



Simulating individual variability in pharmacokinetics as a risk factor for drug toxicity

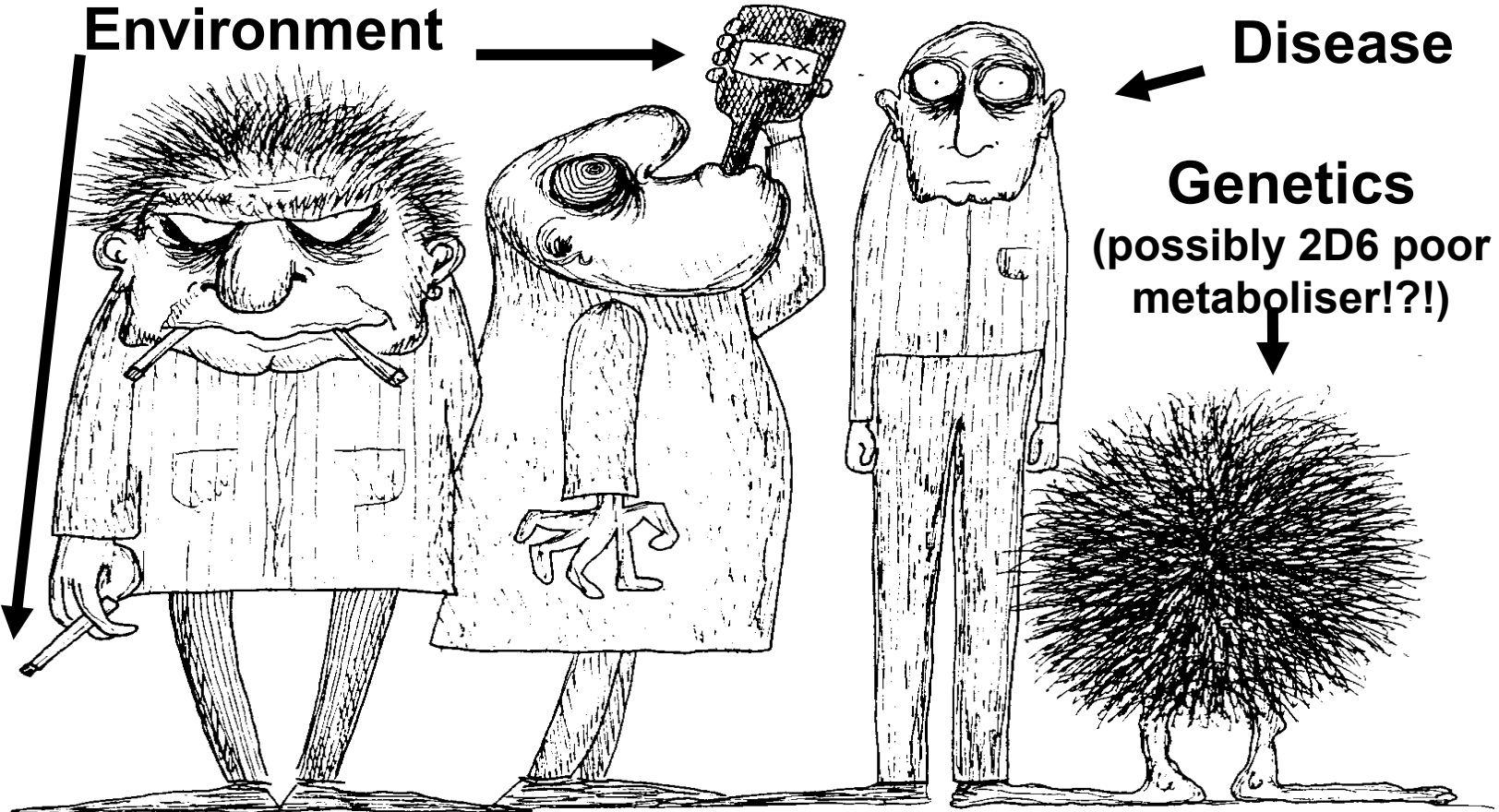
Iain Gardner
Head of Translational DMPK Science
Simcyp Limited

New Perspectives in DMPK: Informing Drug Discovery,
Royal Society of Chemistry, London
11 February 2014

Outline of the presentation

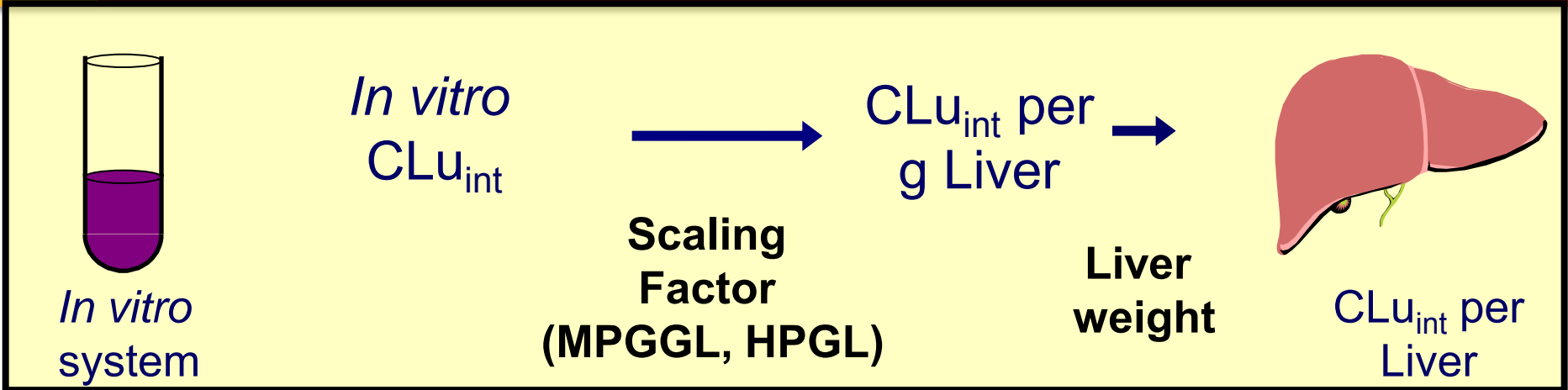
- Integrating population variability into PK simulations
 - creation of virtual populations
- Sources of inter-individual variability in Pharmacokinetics
 - Drug Clearance
- Using virtual populations to predict risk factors for drug toxicity
 - Cardiac Safety

The Challenge of Population Variability



Cross Section of Patients in the Royal Hallamshire Hospital

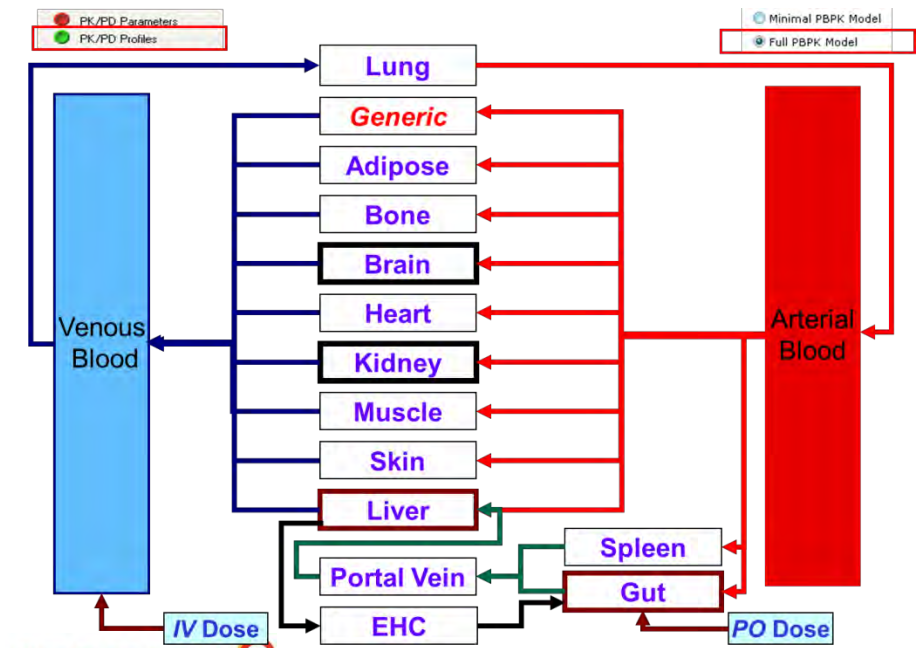
Prediction of human PK (PD) in virtual individuals



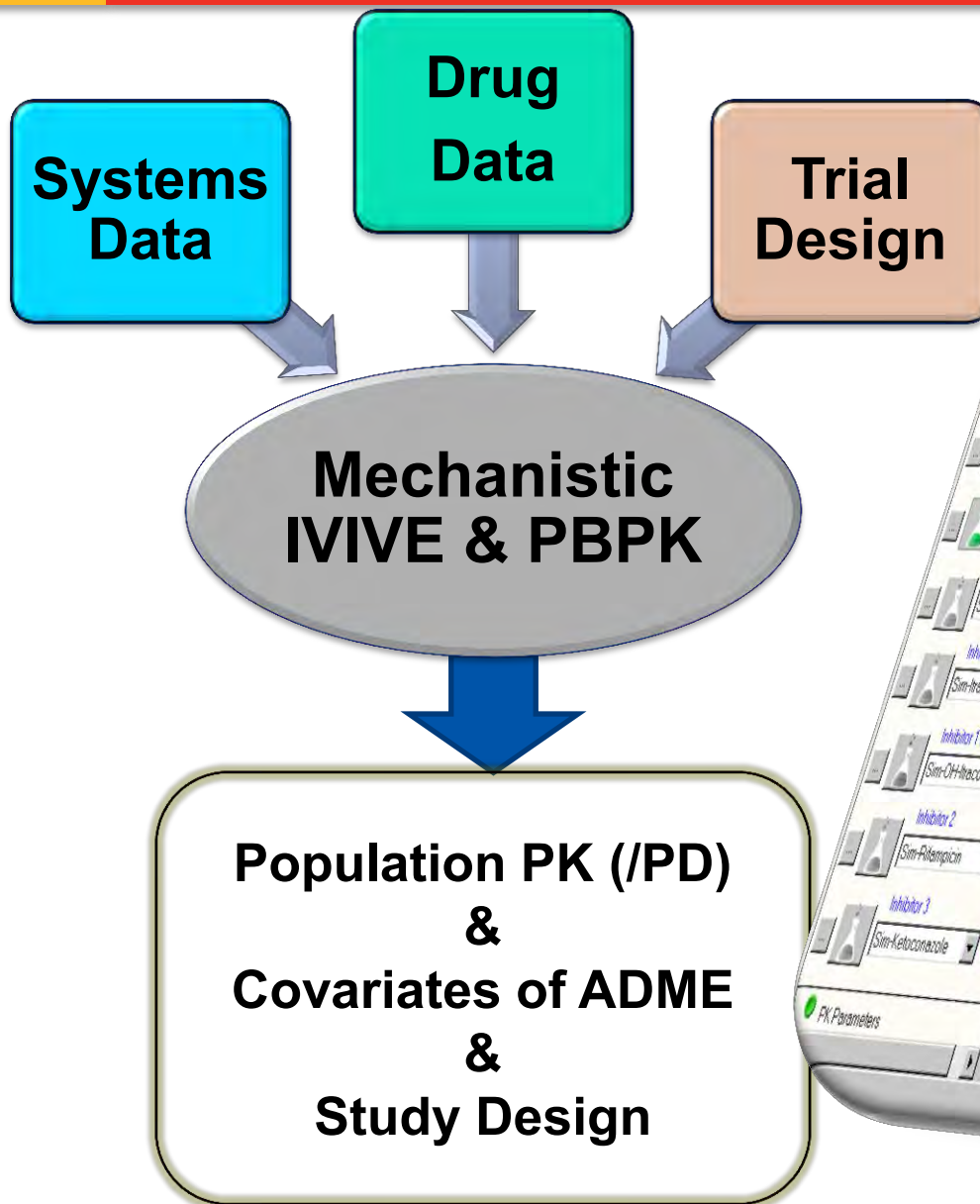
Simcyp approach

Combine in vitro-in vivo extrapolation (IVIVE) and PBPK approaches in virtual individuals to predict drug concentration and effect

