

In Vitro In Vivo Extrapolation and its Applications in Predicting PK Population Variability

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Outline

- **Clearance concept**
- **In Vitro In Vivo Extrapolation (IVIVE)**
- **Linking PBPK and IVIVE, accounting for variability**
- **Transporters**
- **Industry/Regulators views**
- **Future prospects**

Well-stirred liver model

FACTORS AFFECTING DRUG METABOLISM

James R. Gillette, Ann N Y Acad Sci. 1971

**Commentary: A physiological approach to hepatic clearance
Wilkinson and Shand , CPT, 1975**

$$CL = Q \left[\frac{f_{B,Out} CL_{int}}{f_{B,Out} CL_{int} + Q} \right]$$

Pang and Rowland, JPK Biopharm 1977

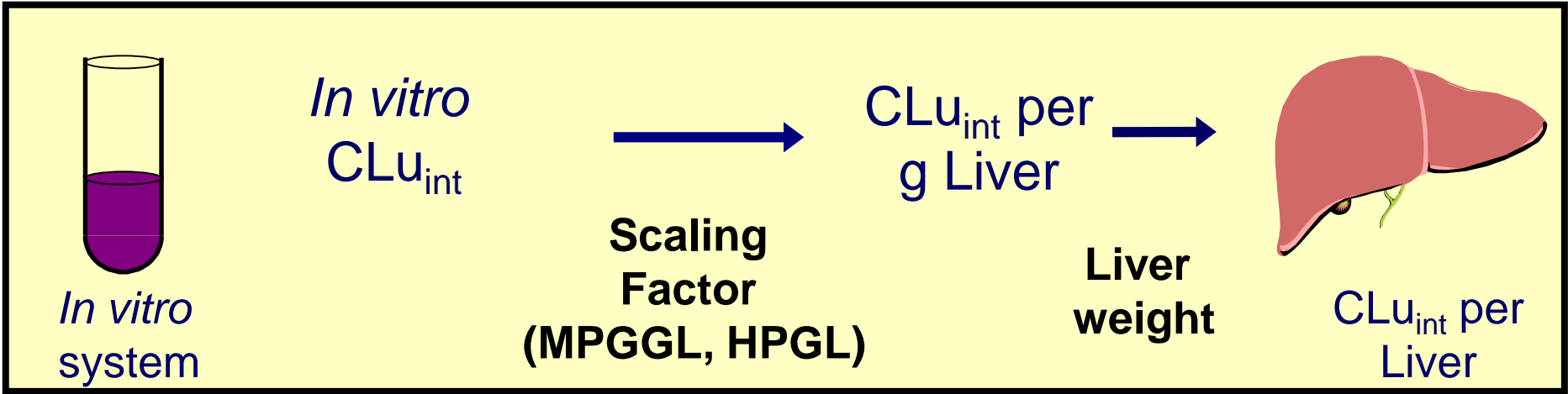
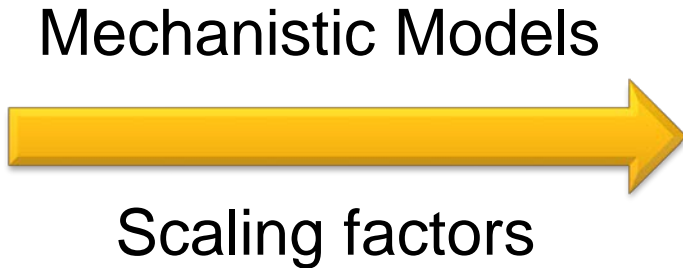
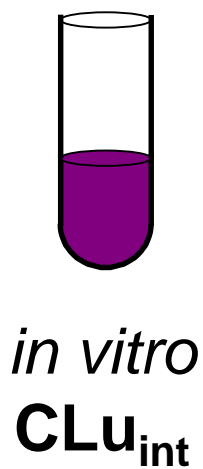
$f_{B,Out}$ = Unbound drug in venous blood /
Whole emergent blood concentration

- Unbound concentration of drug in blood cells equates to the unbound concentration in plasma.
- Emergent venous blood is in equilibrium with that in the liver.

