

Introduction to Population Pharmacokinetic modeling

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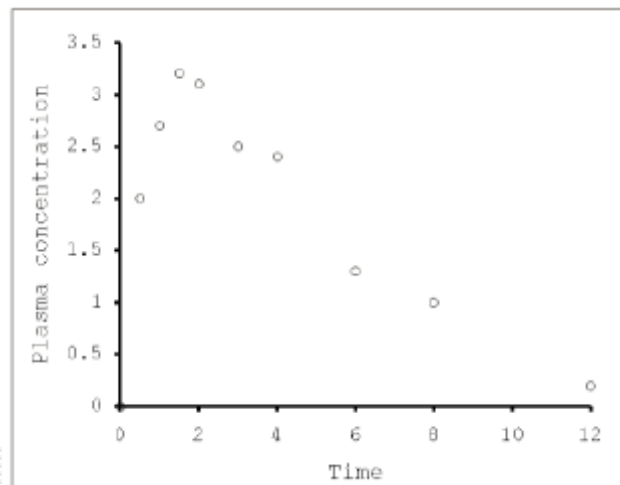
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No conflicts of interest to report

Questions:

- › Population PK is the description of the PK of an individual in a population. Yes (red) or No (green)
- › Non-compartment analysis is suitable to study the influence of physiological changes on the of behaviour of the drug. Yes (red) or No (green)
- › PK curves of healthy volunteers are aimed at model identification (red) or identification of interindividual variability (green)

Typical data in an individual PK curve



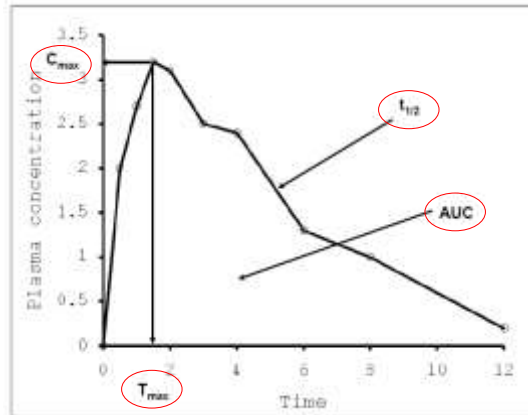
Non-compartmental pharmacokinetic analysis

Advantage:

- Simple

Disadvantage:

- Only descriptive
- Not suitable for study of the influence of physiological changes on the behaviour of the drug



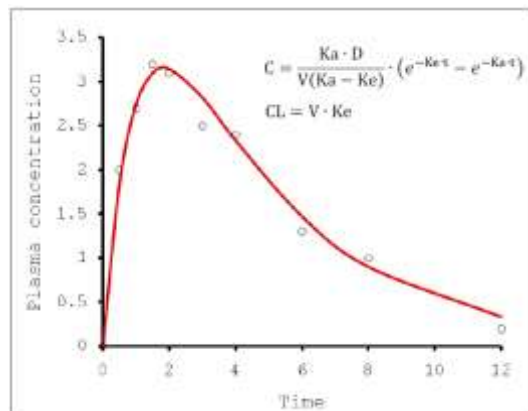
Compartmental pharmacokinetic analysis

Advantage:

- Can be used for mathematical modelling in other doses and patients

Disadvantage:

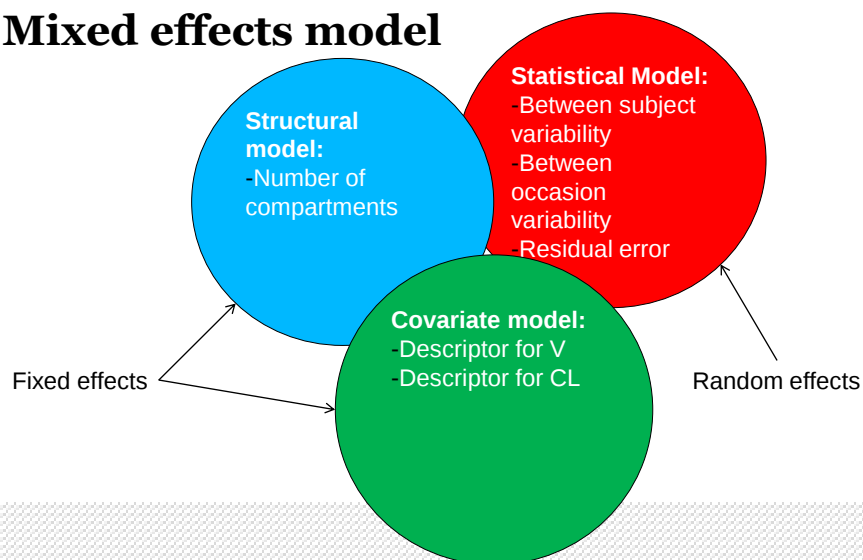
- Requires mathematical and physiological understanding
- Requires software
- Takes more time



