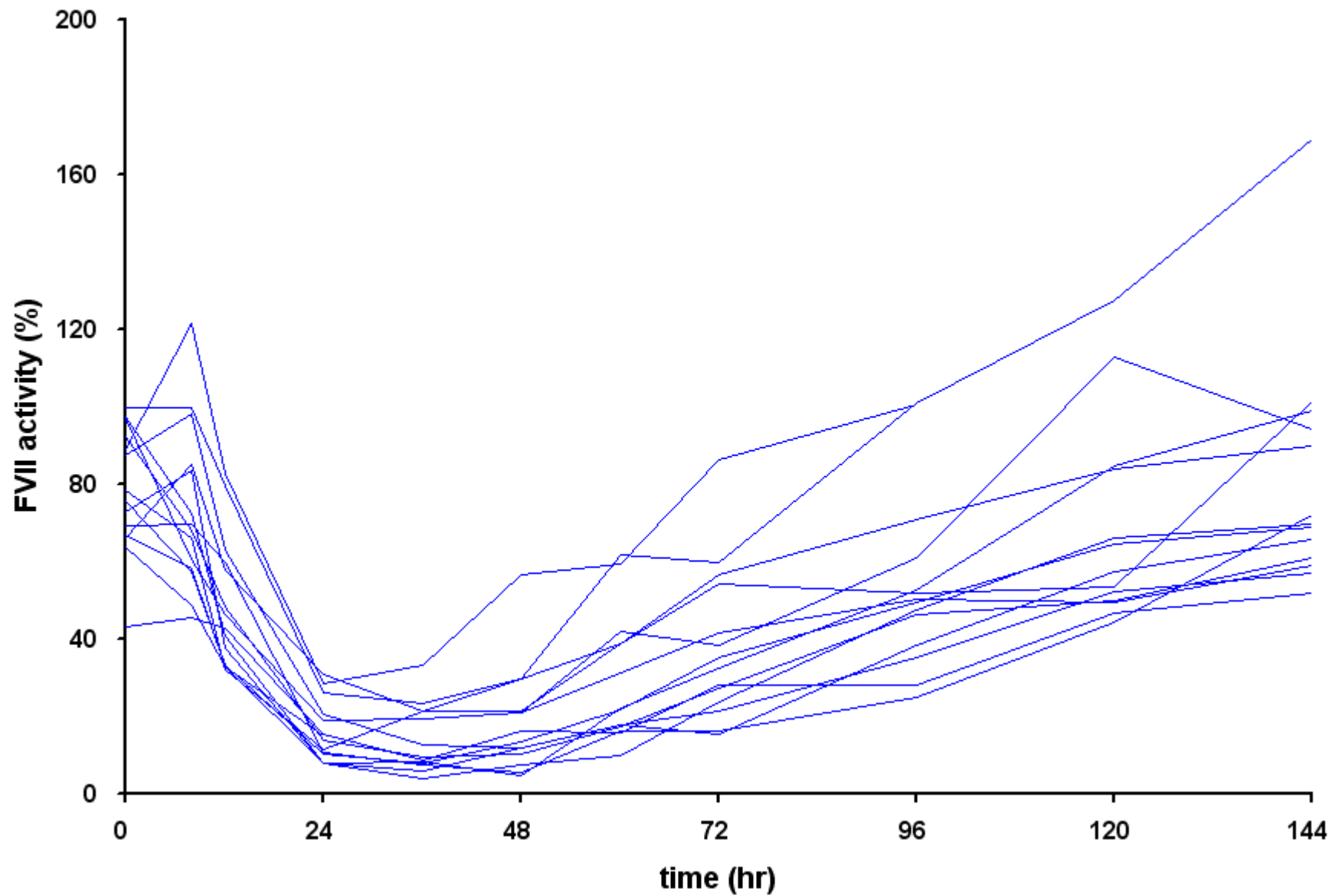


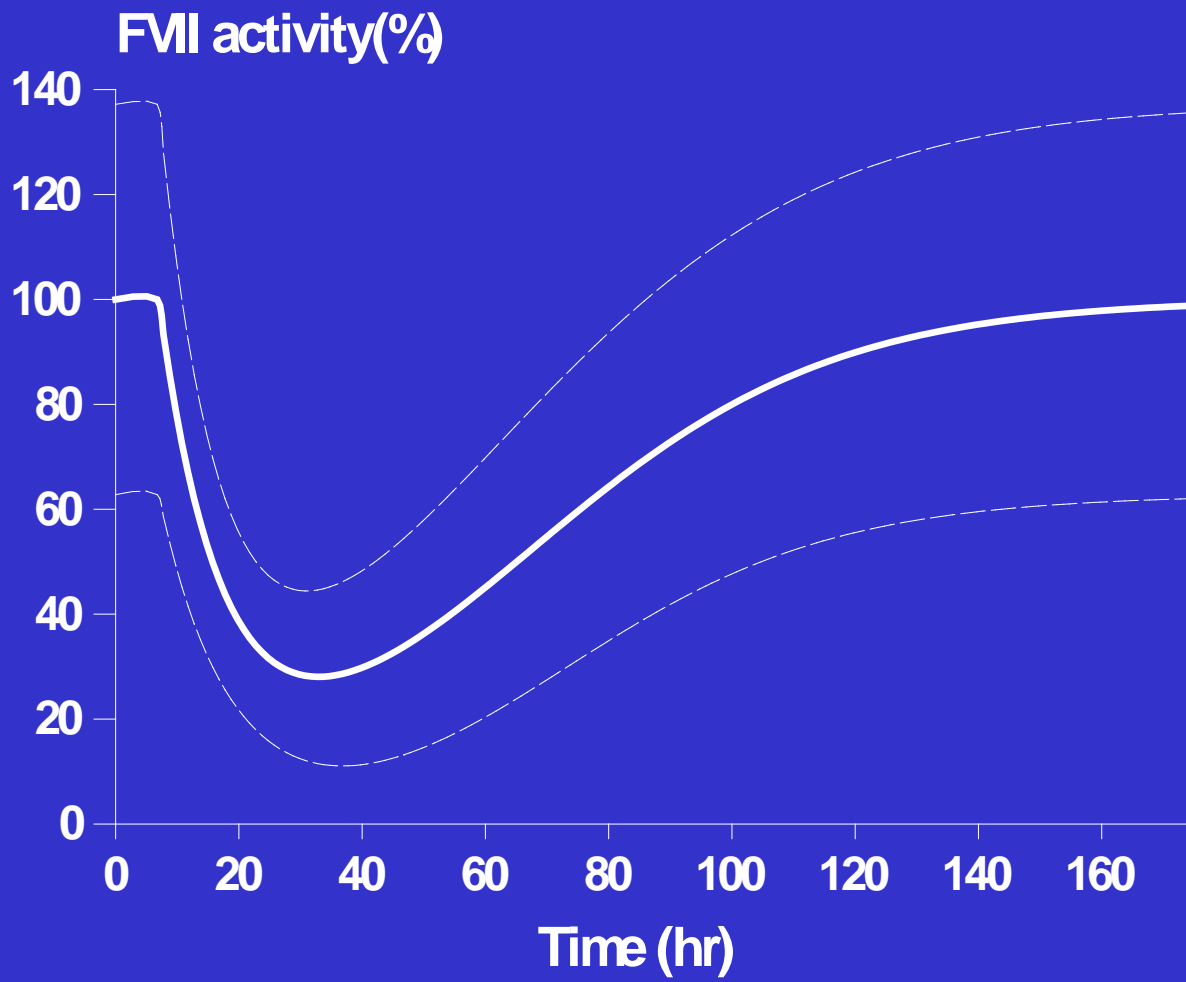
PHARMACODYNAMIC MODEL

rate of change of clotting activity = rate of clotting factor synthesis - rate of clotting factor degradation