Rowland, Benet and Graham, JPK Biopharm 1973

Yang et al., DMD, 2007

In Vitro - In Vivo Extrapolation (IVIVE)



Accuracy of IVIVE approaches for human **CL** or **CLint**

System	AFE	Ref
HLM	2.3	Obach DMD 27, 1350, 1999
	6.2	Ito, Pharm Res, 22, 103, 2005
	2.3	Stringer, Xeno, 38, 1313, 2008
	2.2	Ring, J Pharm Sci, 100, 490, 2011
	5	Hallifax, Pharm Res, 27, 2150, 2010
	2	Jones, Clin Pk, 50, 311, 2011
Heps	2.4	De Buck DMD, 35, 1766, 2007
	5.2	Stringer, Xeno, 38, 1313, 2008
	5	Hallifax 2011
	7.6	Naritomi DMD 31, 580, 2003
Recombinant CYP	1.53 PT 2.15 WS	Stringer DMD, 37,1025, 2009

Generally many literature studies shows under-prediction from *in vitro* systems.
 Can be corrected using an empirical scaling factor.
 Need to understand for your *in vitro* system if this is necessary.

IVIVE predictions – Improvements over years

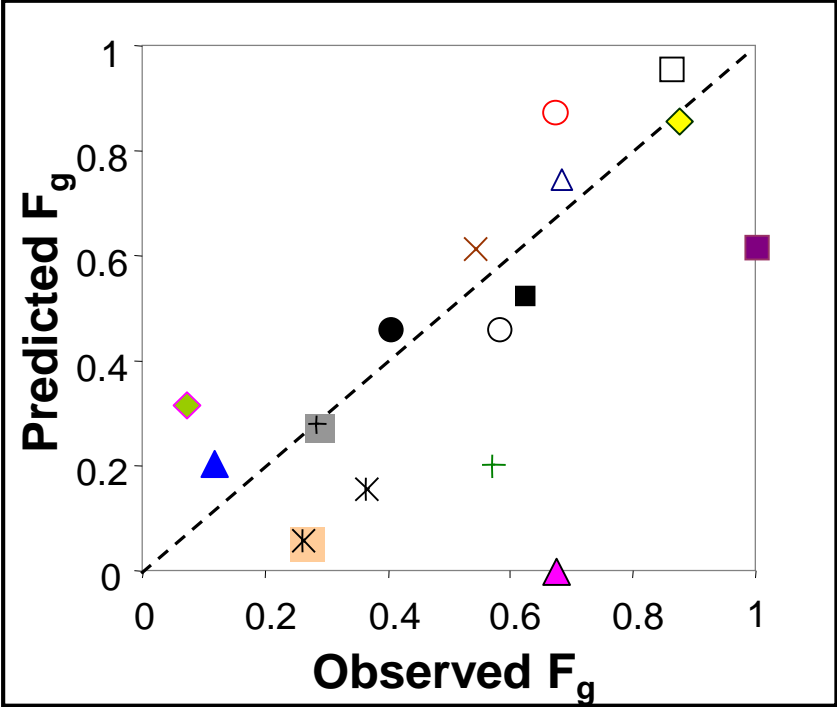
- Non-specific binding (Obach, DMD, 1999, Riley et al., DMD, 2005)
- Recombinant CYPs and ISEF values (Galetin et al., DMD, 2004; Proctor et al. Xenobiotica, 2004)
- In vitro modelling to account for hepatic uptake (Soars et al., DMD, 2007)
- Adding BSA and HAS-FAS to HLM (Rowland et al., DMD, 2008)
- Accounting for the difference in drug ionization in extracellular and intracellular tissue water (Berezhkovskiy, J Pharm Sci, 2011)
- Integrating uptake, metabolism, biliary excretion, and sinusoidal efflux (Umehara and Camenisch, Pharm Res, 2012)
- Incorporating ionisation and protein binding (Poulin et al., J Pharm Sci, 2012)