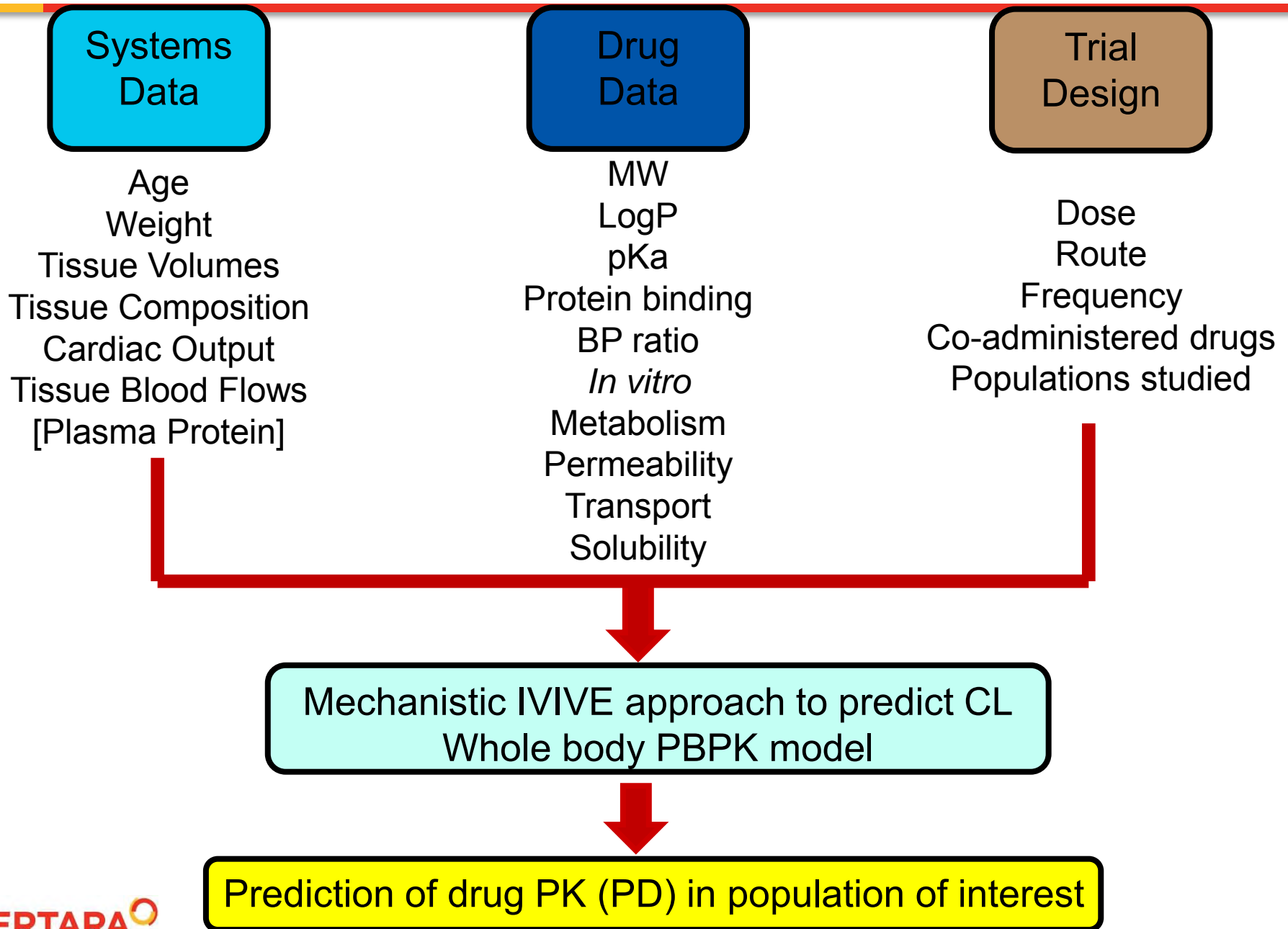
Identifying relevant DISTRIBUTION of values for demographical, biological, physiological and genetic parameters in target population & the COVARIATIONS between the parameters in target POPULATION