Population Pharmacokinetics is

- › A description of pharmacokinetic behaviour of a drug in a population
 - › Using a pharmacokinetic model, describing the typical relationships between physiology and pharmacokinetics
- › A description of the interindividual variability in these relationships
 - › Using a statistical model for parameter distribution and error

Mixed effects model



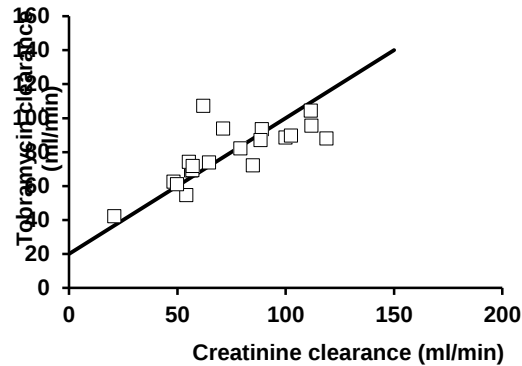
Population modeling allows you to:

- › Find interindividual characteristics on PK
 - › Covariates like renal function, pharmacogenetics
- › Find interoccasion or intraindividual variability
 - › Bioavailability, altering renal function
- › Find other sources of error
 - › Model misspecification
- › Define MAP Bayesian parameter estimation
 - › for dose adjustment in individual patient (TDM)

Model parameters

- › Model parameters
 - Measure of central tendency ('mean value')
 - Measure of inter-individual variability ('sd')
 - Covariance between parameters (often ignored!)
 - Assessment of covariates

$$CL = CL_m + fr \cdot CL_{cr}$$



Parameter distribution

- › Parametric methods require assumptions, e.g.
 - normal distribution
 - log-normal distribution
- › Nonparametric methods

Data (measurements)

- › Rich data
- › Sparse data

Rich data

- › Large number of blood samples from each subject
- › Small number of subjects
- › Experimental environment
- › Healthy volunteers

- › Aim: model identification

Sparse data

- › Small number of blood samples from each patient
 - › Large number of subjects
 - › Clinical environment
 - › Patients
-
- › Aim: Identification of model parameters and covariates for TDM

Methods

- › Naive pooling
- › Standard Two-Stage (STS)
- › Mixed-Effect modeling (eg: NONMEM)
- › Nonparametric methods (eg: NPEM)
- › Iterative Two-Stage Bayesian (ITSB)

Naive Pooling

- › Data of all patients pooled
- › Inter-individual variability ignored
- › No information on inter-individual variability obtained

Standard Two-Stage (STS)

Step 1

Data of each patient analysed separately

Step 2

Mean and SD of model parameters calculated from results of step 1

Standard Two-Stage (STS)

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- Conceptually and computationally simple

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- Inter-individual variability overestimated
- Not applicable to sparse data
- Problems with ‘non-fittable’ patients

Mixed-Effect Modeling

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- Statistically sophisticated
- Generally accepted (FDA)
- Inter- and intra individual variability estimated
- Rich and sparse data

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- ‘Black box’

Iterative Two-Stage Bayesian

- › Prior knowledge of parameter distribution is needed to estimate posterior distribution
- › Assume a reasonable set of population data (e.g. from STS)
 - means $\pm$ sd
 - covariance matrix (usually zero)
 - residual error (e.g. assay error pattern)

Iterative Two-Stage Bayesian

- › Step 1: Perform Bayesian analysis on each subject separately
- › Step 2: Calculate new set of population data
 - means $\pm$ sd
 - covariance matrix
 - residual error

Iterative Two-Stage Bayesian

- › Repeat step 1 and step 2 until convergence is reached, i.e.
stable values for:
 - means $\pm$ sd
 - covariance matrix
 - residual error

Iterative Two-Stage Bayesian

- +
 - Conceptually and computationally simple
- - Results may be less precise and/or less accurate for sparse data

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Literature

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