Gut wall metabolism

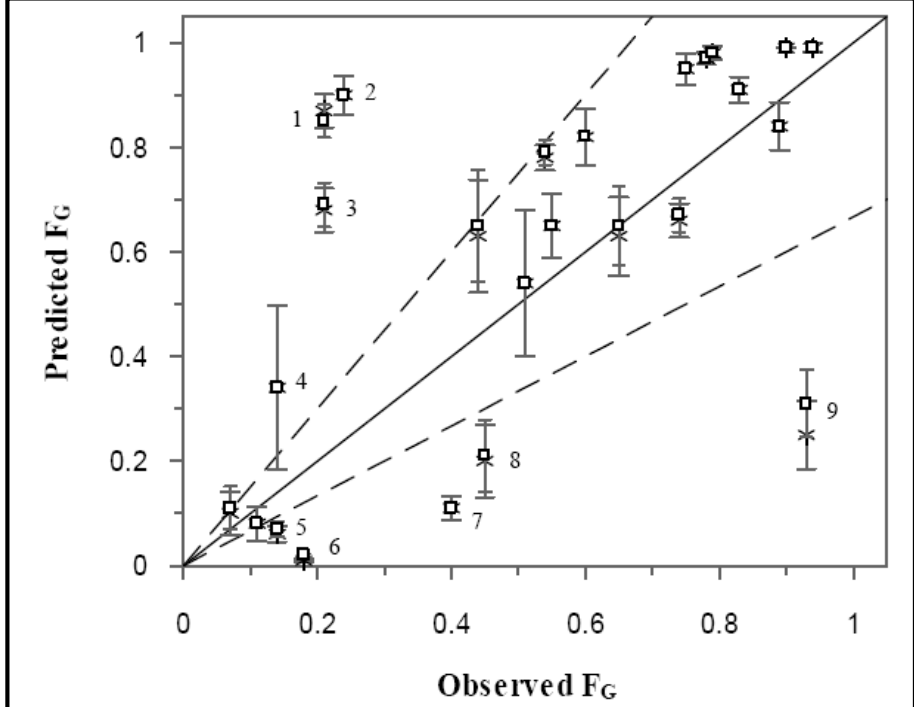
'Q_{gut}', a minimal model

$$F_g = \frac{'Q_{gut}'}{'Q_