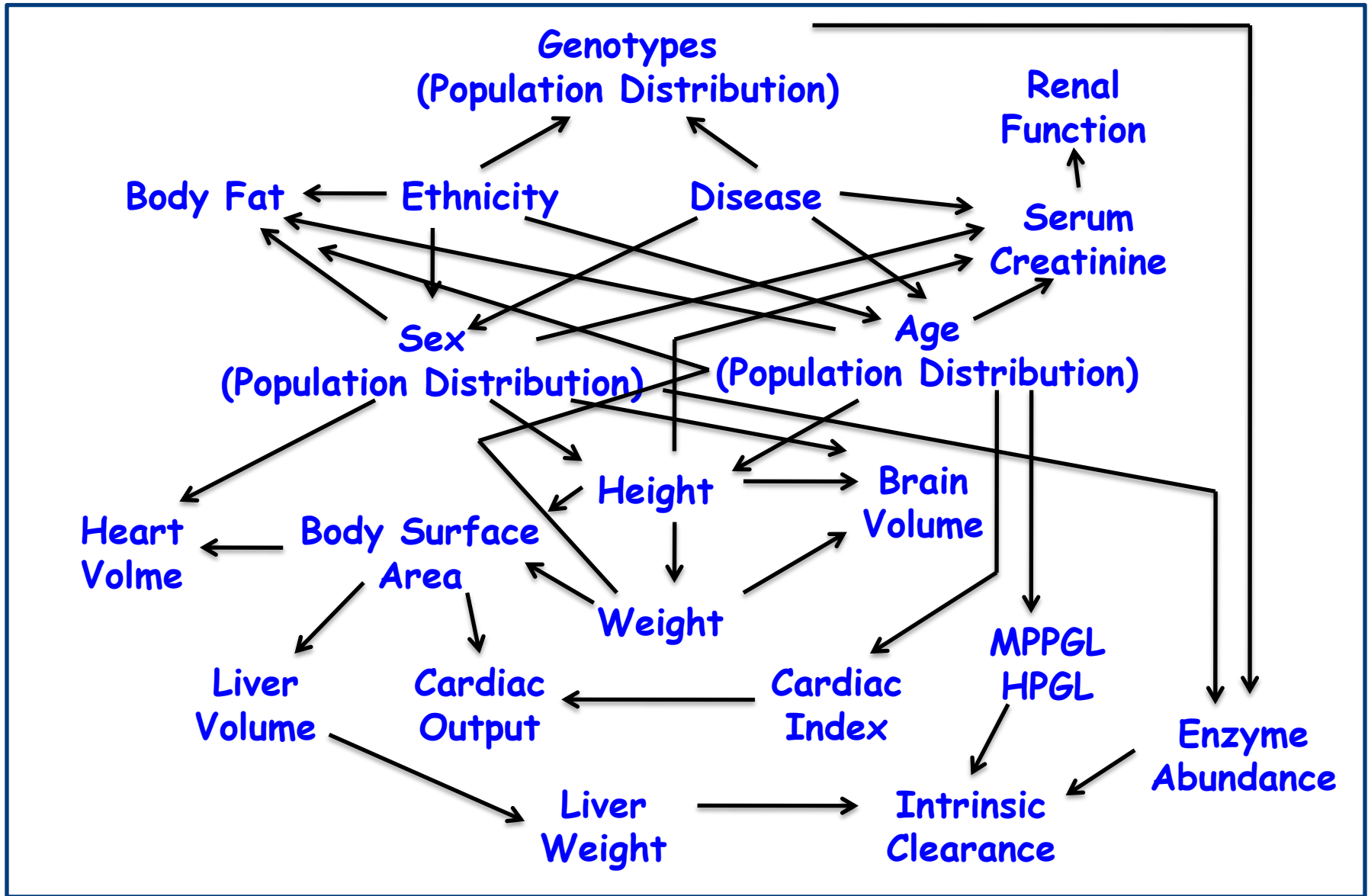
Separating Systems & Drug Information



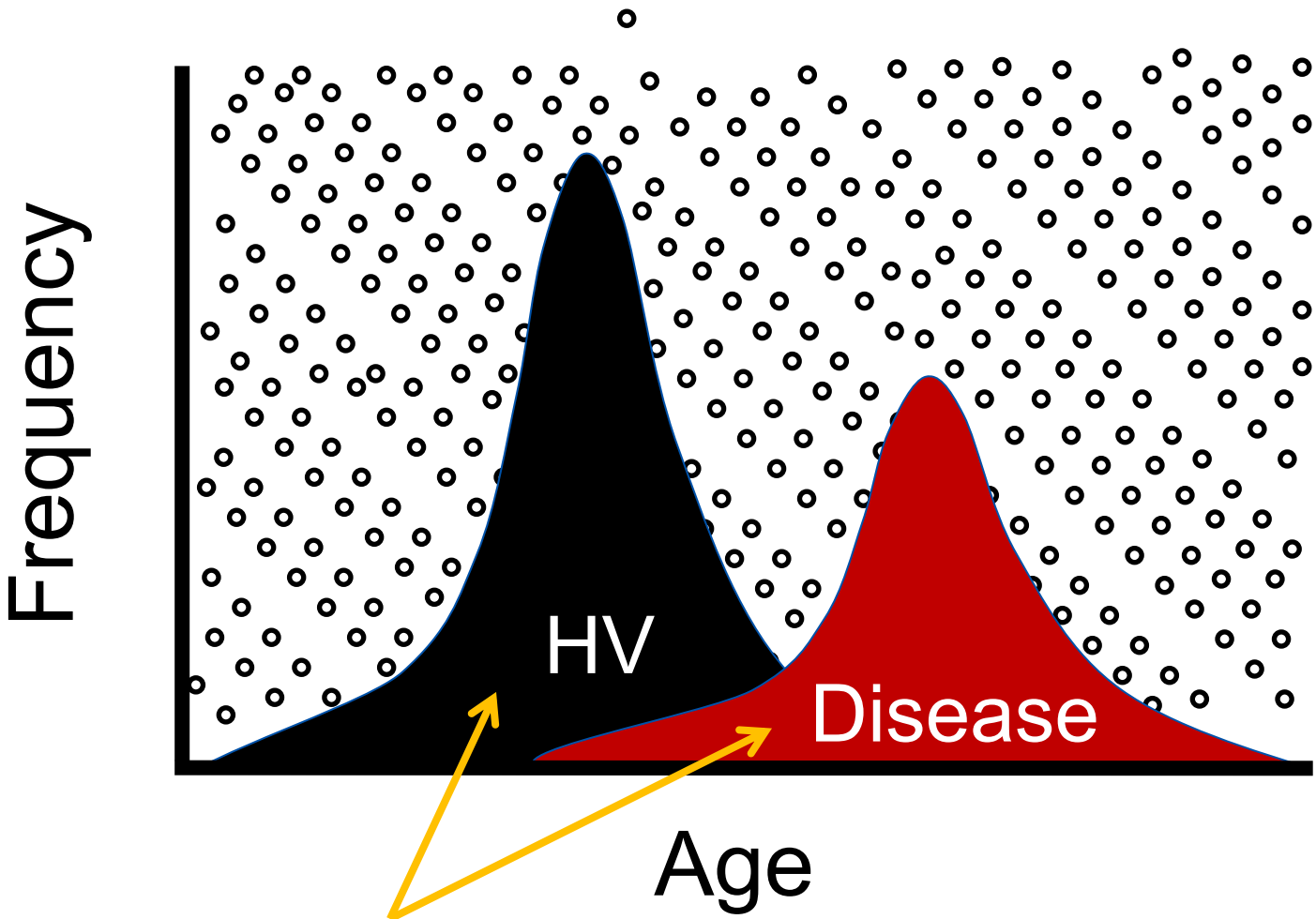
Separating Systems & Drug Information



Building virtual populations

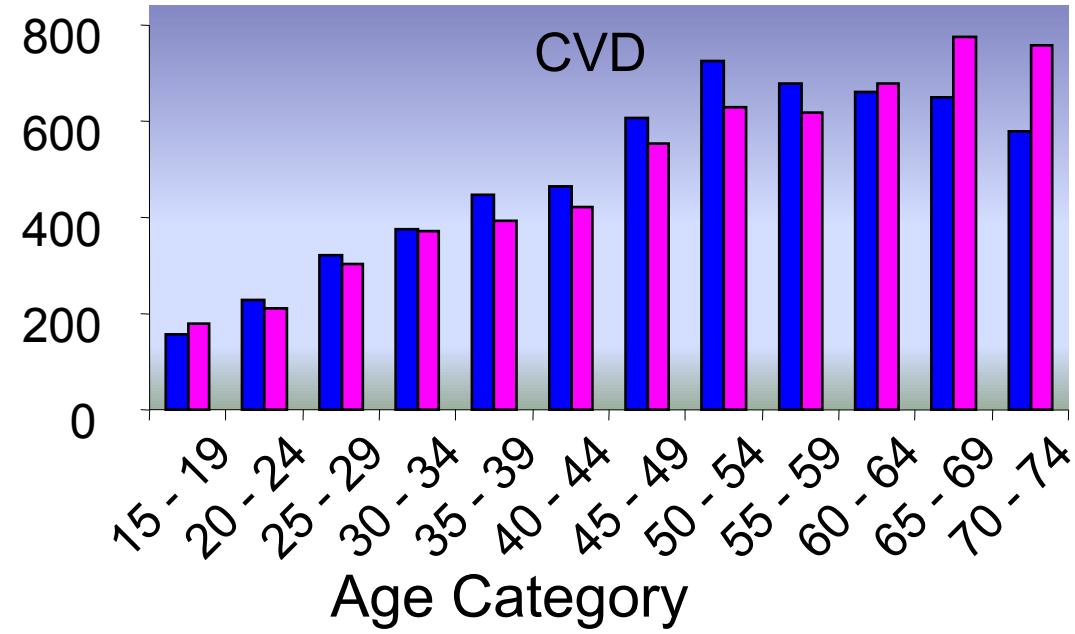
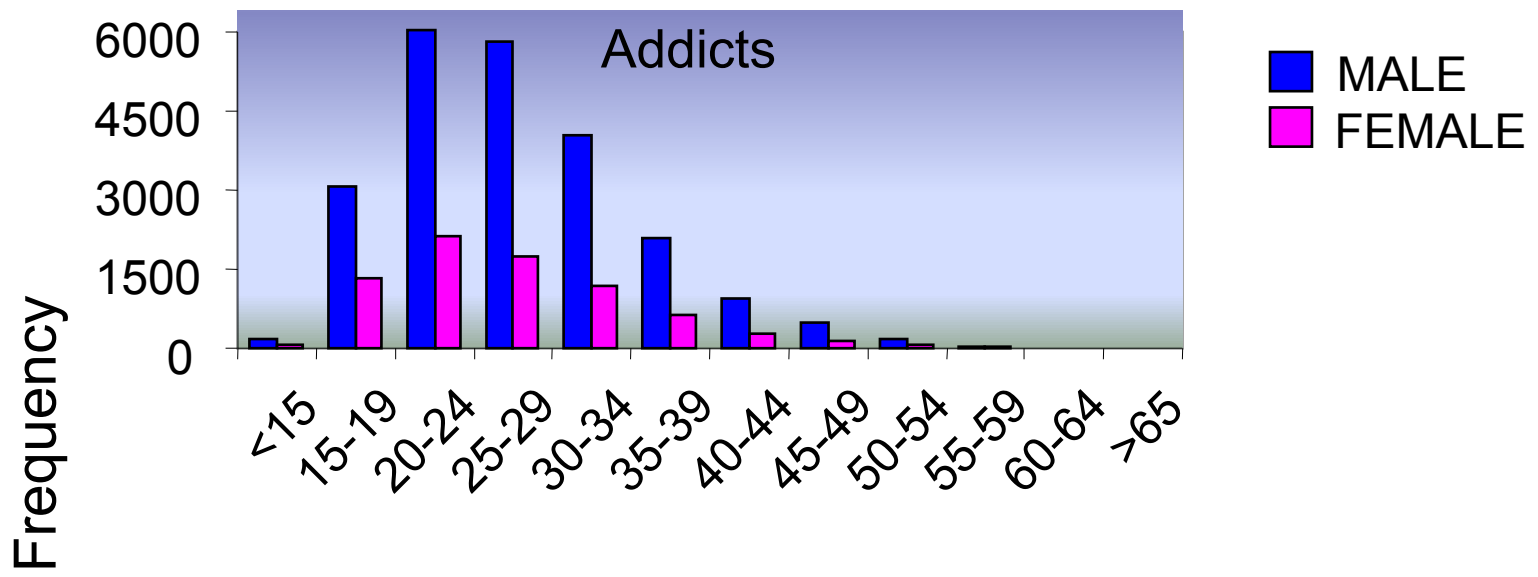


Demographic Features of Healthy and Disease Populations



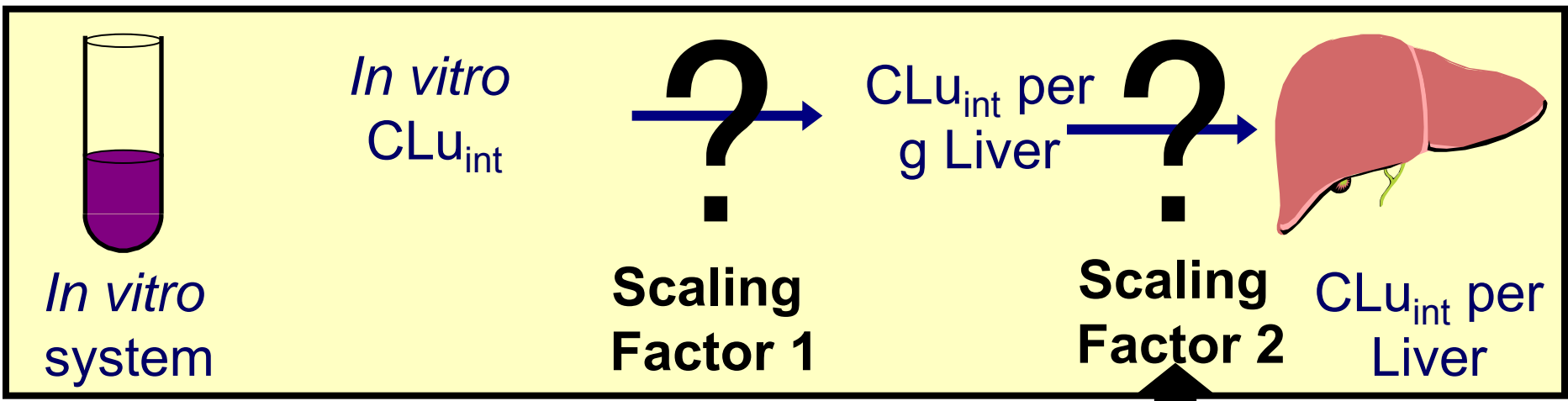
Defined by real data

Age Distribution in Target Population



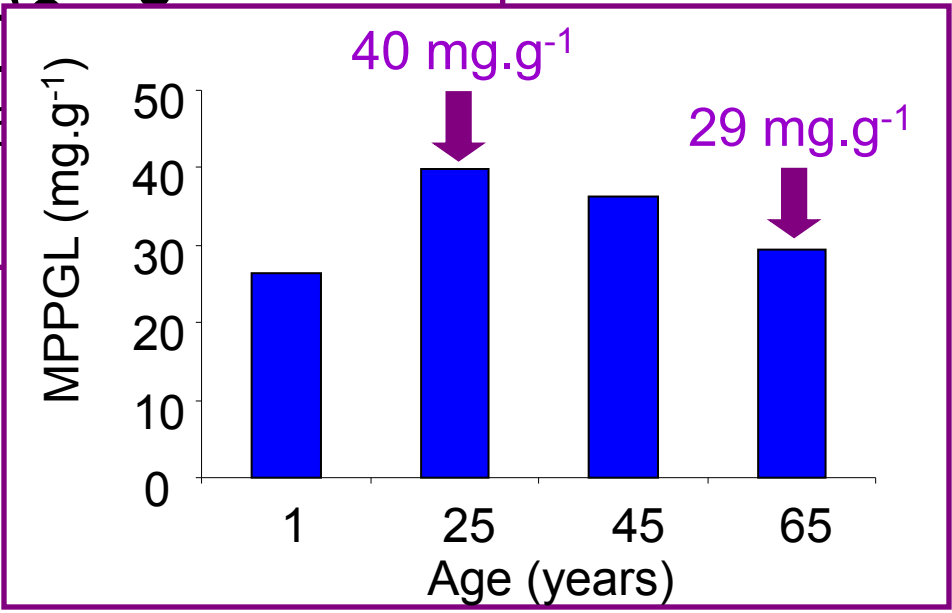
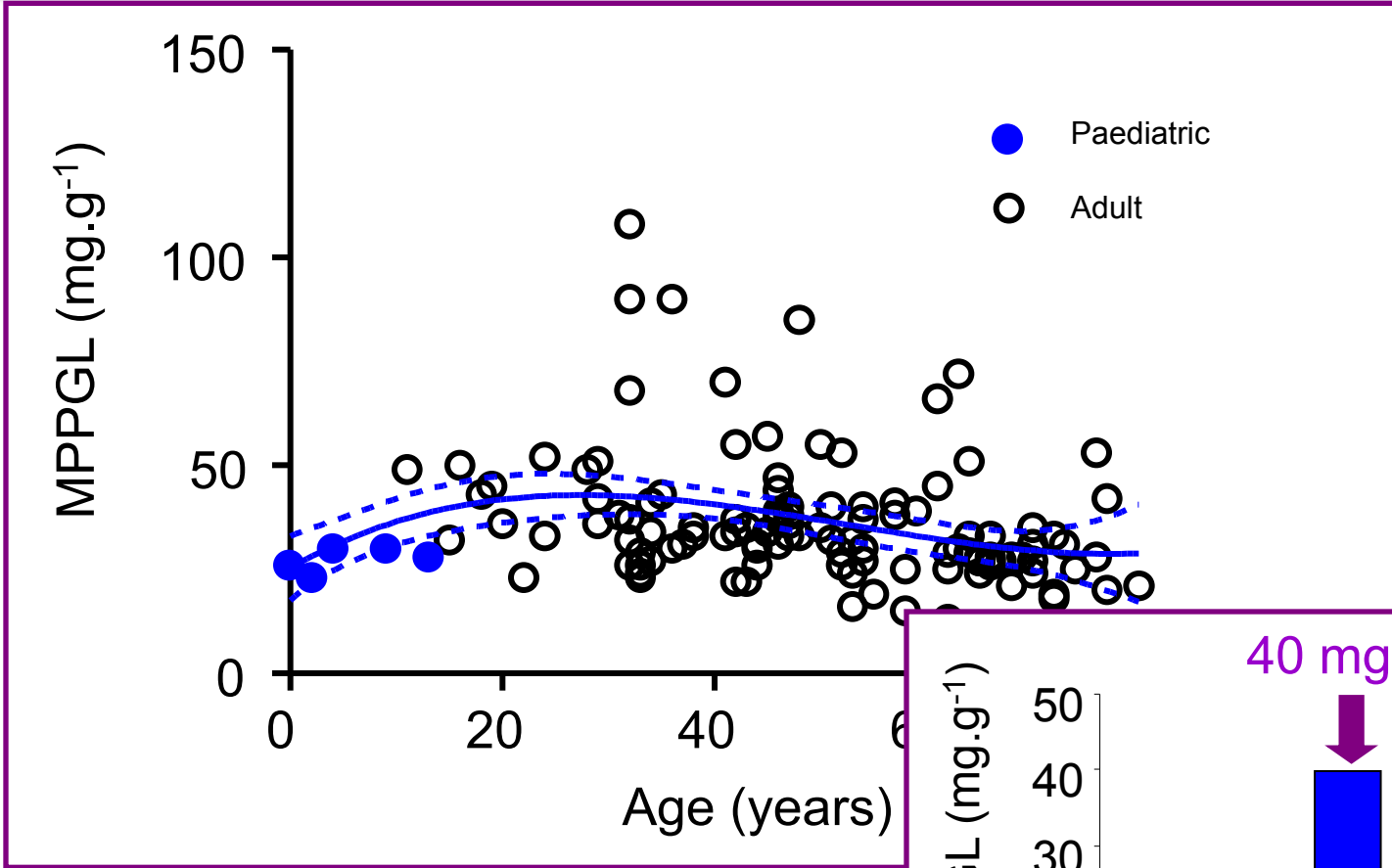
CLEARANCE: *In Vitro – In Vivo* **Extrapolation**