$$\frac{dCA(t)}{dt} = k_d \left[\frac{CA_{norm}}{1 + \left(\frac{C_s(t)}{C_{50,s}} \right)^\gamma} - CA(t) \right]$$

k_d = clotting factor degradation rate constant
 CA_{norm} = normal clotting activity
 $C_{50,s}$ = drug concentration giving 50% of maximum effect
 γ = slope factor
 $C_s(t)$ = plasma concentration of s enantiomer





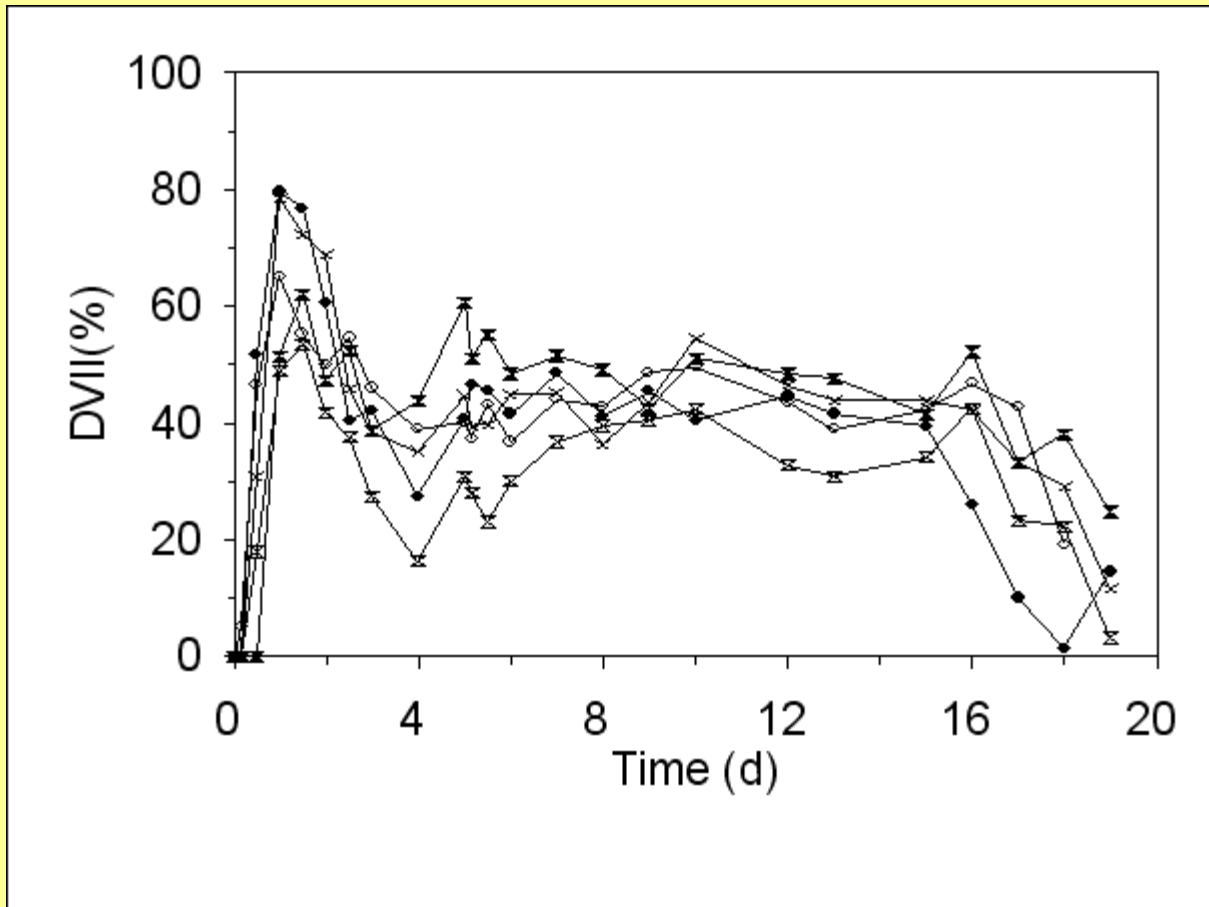
Prospective study

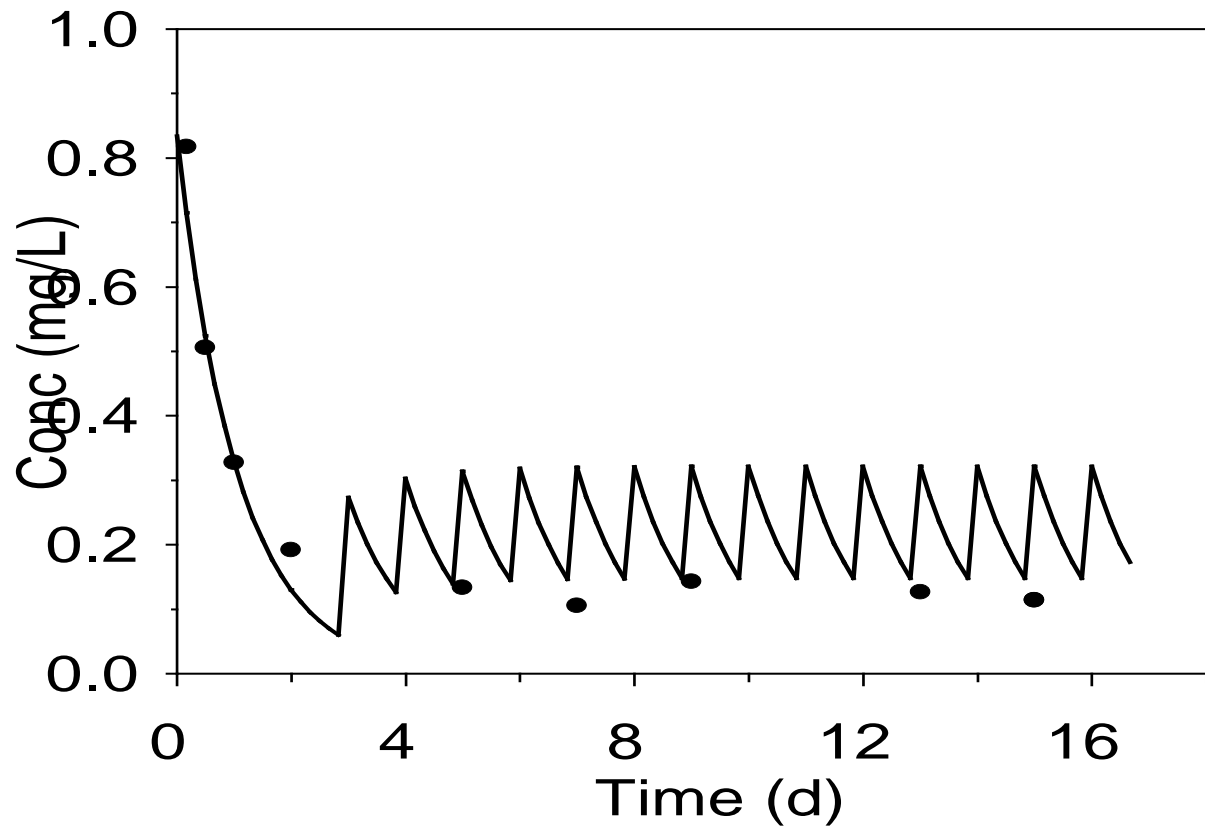
- n** **5 normal male volunteers**
- Study design** **15mg single dose followed by
maintenance dosing from day 3 to
day 16 designed to achieve 50% of
clotting factor activity**
- Maintenance dose** **$DM/\tau = k_s \cdot V_s \cdot C_{50,s}$**