{gut}' + fu_{gut} \cdot CLu_{int-gut}}$$

$$'Q_{gut}' = \frac{CL_{perm} \cdot Q_{villi}}{CL_{perm} + Q_{villi}}$$

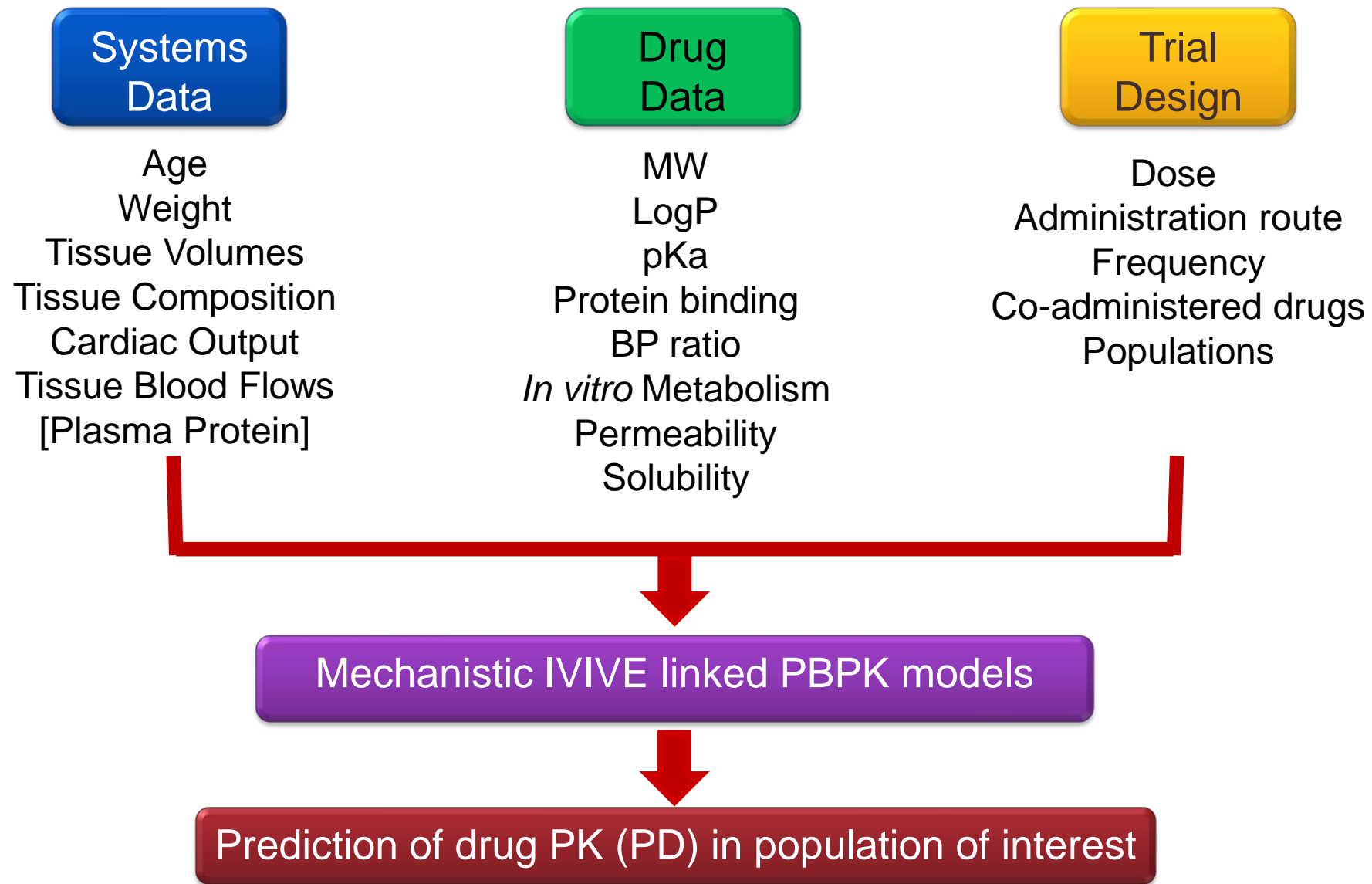


Yang et al., CDM, 2007



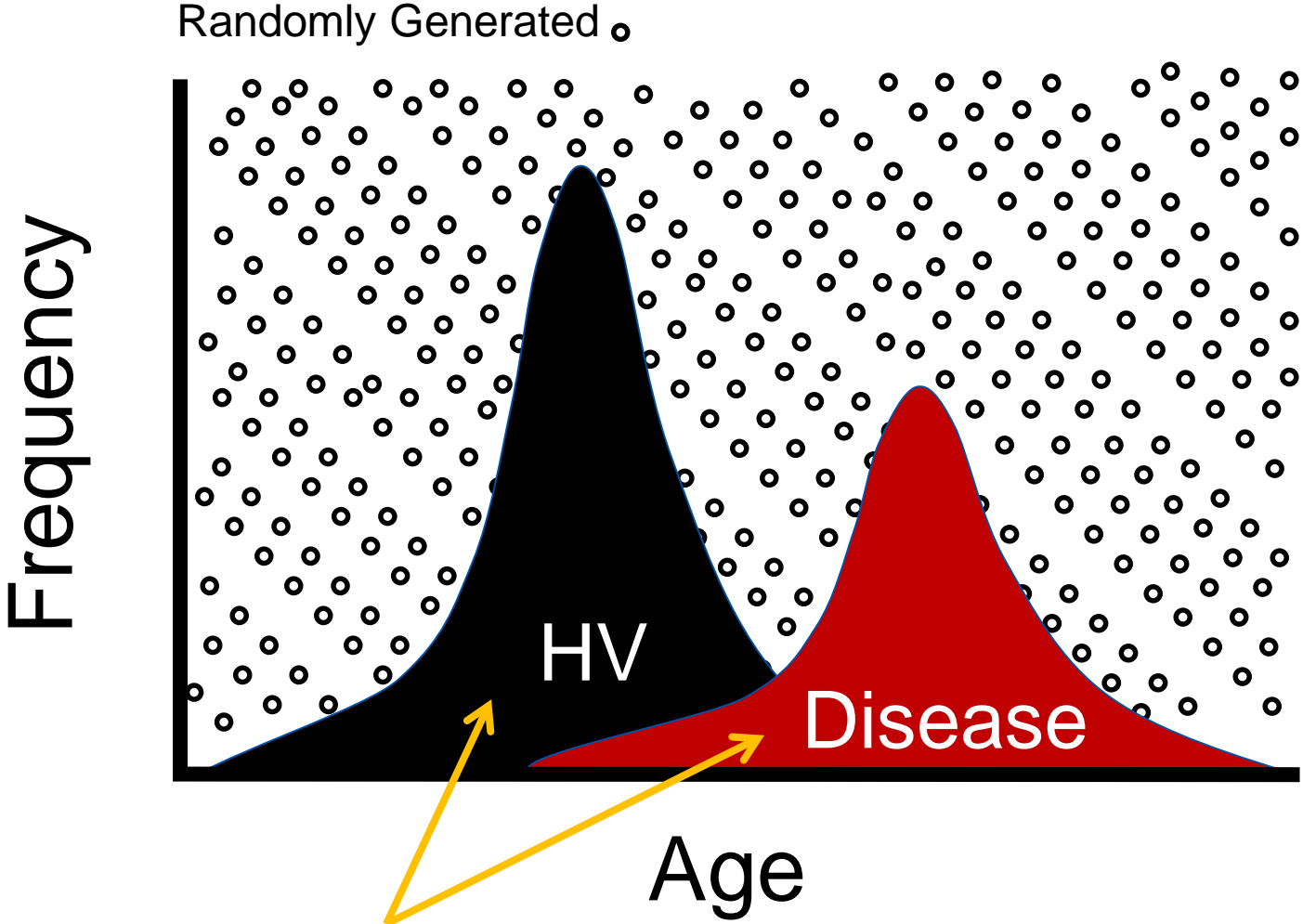