Scaling Factors in Human IVIVE



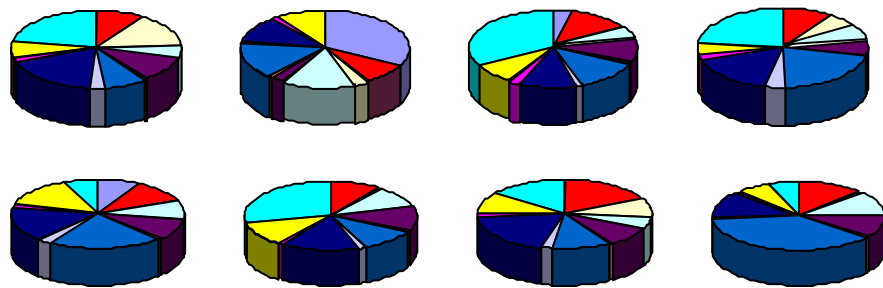
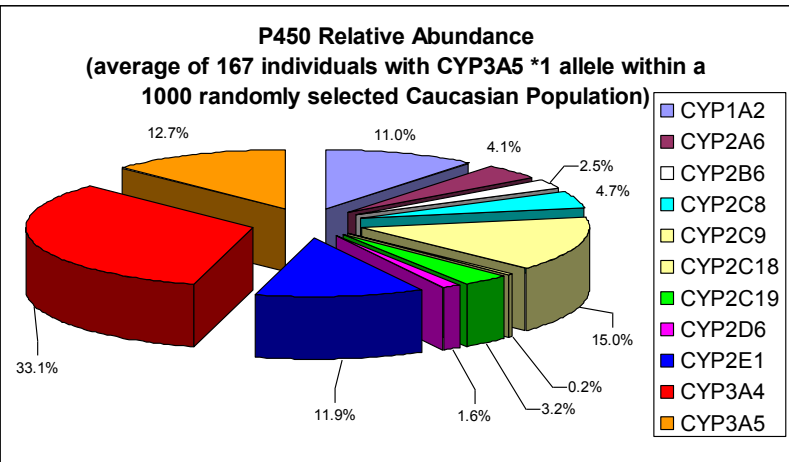
HHEP	$\frac{\mu L \cdot min^{-1}}{10^6 \text{ cells}}$	X	HPGL] X [Liver Weight
HLM	$\frac{\mu L \cdot min^{-1}}{mg \text{ protein}}$	X	MPPGL	
rhCYP	$\frac{\mu L \cdot min^{-1}}{pmol \text{ CYP}}$	X CYP abundance	X MPPGL	

Sources of Variability: MPPGL and Donor Age



Barter *et al.* 2007 *Curr Drug Met*
Barter *et al.* 2008 *Drug Met Disposition*

CYP Abundance Variability

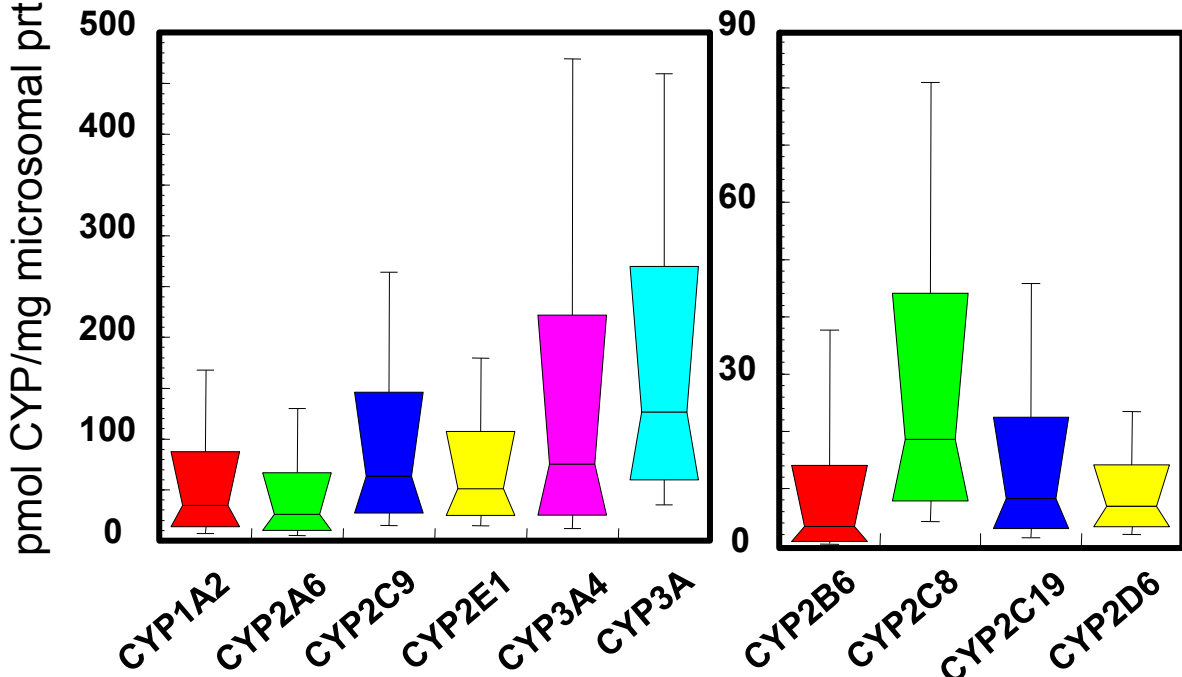


Different Individuals

- Genetics (Ethnicity)

PLUS

- Age (Ontogeny)
- Environment (Ethnicity)
- Sex / Co-medications

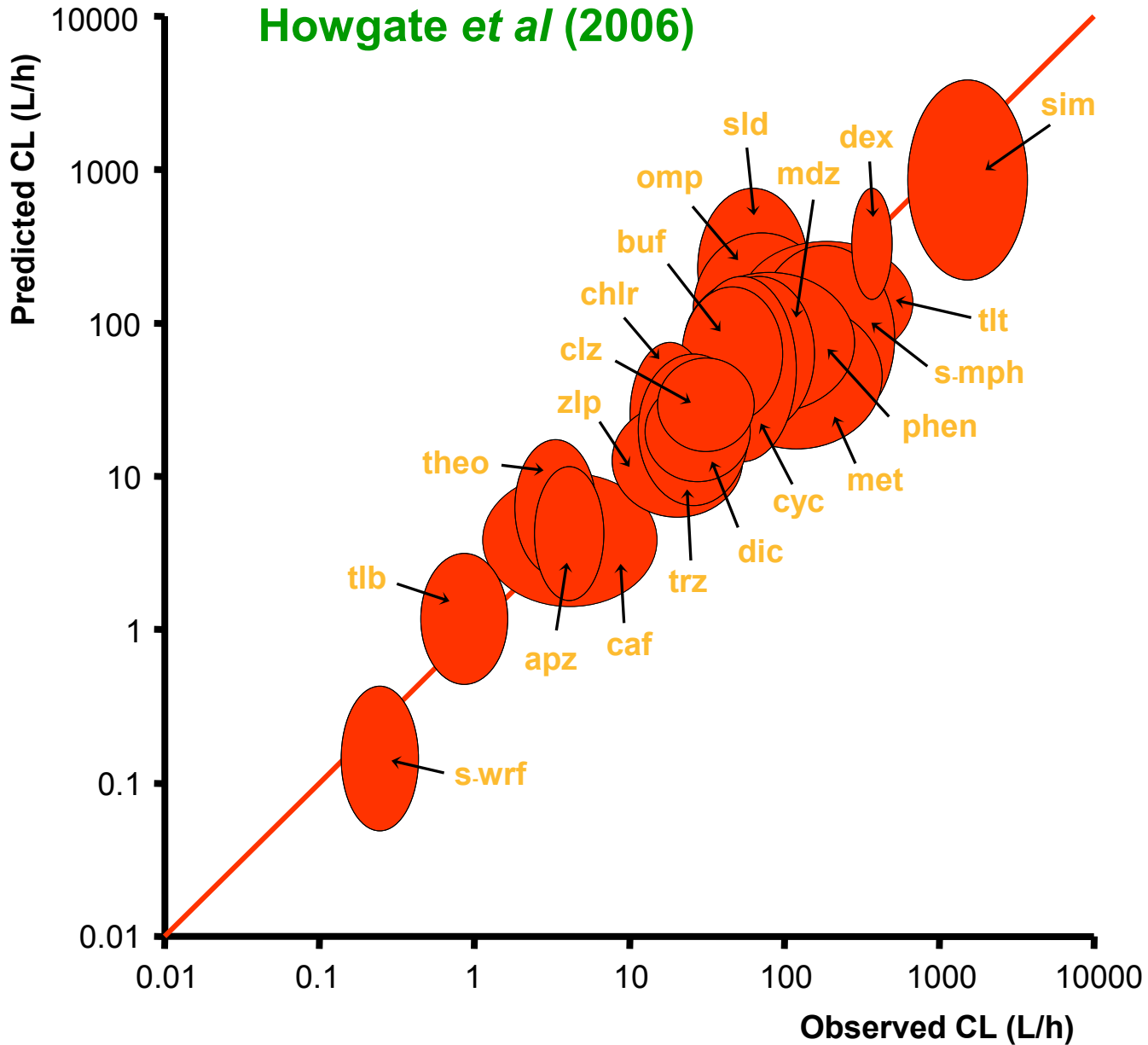


PREDICTION OF CLEARANCE (Oral)