Dose prediction

$$\Phi_i = \sum_{k=1}^{nparam} \frac{(\log \hat{p}_k - \log p_{k,i})^2}{cv(\hat{p}_k)^2} + \sum_{j=1}^{nobs_i} \frac{(\log y_{j,i} - \log f_{j,i})^2}{cv(y_{j,i})^2}$$

pharmacokinetic parameters set to population values



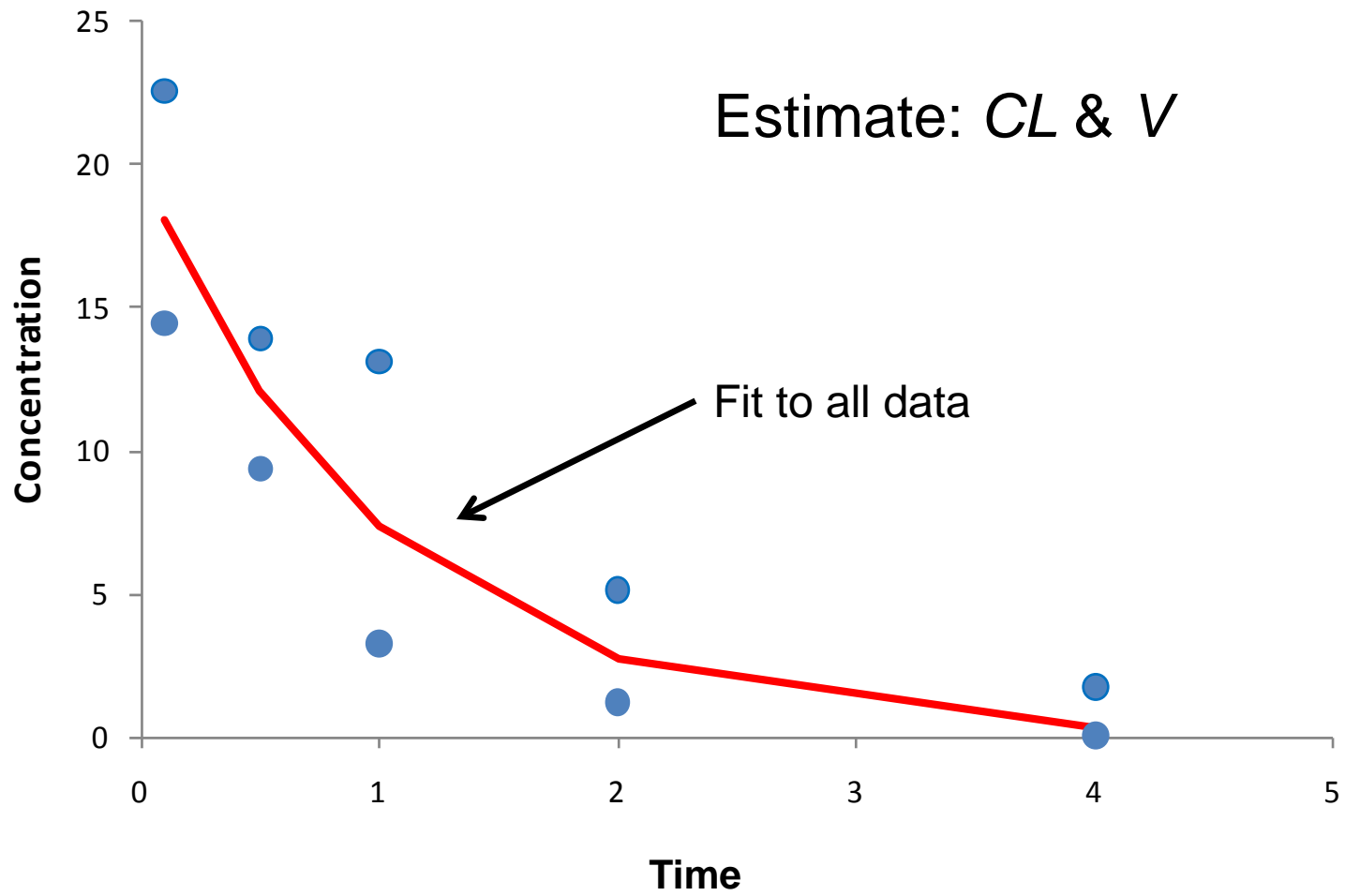


Approaches for population analysis

- Naïve pooled data approach
- 2-stage approach
- A fully population approach

Naïve Pooled Data Approach

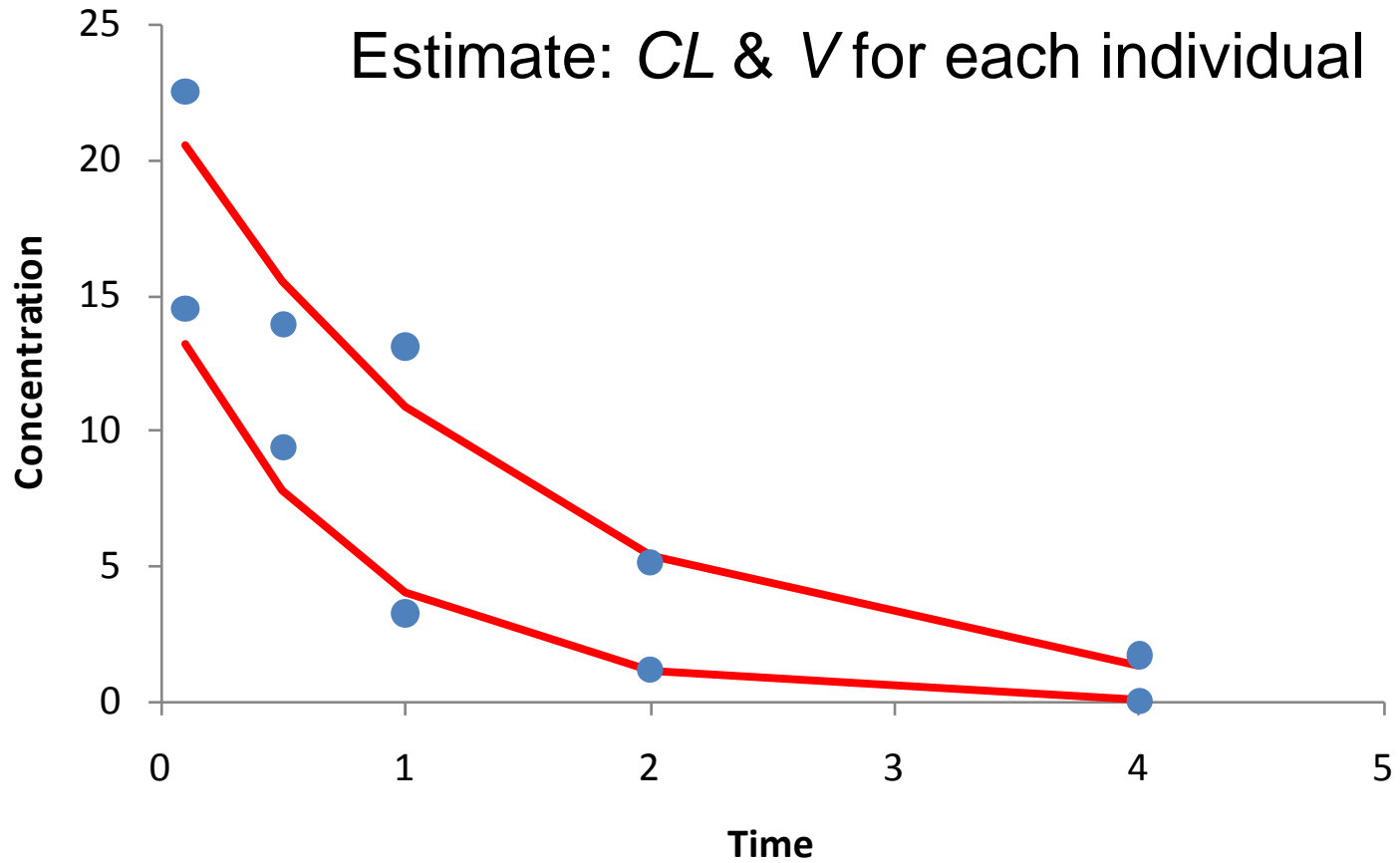
- Estimate parameters for the model assuming all data arise from a single individual
- Can be biased
 - If individuals have different amounts of data
 - If model very nonlinear
- No estimation of variability components
- Influential factors on PK cannot be ascertained



Two-Stage Approach

- The “traditional” approach for PK analyses
- Model the PK of each subject’s data (Stage 1)
- Obtain summary statistics, e.g. mean value of clearance (Stage 2)
 - Between-subject variance is overestimated because estimation of the parameters includes residual error
 - Assumes all individuals contribute equally
 - Is not helpful for identifying sources of variability (e.g. renal function)
 - Essentially unusable where there are “sparse” data