Gertz et al., DMD, 2010

Special populations

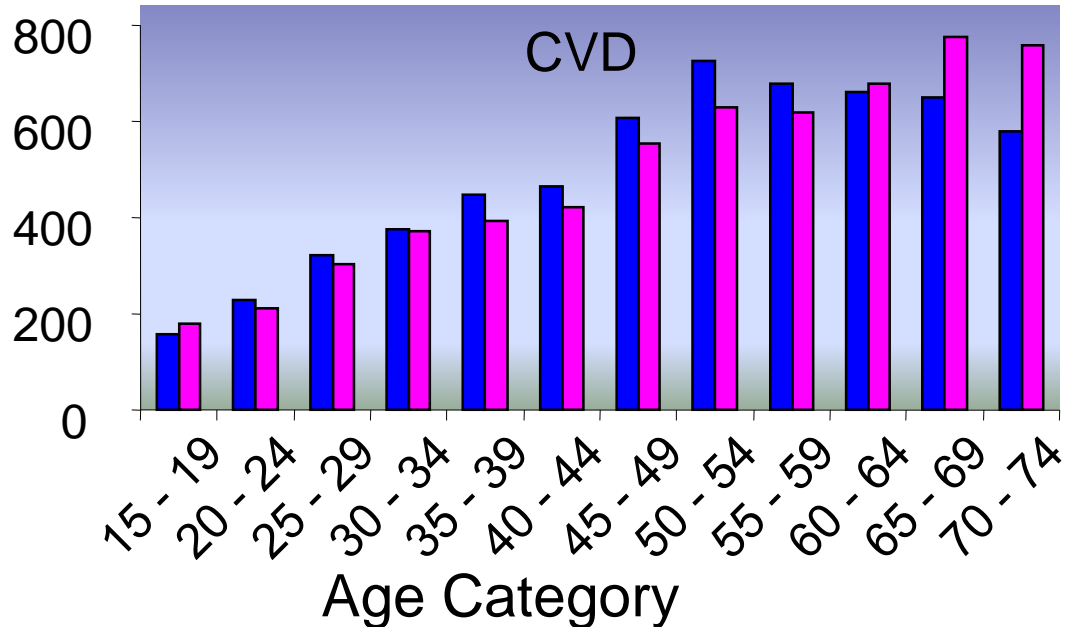
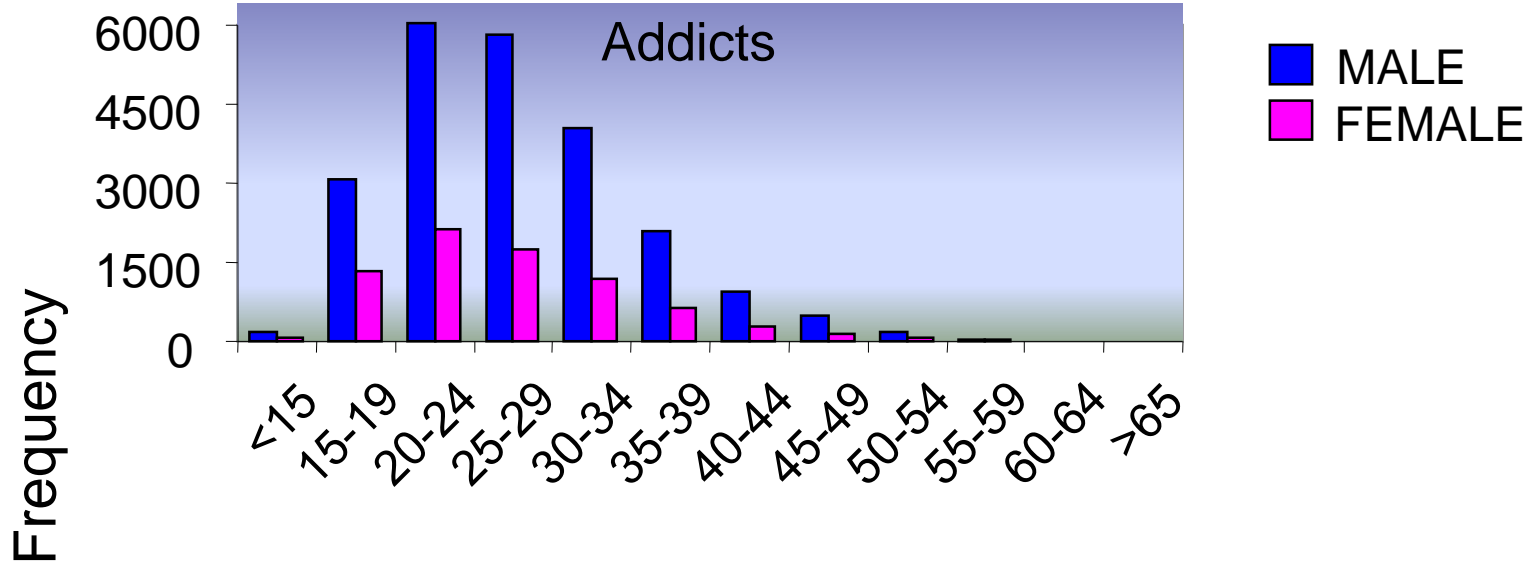