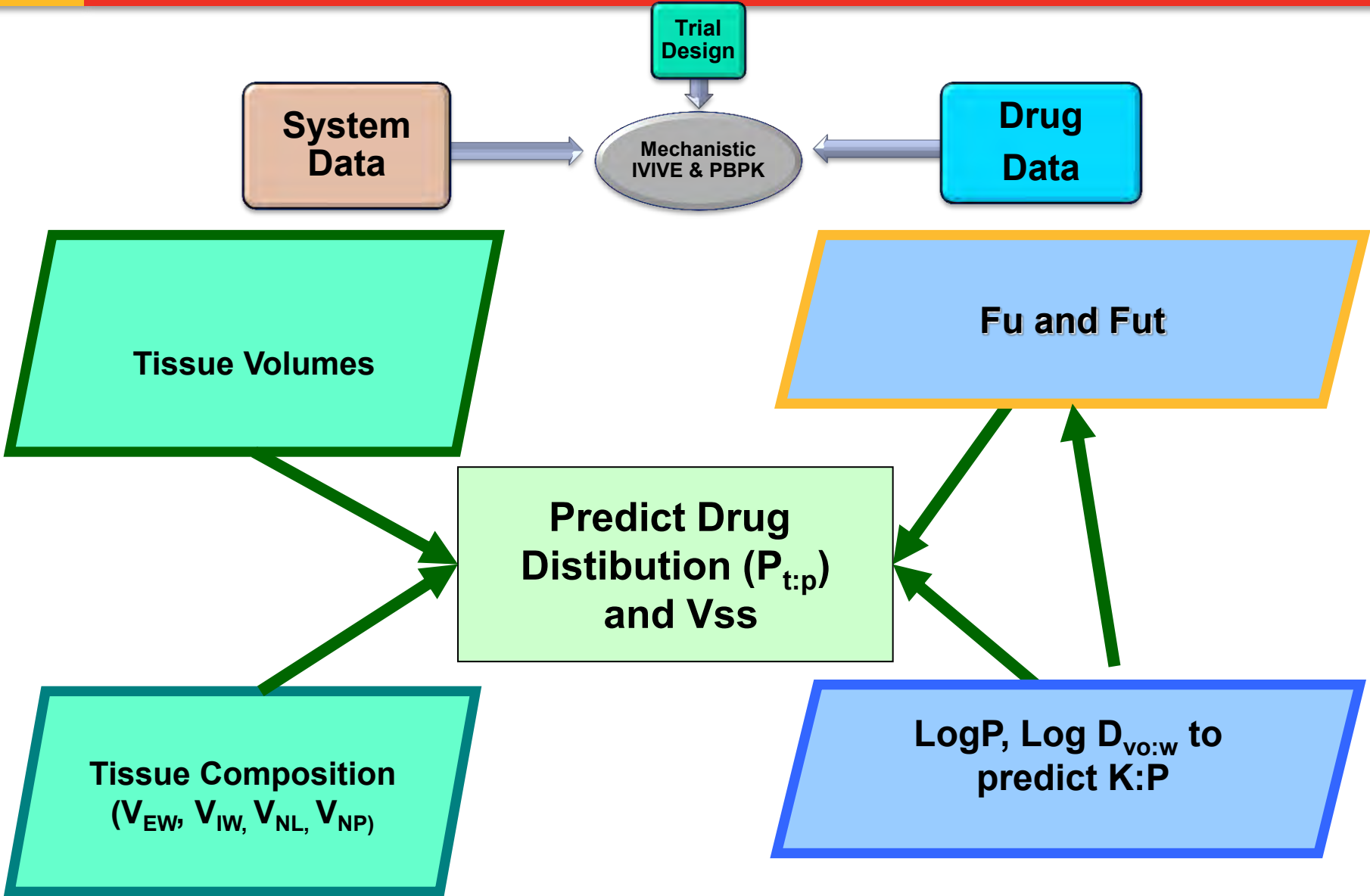
Central
tendency

&

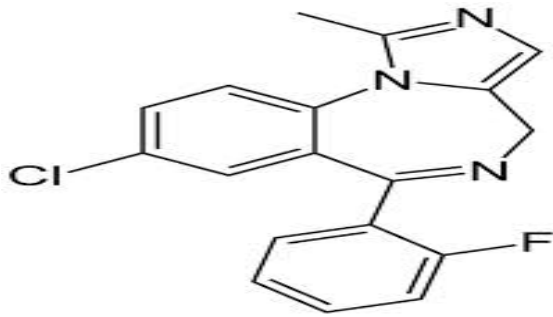
Measure of
variability



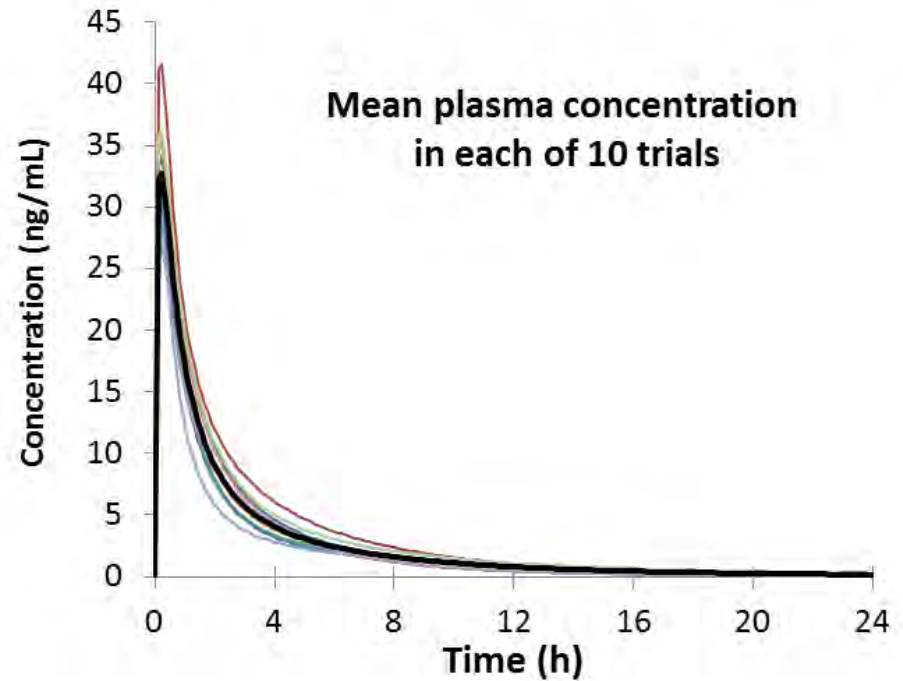
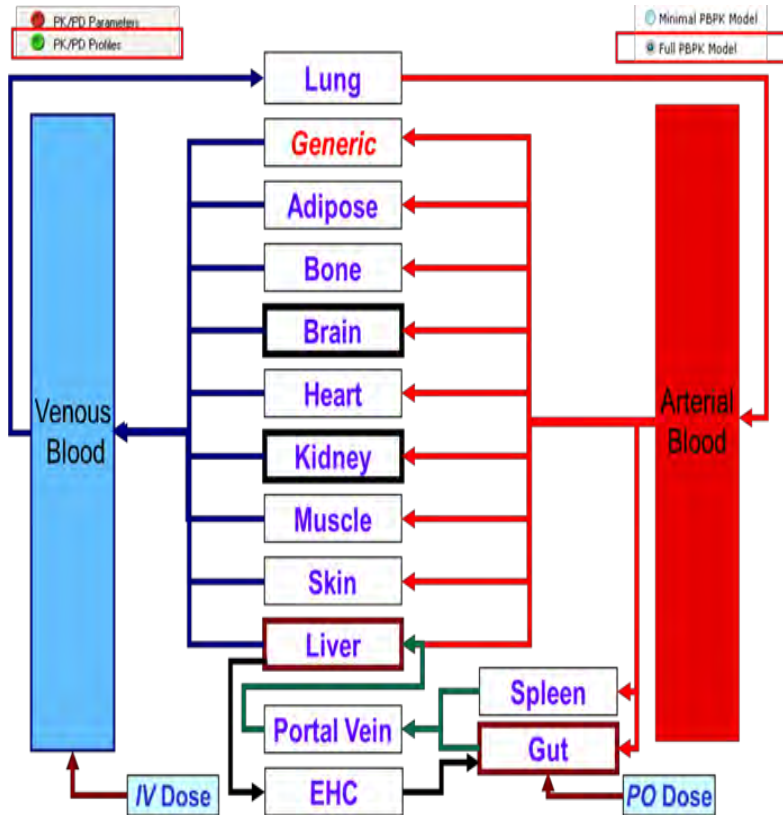
Drug Distribution



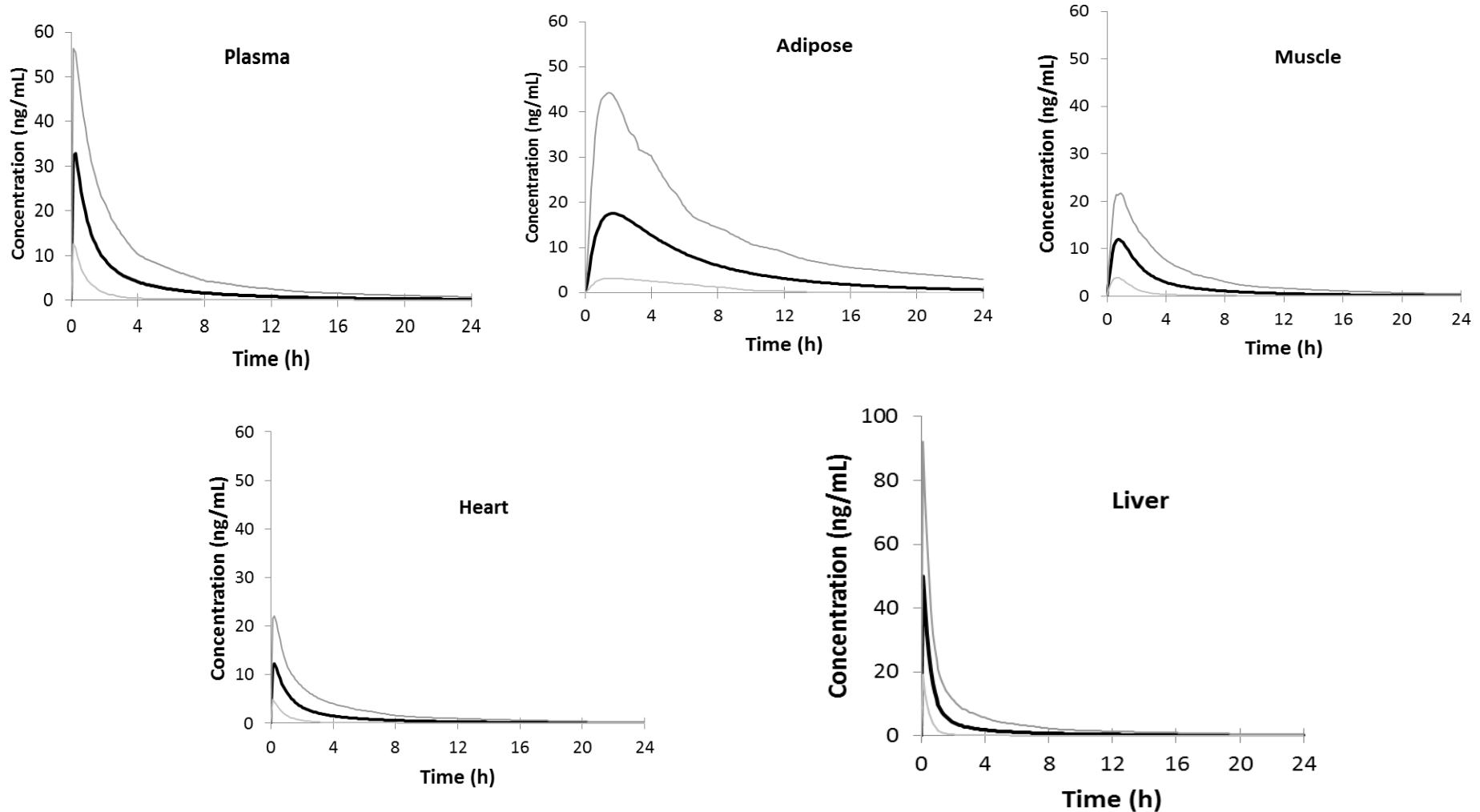
Example: Midazolam pharmacokinetics (simulations in 10 trials of 10 individuals)



Can describe Midazolam Pharmacokinetics using *in vitro* metabolism data together with systems physiology data



Midazolam Tissue concentrations (mean and 5 and 95 percentiles)



Concentrations in the tissues can also be linked to Pharmacodynamic or Toxicological effects

Cardiotoxicity

- Cardiac side effects major cause of drug withdrawal (regardless of the development level – from pre-clinical up to the post-approval)
 - E.g. Torsade de points and terfenadine
- Various mechanisms and effects involved
 - pro-arrhythmia, cardiac cell toxicity
- Drug interactions – important element causing serious adverse events
 - E.g. astemizole (CYP related)
- PK variability important for safety assessment
 - E.g. tolterodine (CYP 2D6 mediated metabolism – genetic variability)