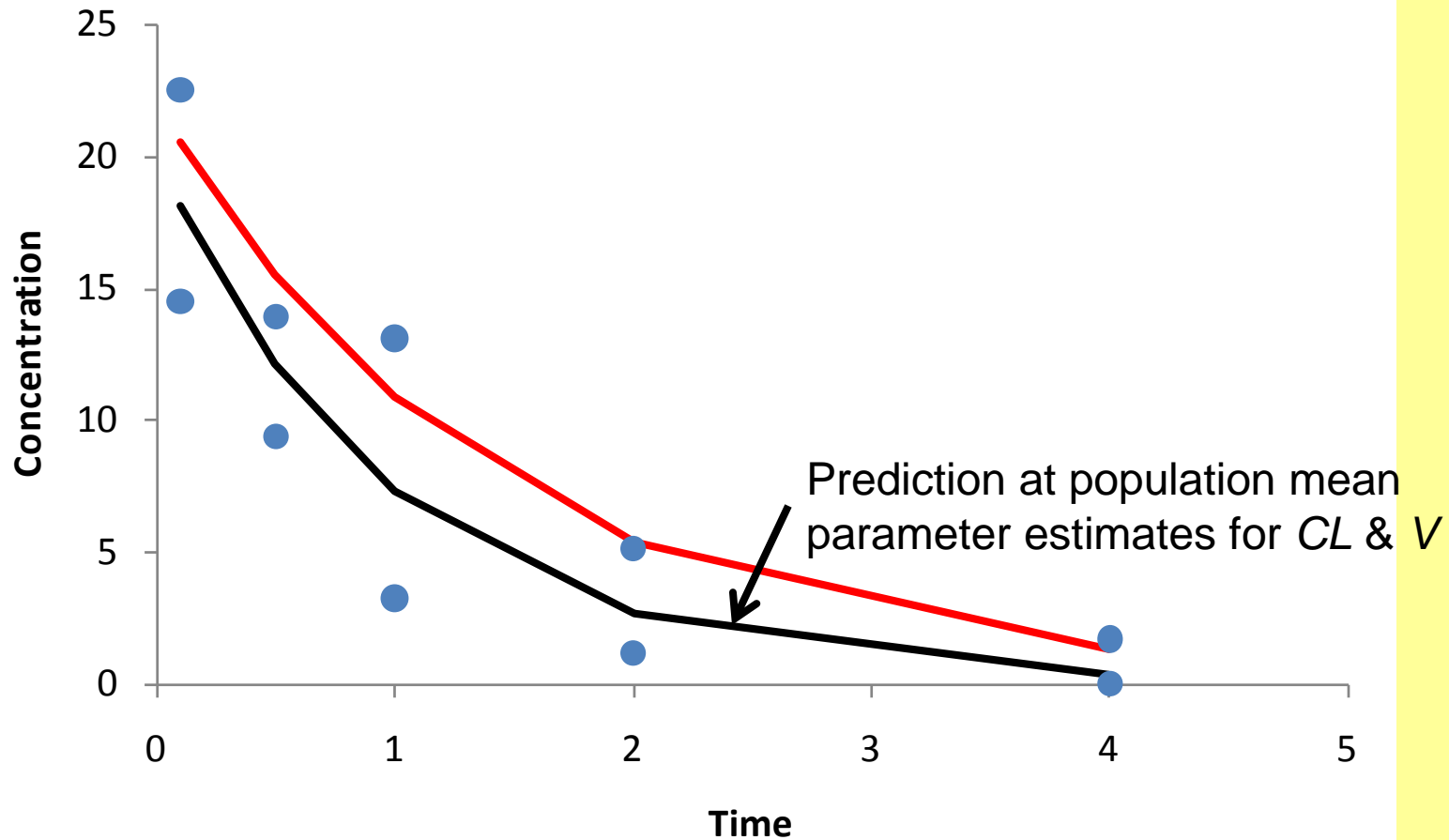
Two-Stage Approach



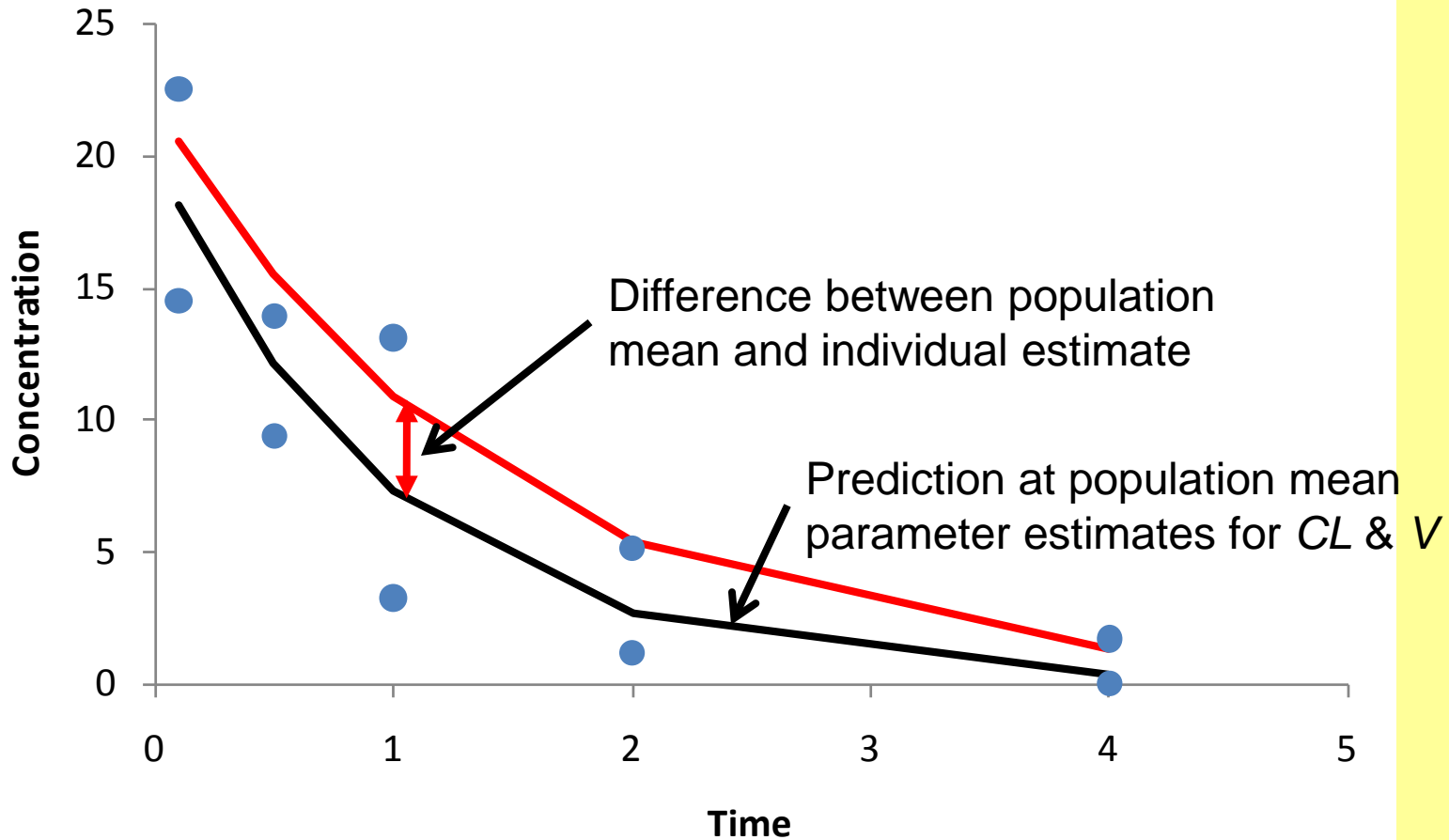
The population approach

- Estimate the overall mean parameters and the variability between individuals as well as the residual variability
- The “standard” population analytical approach
- Sparse/rich, unbalanced/unstructured data
- Considers the **population** (rather than the individual) as the unit for the PK analysis
- “Individuality” of the subjects is maintained