Jamei *et al.*, DMPK, 2009, Rostami-Hodjegan, CPT, 2012

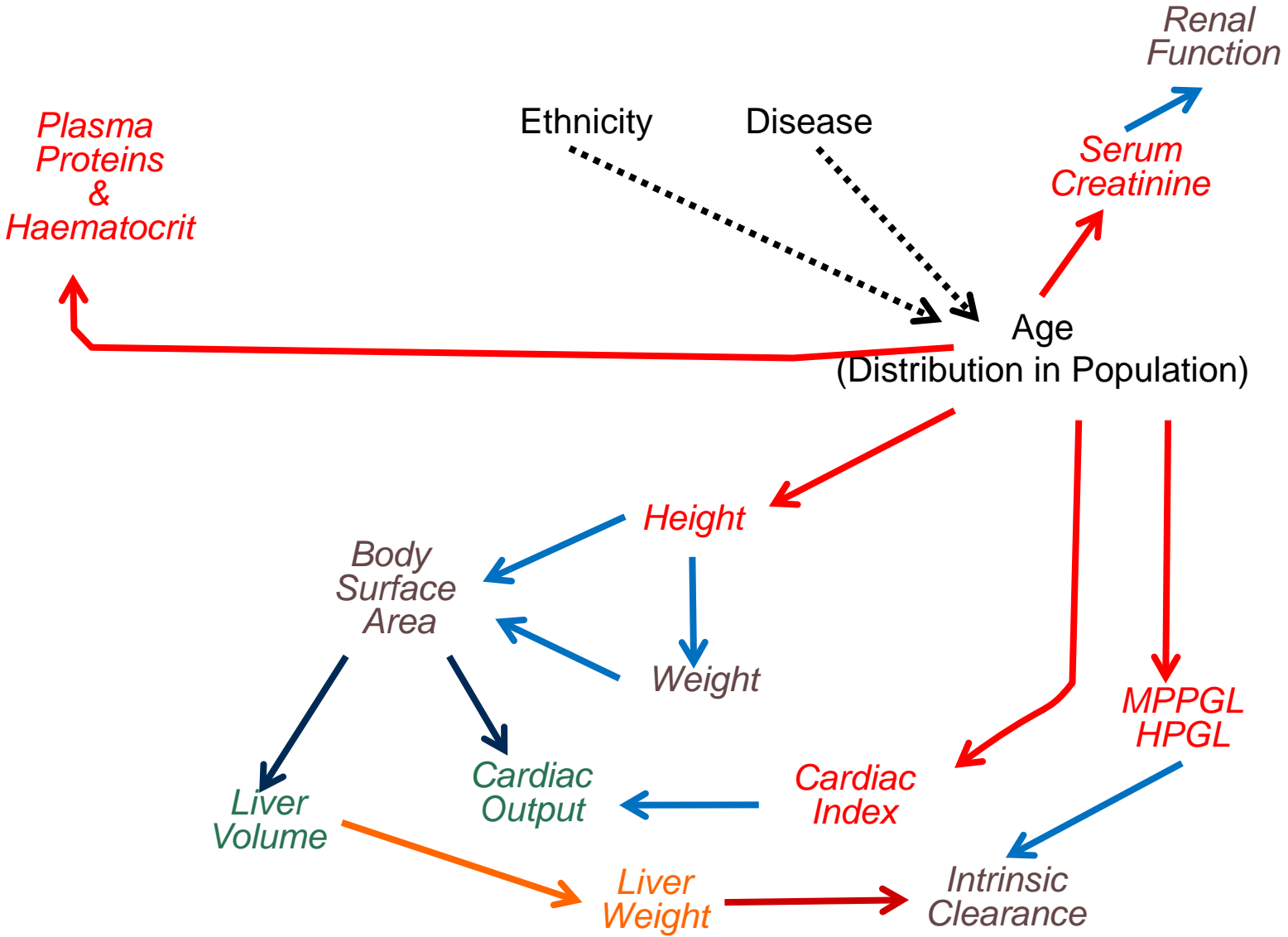
Demographic Features of Healthy and Disease Populations



Age Distribution in Target Population



The Complexity of Covariate Effects as Applied to CL



Converting CLu_{int} to CL_H

$$MPPGL = 10 \left(1.407 + 0.0158 \times \text{age} + 0.00038 \times \text{age}^2 + 0.0000024 \times \text{age}^3 \right)$$

$$CLu_{int} = CLu_{int} \cdot MPPGL \cdot \text{Liver Weight}$$

(whole liver)