Screening for cardiac safety

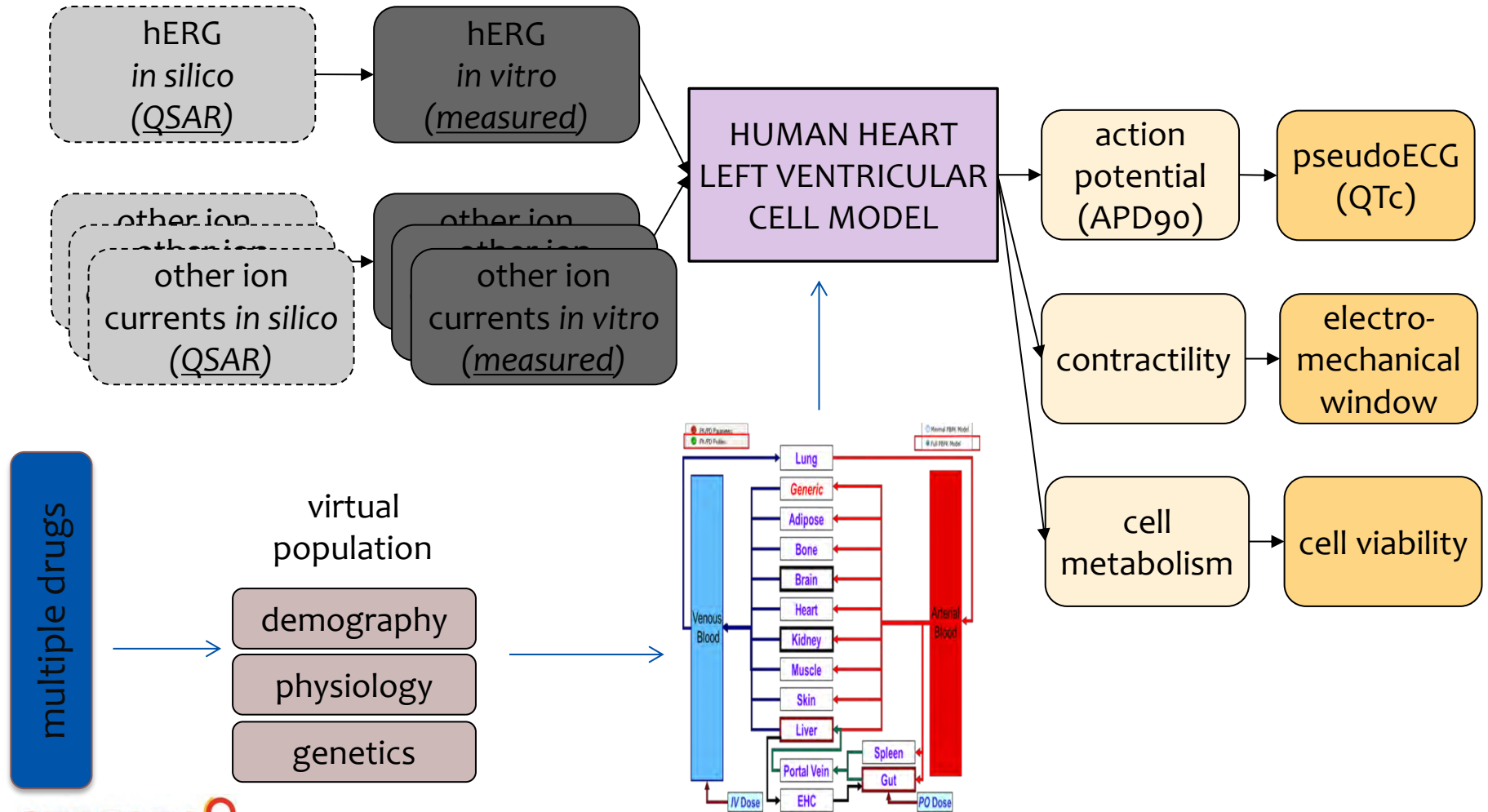
- Effects of new compounds on IKR/Herg extensively screened for in drug discovery/development
 - QSAR models
 - IKR binding
 - HERG inhibition
 - Purkinje fiber studies
- Cardiac safety also often investigated in vivo
 - Pre-clinical studies
 - Thorough QT study in humans (~\$1000000)
- Can cardiotoxicity also be assessed using mechanistic *in silico* models?

Links to PD: Assessment of Proarrhythmic Potency

Virtual population generator for human cardiomyocytes
parameters: *in silico* drug cardiotoxicity assessment

Toxicology Mechanisms and Methods, 2012

Sebastian Polak¹, Kamil Fijorek², Anna Glinka¹, Barbara Wisniowska¹, and Aleksander Mendyk³

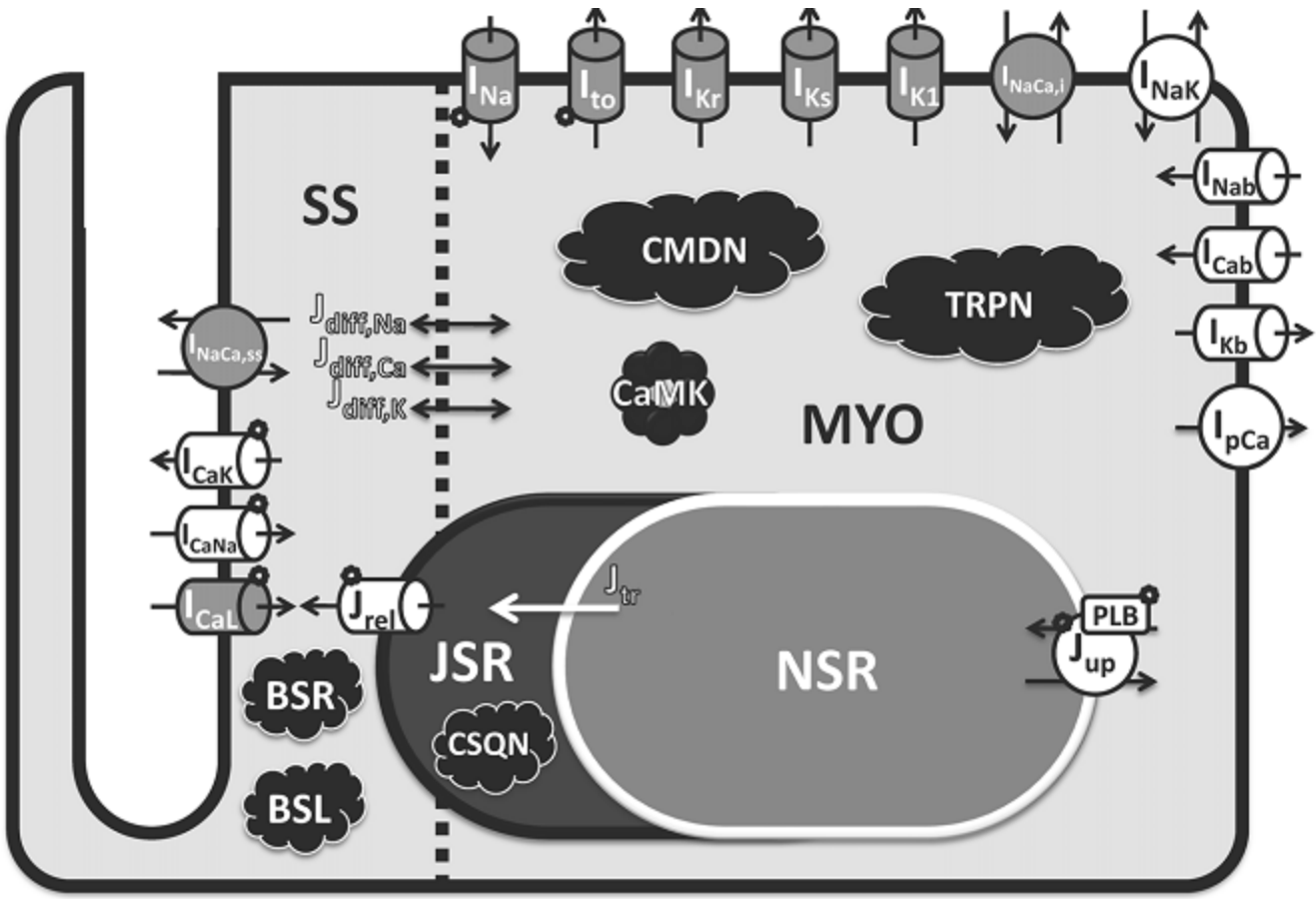


Structure of left ventricular cell model

- molecular structure —> QTc prolongation/TdP
 - mechanistic/physiological

$$\frac{dV}{dt} = \frac{I_{ion} + I_{stim}}{C_m}$$

HUMAN HEART
LEFT VENTRICULAR
CELL MODEL



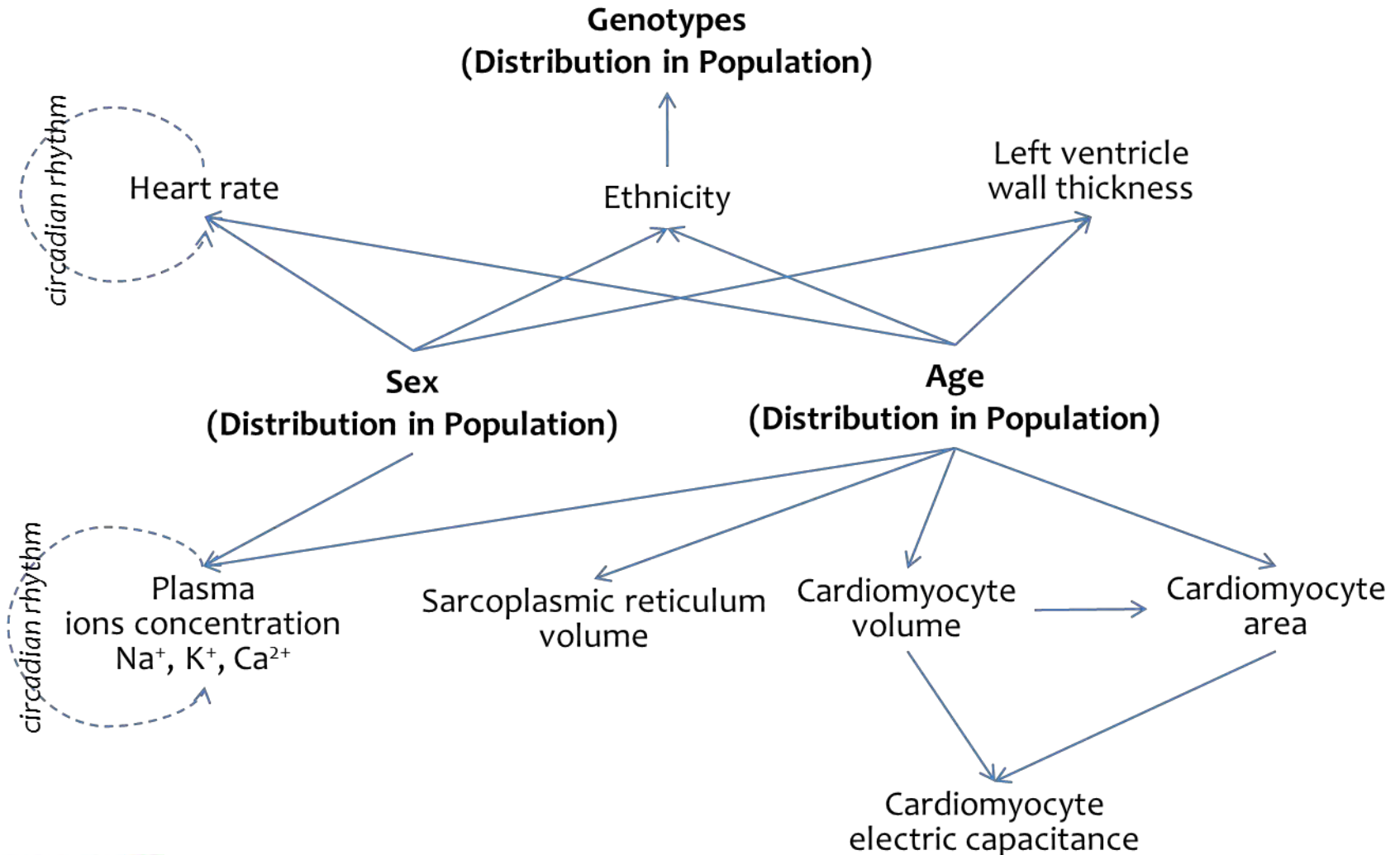
O'Hara and Rudy *PLoS computational Biology* 7, 2011
Ten Tusscher et al. *AmJPhys-HeartPhys* 286, 2004

Variability matters

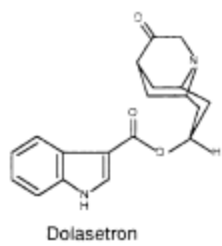
- Clinical evidence

Group	Parameter	Influence on ECG	Reference
Demography	age	↑ age - ↑ QT/QTc	Pham 2002
	gender	Females have longer QT/QTc as compared to males.	James 2007
Anatomy/ physiology	plasma ions concentration (K ⁺ , Ca ²⁺)	↑ K ⁺ - ↓ QT/QTc ↑ Ca ²⁺ - ↓ QT/QTc	Etheridge 2003 Covis 2002
	cardiomyocyte size (volume, area)	↑ size - ↑ QT/QTc	Pacifico 2003
	heart wall thickness	↑ thickness - ↑ QT/QTc	Jouven 2002
	cells heterogeneity across heart wall	M cells presence influence the T wave and ECG in general.	Antzelevitch 2010
	heart rate	↑ RR - ↑ QT	Harris 2003 Malik 2002
	sex hormones	↑ testosterone - ↓ QT/QTc ↑ progesterone - ↓ QT/QTc	Pham 2002 Sedlak 2012
Genetics	common polymorphisms (ion channels)	Usually slight modification of ECG, may act as a genetic modifier (with different mutation both protective effect and QTc prolongation were observed).	Crotti 2005
	mutations (ion channels)	↑ QT/QTc	Etheridge 2003 McPate 2005