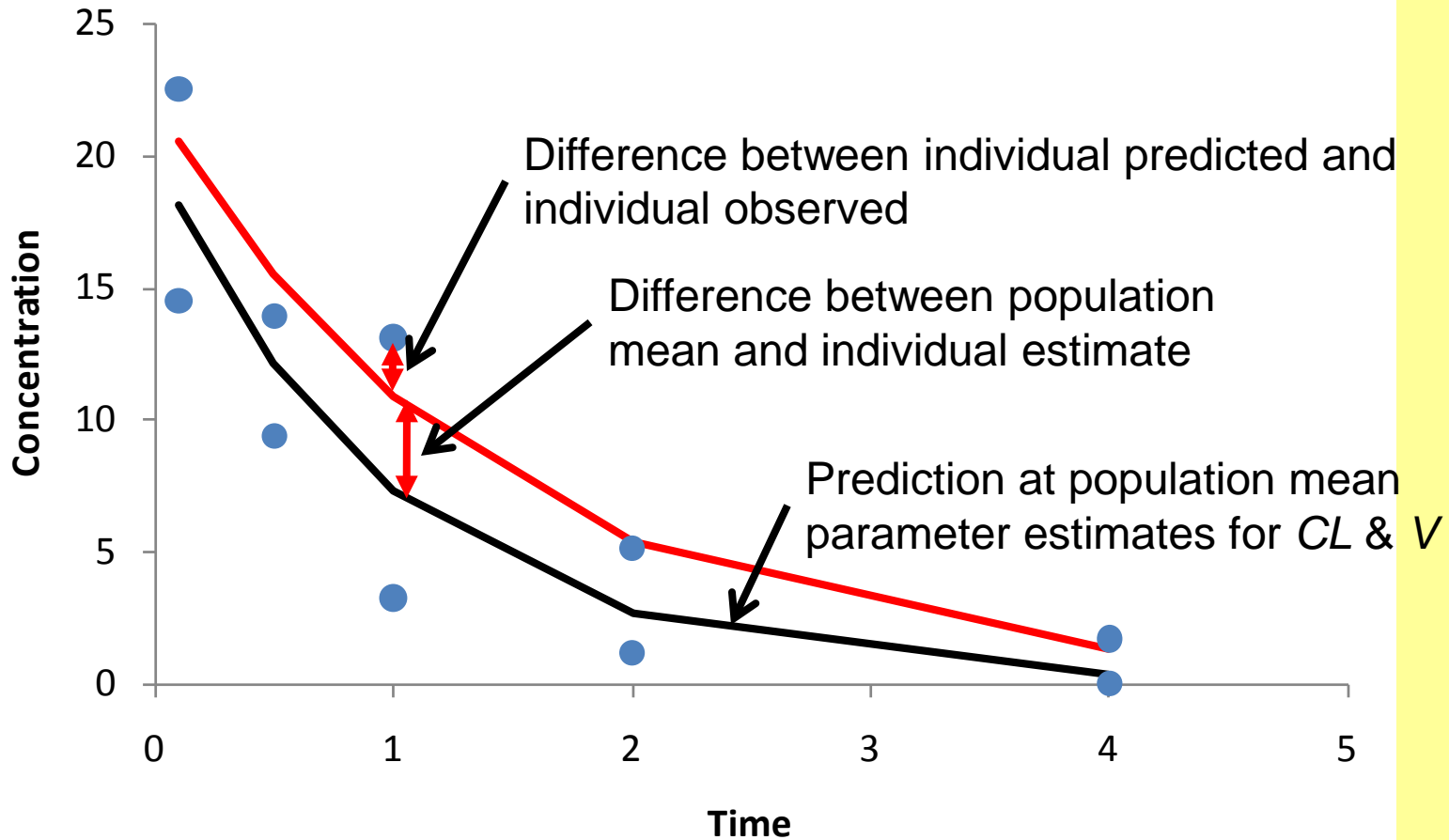
The population approach



The population approach



The population approach



Variability

- Two levels
 - Variability between people (heterogeneity)
 - Uncertainty about the observations (residual unexplained variability)

Heterogeneity

(Between Subject Variance [BSV])

- There are two sources of heterogeneity between patients
 - Predictable = BSVP
 - Unpredictable (random) = BSVR
- Predictable variability can arise from (examples):
 - differences in renal function
 - differences in body size and composition
- The population parameter variability (PPV) in the population is therefore the sum of these two sources:

$$PPV = BSVR + BSVP$$

Residual unexplained variability

[RUV]

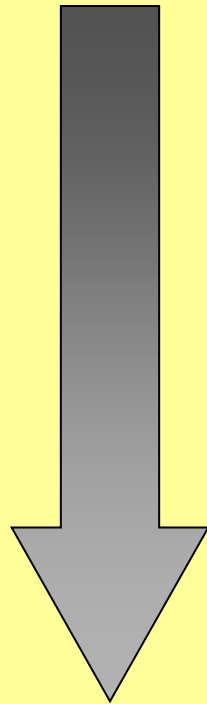
- This is the variability of the observed concentration around the model predicted
- The observed will seldom equal the model predicted due to many process such as assay error

Nonlinear mixed effects modelling

- Nonlinear: almost (if not all) PK and PKPD models are nonlinear in the parameters
- The population approach is interested in the mean parameter values (“fixed effects”) as well as the variability between and within individuals (“random effects”)
- Combining fixed and random effects is termed “mixed effects”
- Essentially all population PK studies are therefore examples of nonlinear mixed effects modelling

Software for Population Analyses

Decreasing
Market Share



- **NONMEM (\$\$)**
 - Monolix (free/\$\$)
 - WinBUGS (free)
 - SAS proc nlmixed (\$\$\$)
 - Phoenix nlme (free/\$\$\$)
 - S-ADAPT MC-PEM (free)
 - NPEM (free)
 - P-Pharm (\$?)
 - NPML (dead?)
- } Not used to any extent

Role of Modelling and Simulation in Phase I Drug Development

L. Aarons, M. Karlsson, F. Mentré,
F. Rombout, J.-L. Steimer, A. van
Peer

COST B15

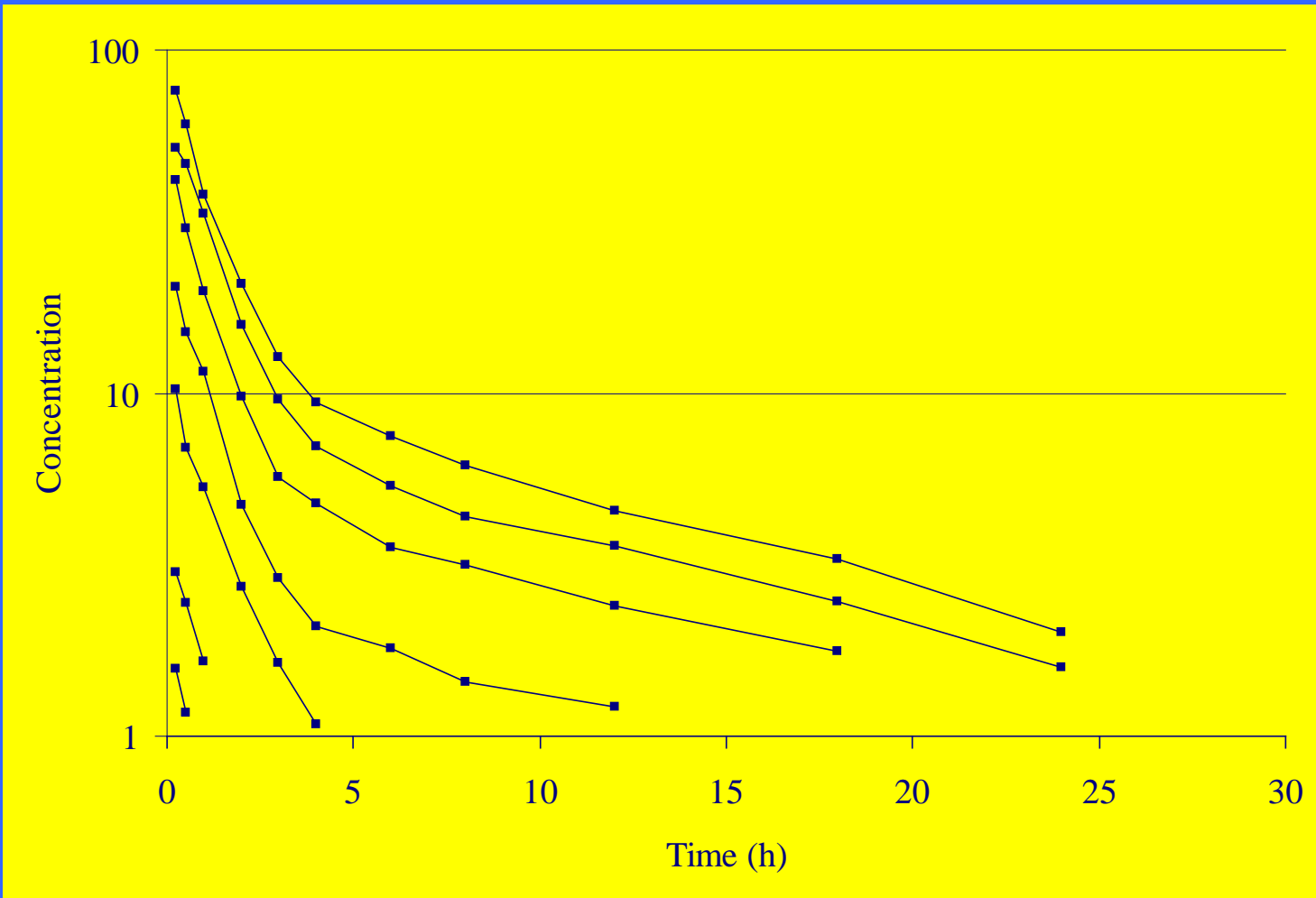
Brussels January 10-11, 2000

When is the population approach useful?

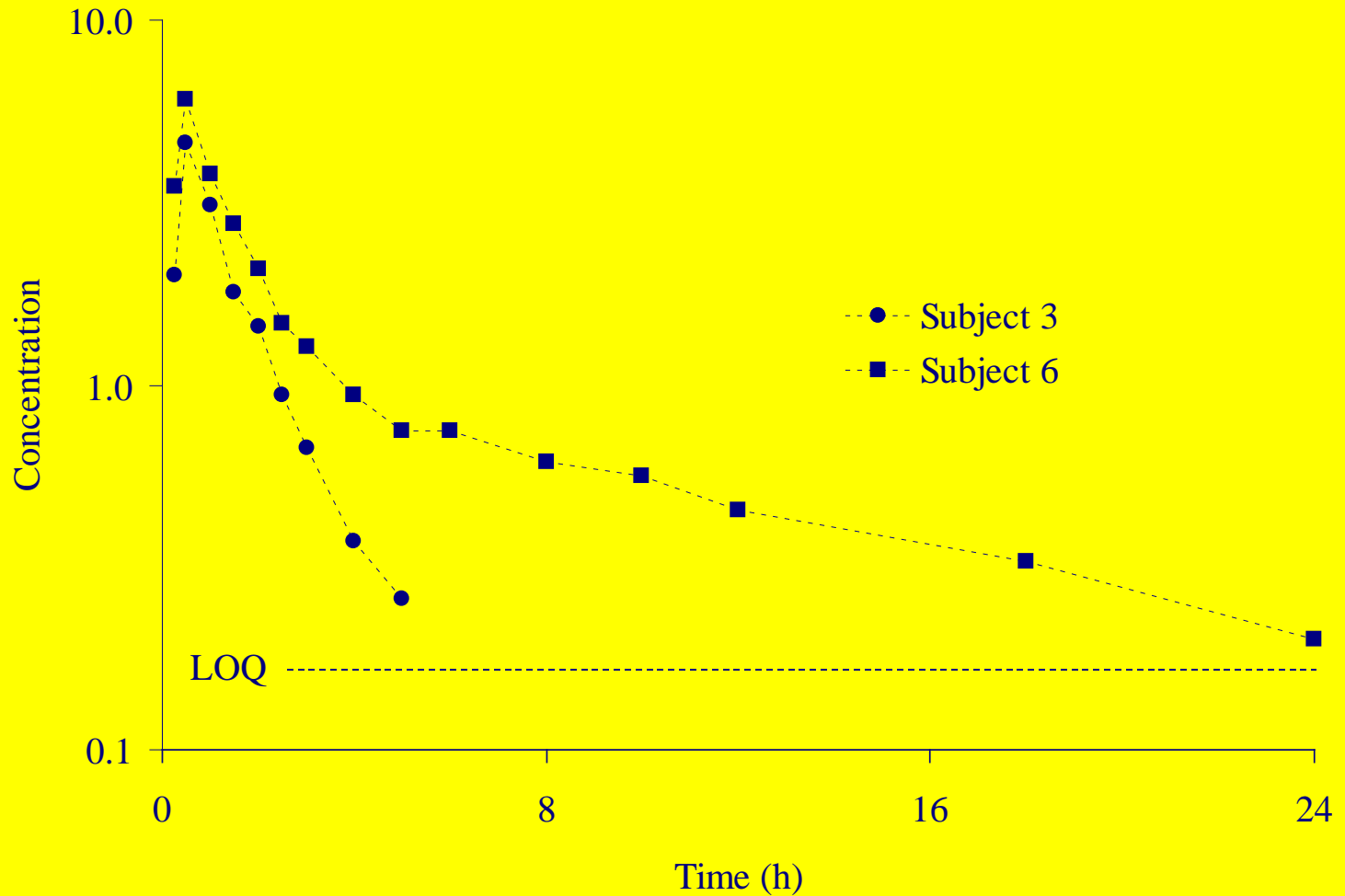
Essentially it is (almost) always useful

- In particular if there are few observations per patient e.g.
 - Outpatients or phase III studies
 - Intensive and emergency care
 - Elderly
 - Children, including premature infants
- When you want to learn about sources of variability
- When you want to develop a model to predict future patient responses
- When you want to use the model to simulate various dosing scenarios
- ...