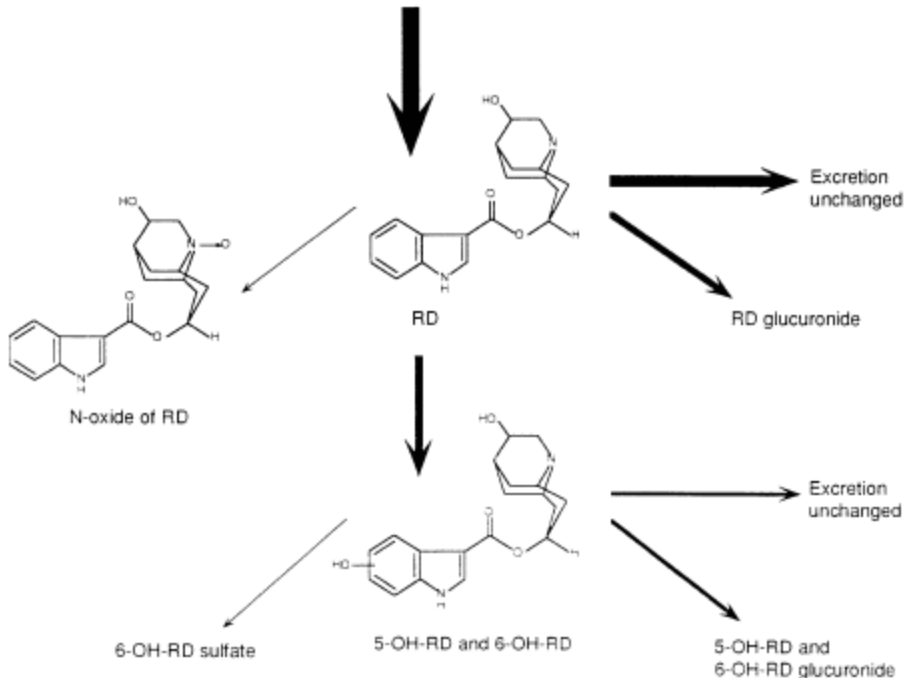
Building virtual populations – cardiac safety assessment



Example 1: Dolasetron – formulation dependent effect



DOLASETRON – CARDIOLOGICALLY ACTIVE
Plasma concentration negligible after po dosing



iv formulations – withdrawn from the market due to the TdP reports

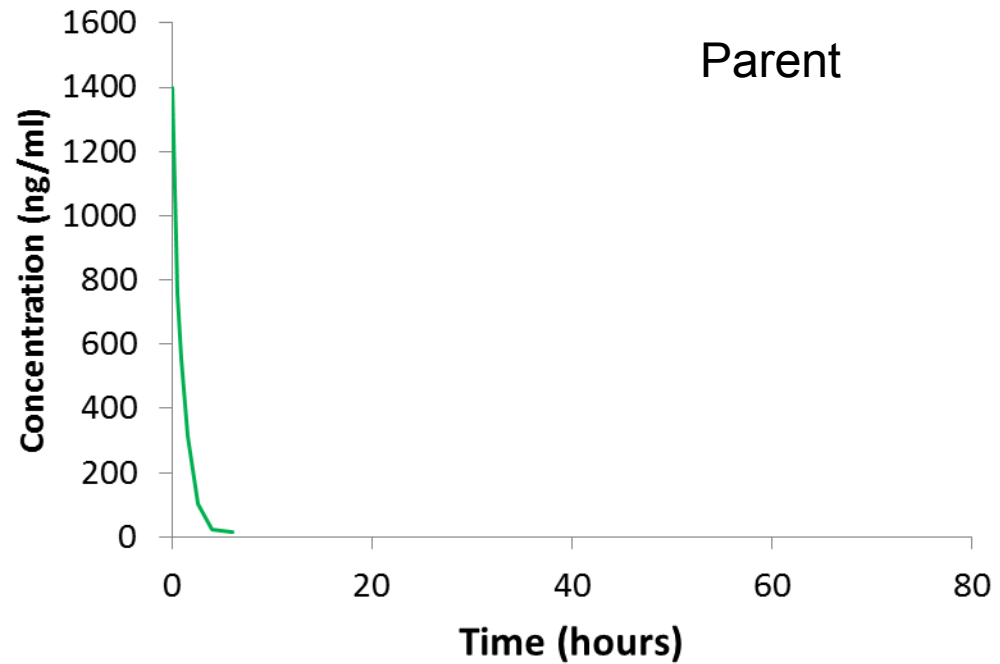
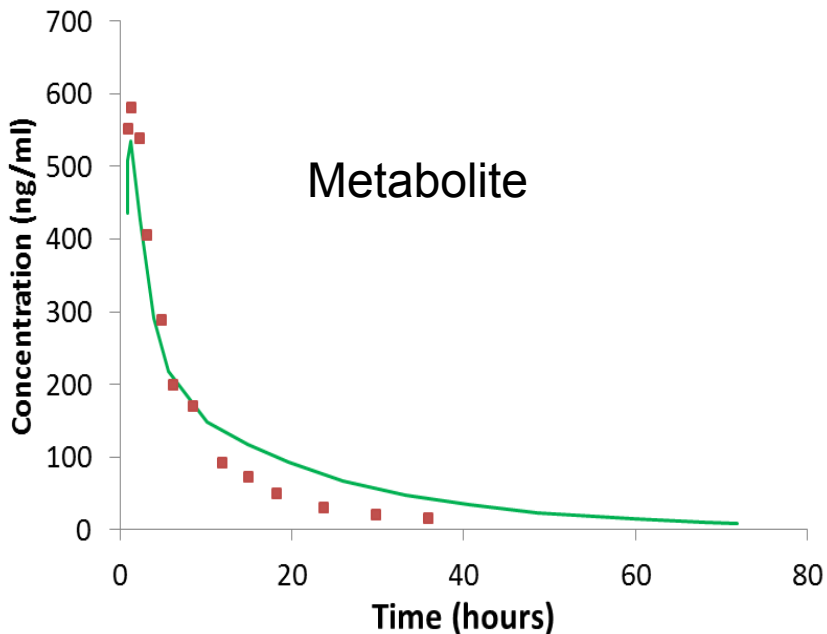
po formulation – considered as safe

Example 1: Dolasetron – Simcyp PK prediction

iv 2.4 mg/kg

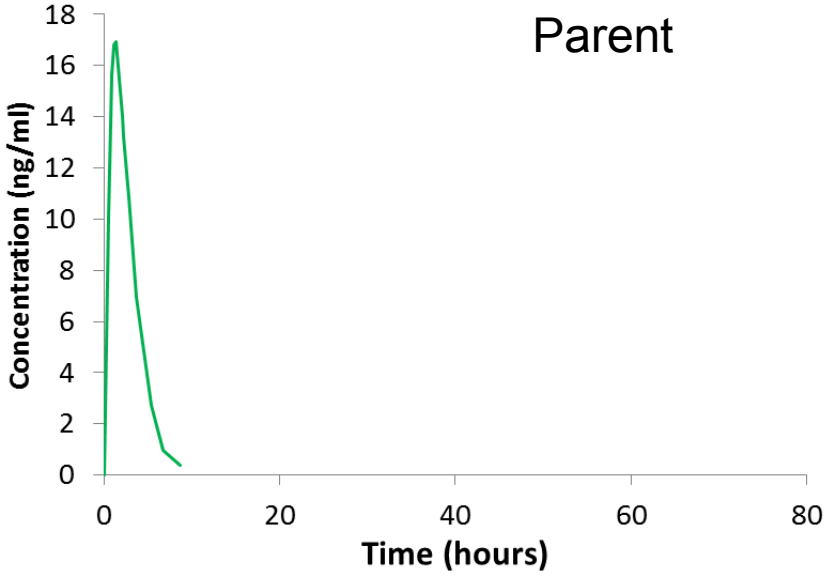
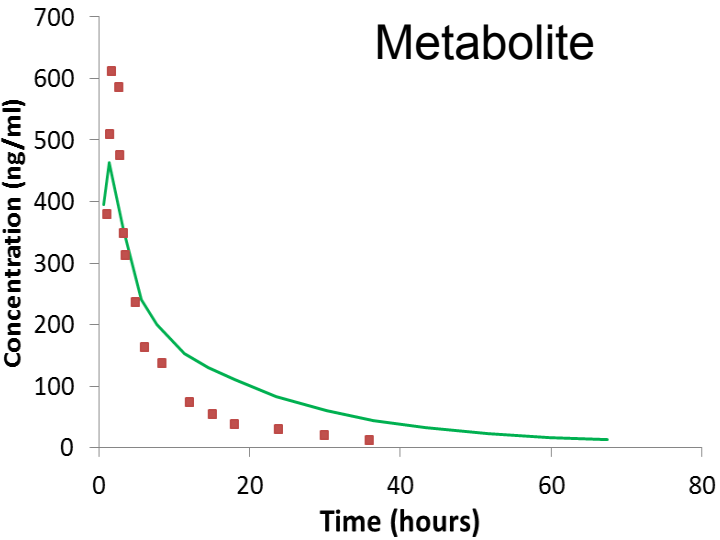
Dolasetron

Hydrodolasetron



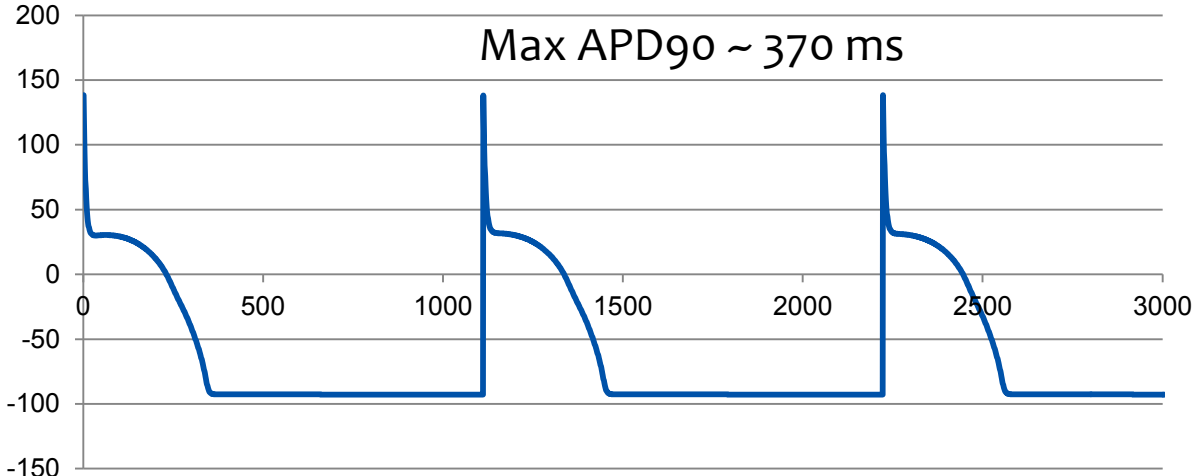
Example 1: Dolasetron – Simcyp PK prediction