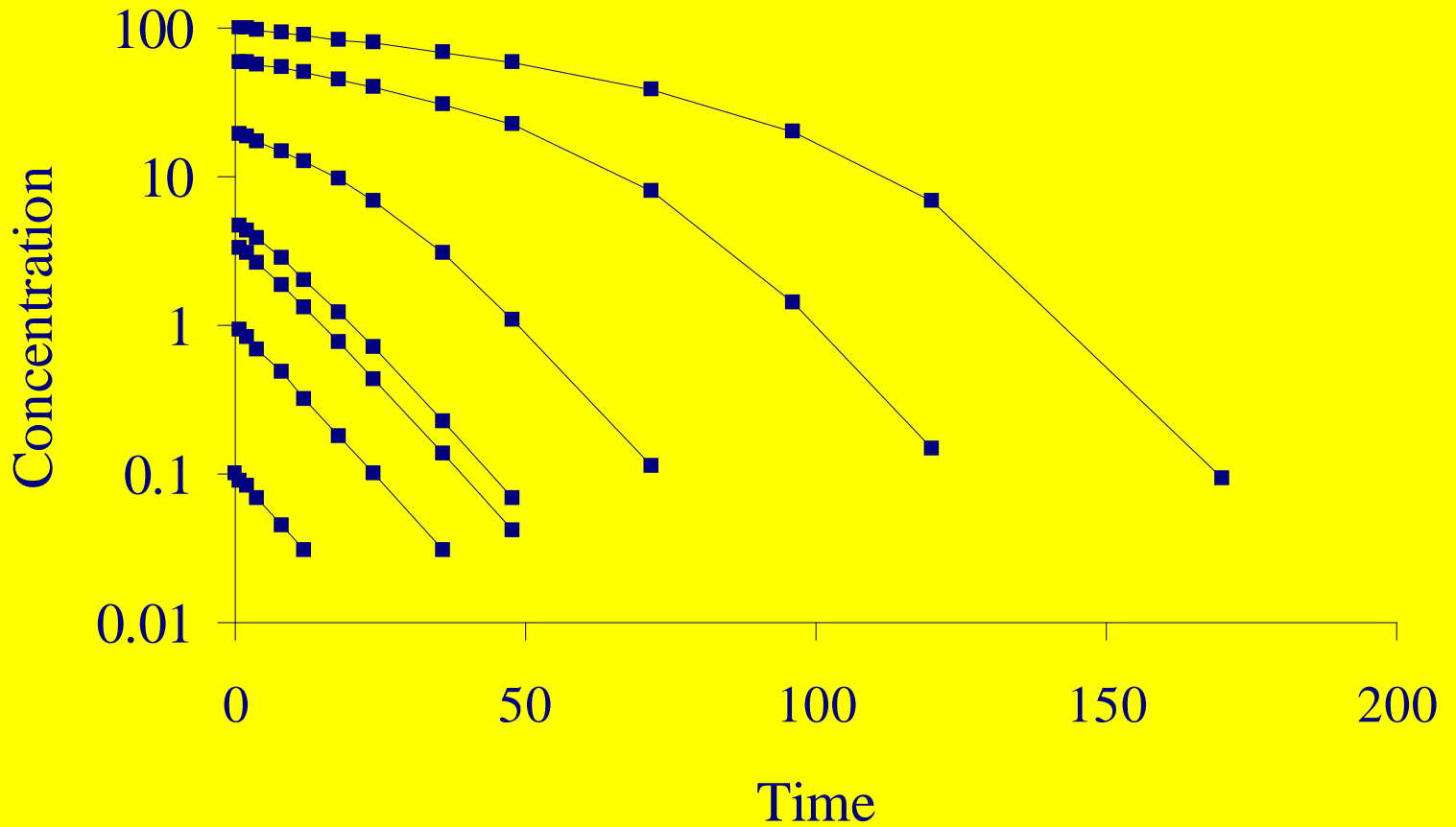
Tolerability studies



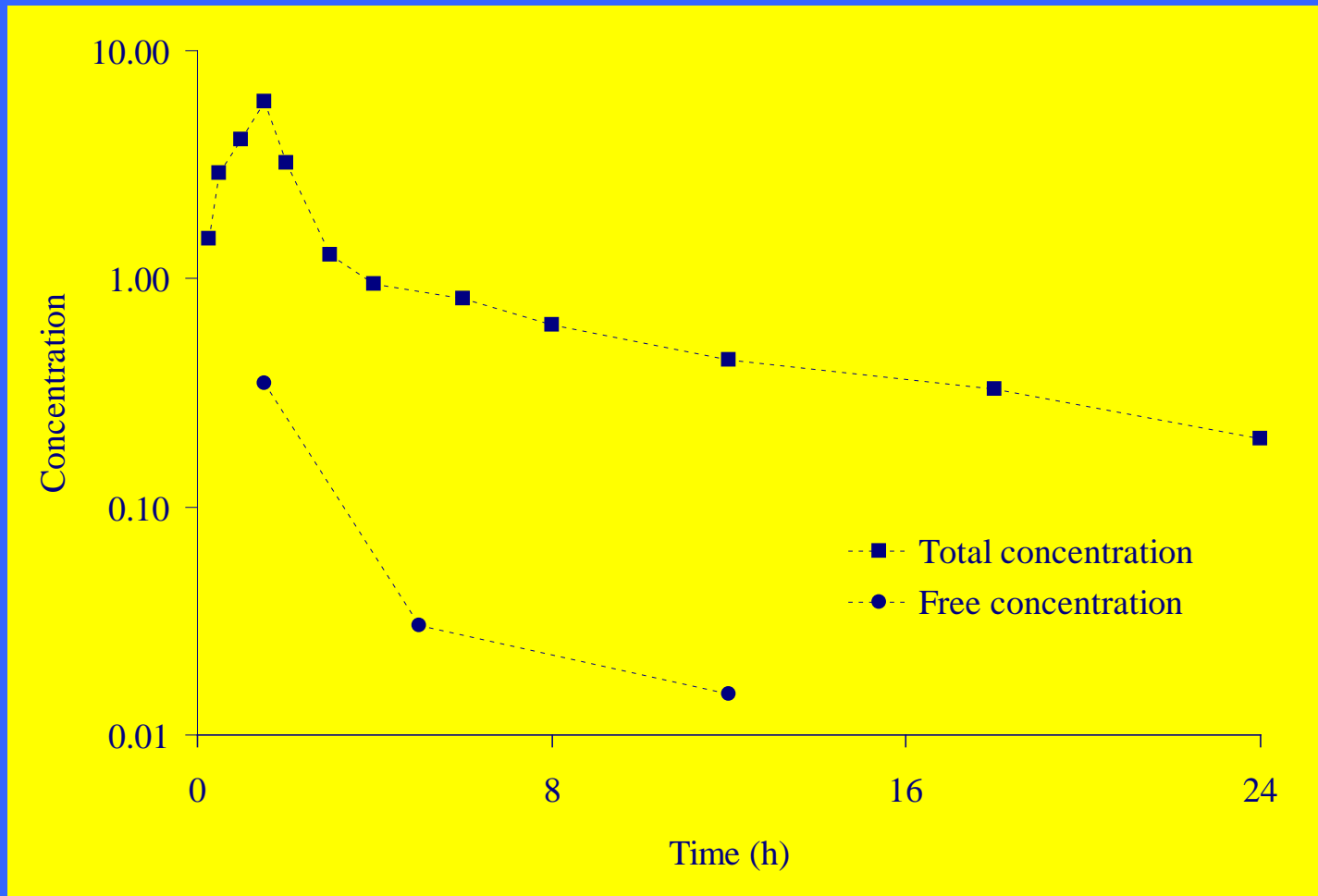
Unbalance and confounding



Non-linear PK



Rich + sparse data



Summary

- PK/PD is model driven
- PK/PD models aid the interpretation of pharmacological data and can be used prospectively to design subsequent studies **learning/confirming**
- Nonlinear mixed effects modelling allows data from a variety of unbalanced, sparse designs to be analysed
- Software for nonlinear mixed effects modelling is now widely available - **even for amateurs!**