po 2.4 mg/kg
Dolasetron
Hydrodolasetron

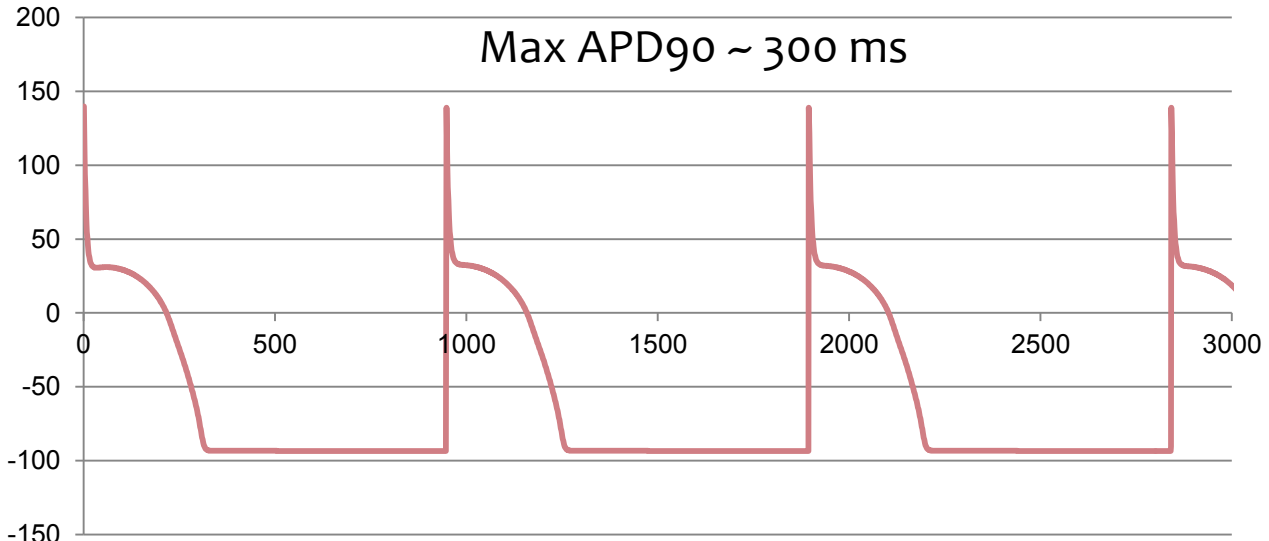


Example 1: Dolasetron – CSS PD effect prediction

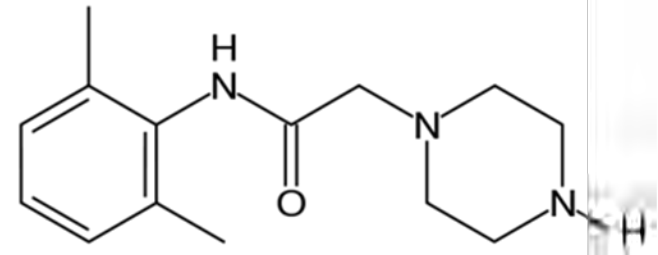
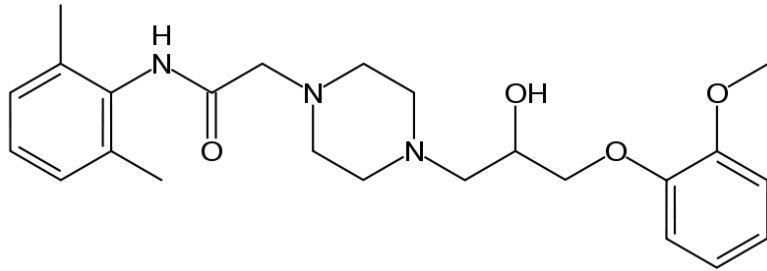
iv 2.4 mg/kg



po 2.4 mg/kg



Example 2 – Ranolazine: IVIVE at the population level



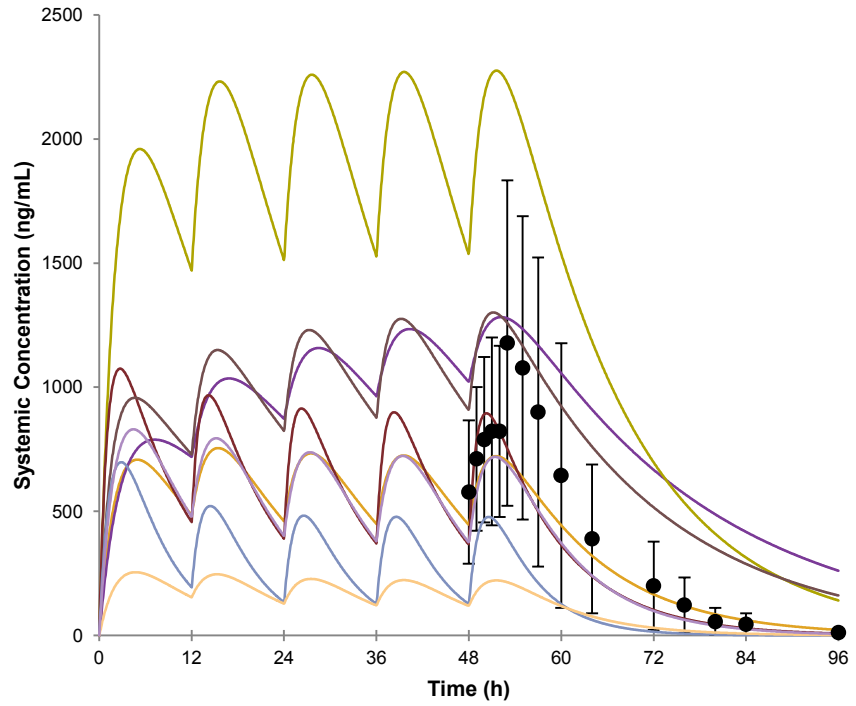
CVT-2738

Ranolazine

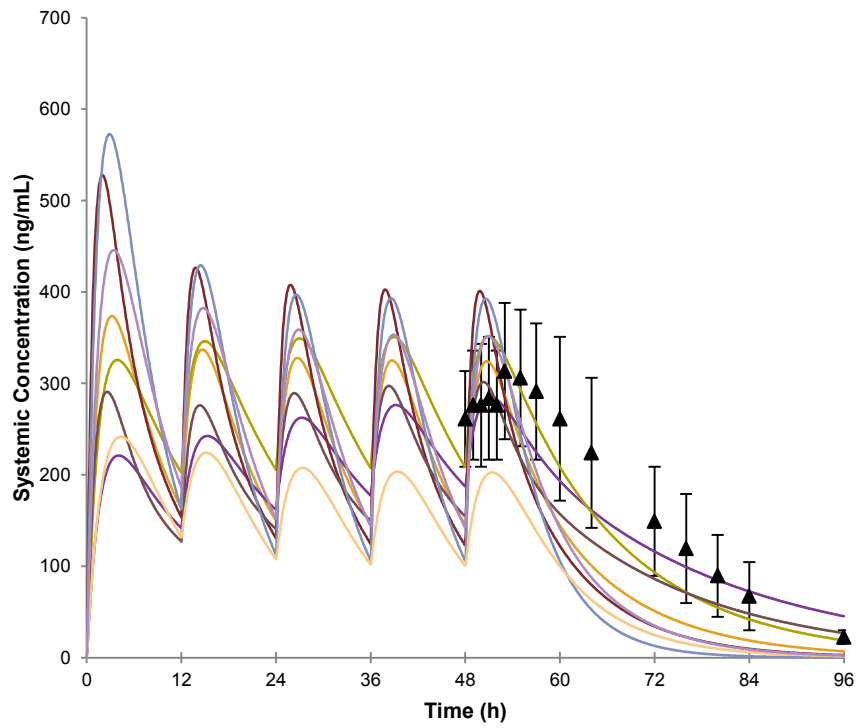
- I_{Kr} inhibitor (*in vitro*)
- multiple other ionic currents inhibition (*in vitro*)
- pharmacologically/electrically active metabolites
- clinically – large variability

Ranolazine and CVT-2738 plasma concentrations

Ranolazine



CVT-2738



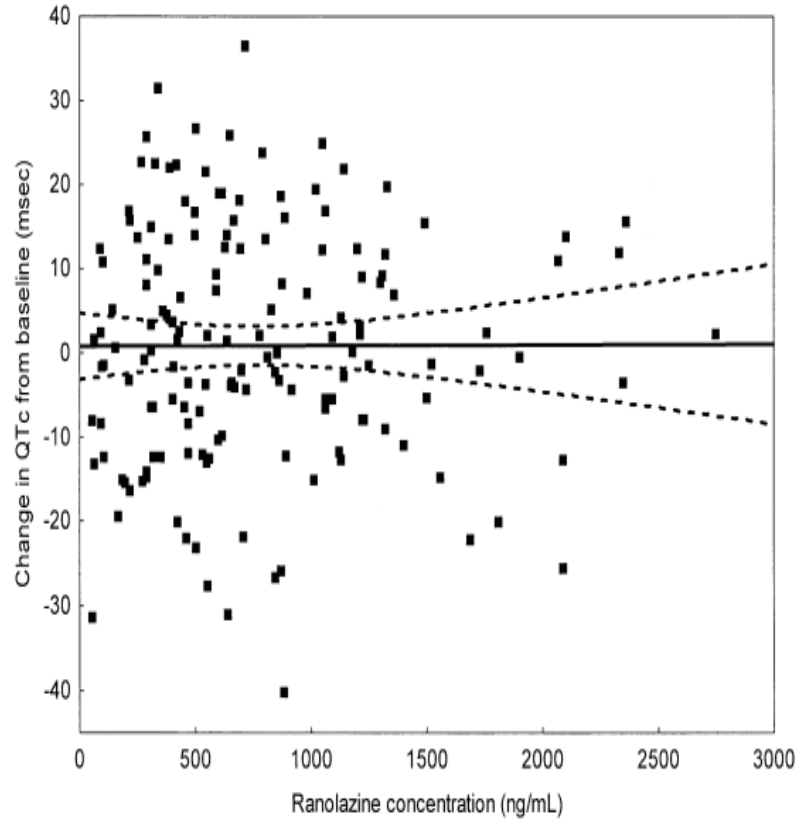
Reasonable description of the plasma concentrations after multiple dosing

Electrophysiology inputs to the model

Ionic current	Ranolazine [IC ₅₀]/n	Source	CVT-2738 [IC ₅₀]	Source
I _{Kr}	12/1	Measured	1.19	QSAR predicted
I _{Ks}	1900/1	Measured	-	
I _{Ca}	311/1	Measured	26.38	QSAR predicted
I _{Na peak}	428/1.63	Measured	-	-
I _{Na late}	6.86/0.71	Measured		

PRO-ARRHYTHMIC POTENCY - IVIVE at the population level - RESULTS

OBSERVED



PREDICTED

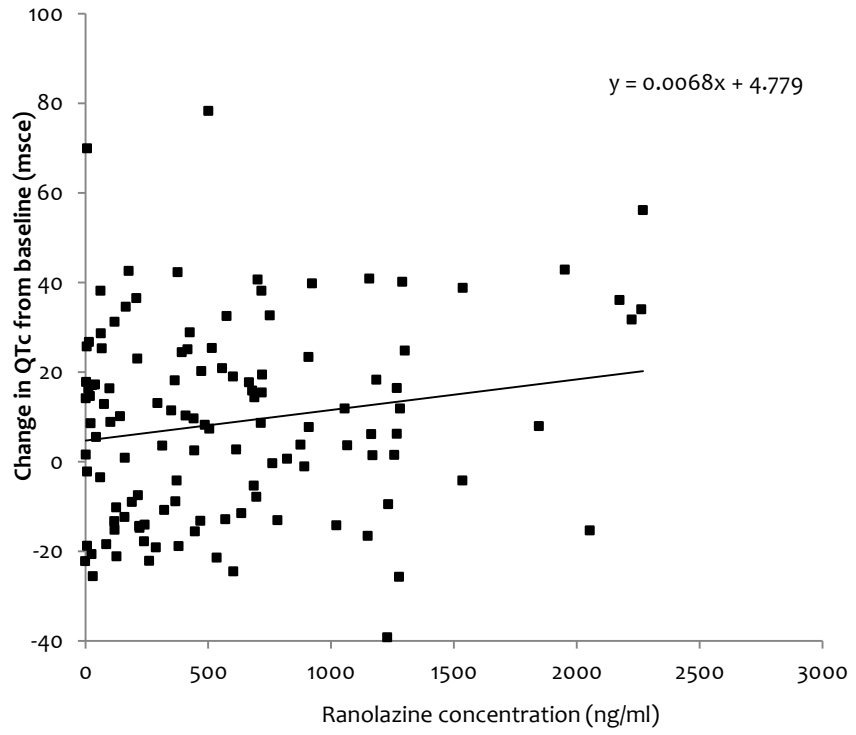


Fig 4. Change in QTc from time-matched baseline values versus ranolazine plasma concentration in healthy control subjects. Linear regression with 95% CI was as follows: $\Delta QTc = 0.82 + 8 \times 10^{-5} \times \text{Ranolazine concentration}$ ($R^2 < 0.001$, $P = 0.97$ for slope [95% CI, -0.004 to 0.004]).

Summary

- Using extrapolated *in vitro* data coupled with PBPK models it is possible to simulate the pharmacokinetics of many drugs
- By incorporating known physiological co-variates it is possible to make simulations in virtual populations rather than an “average” individual
- Concentrations of drugs in tissue compartments of the PBPK model can be linked to mechanistic models to predict side effects/toxicity
- Physiological variability can also be included within the toxicity models

- Acknowledgements
 - Sebastian Polak, Amin Rostami