$$\text{Liver Weight} = \text{Liver Volume} \times \text{Liver Density}$$

$$\text{Liver Volume} = 0.722 \cdot BSA^{1.176} \text{ (L/m}^2\text{)}$$

$$0.00718 \times Ht^{0.725} \times Wt^{0.425}$$

$$f(\text{age}) + x$$

Converting CLu_{int} to CL_H

$$CLu_{int} = CLu_{int} \times MPPGL \times Liver\ Weight$$

$$CL_H = \frac{Q_H \times fu_B \times CLu_{int}}{Q_H + fu_B \times CLu_{int}}$$

$$fu_B = \frac{fu}{C_B/C_p}$$

$$Q_H = \%CO$$

$$C_B/C_p = (E:P) \times HC + (1 - HC)$$

$$CO = f(\text{age}, BSA)$$

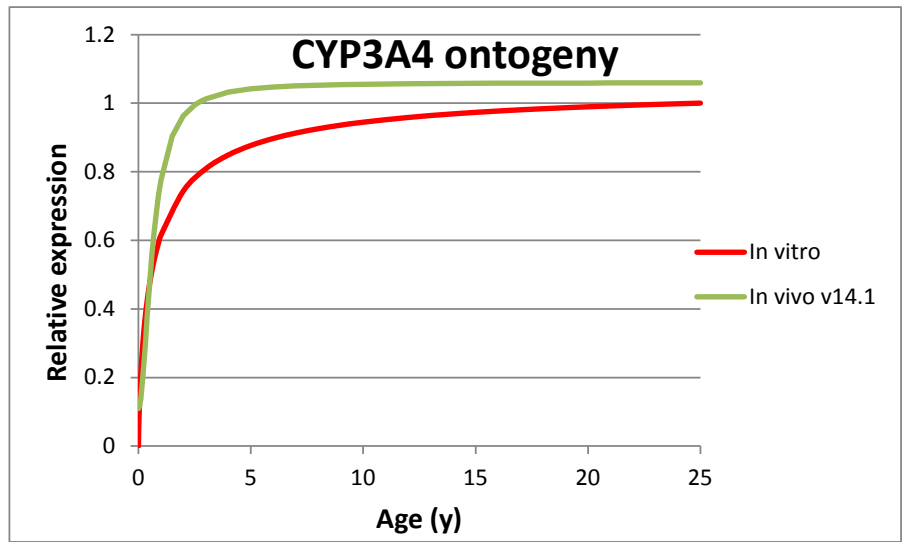
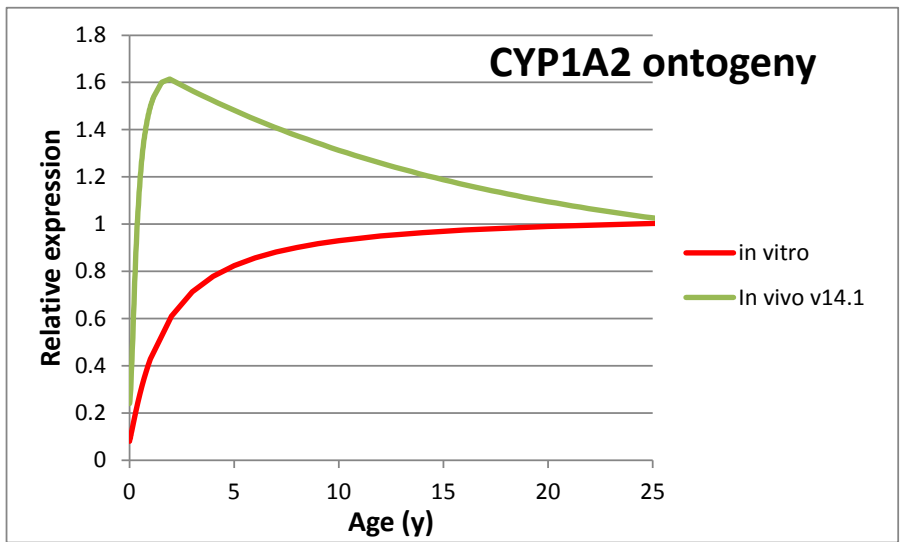
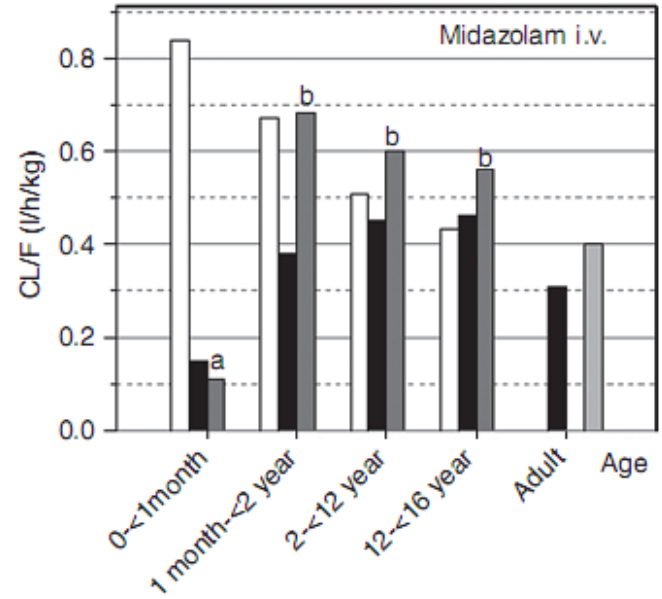
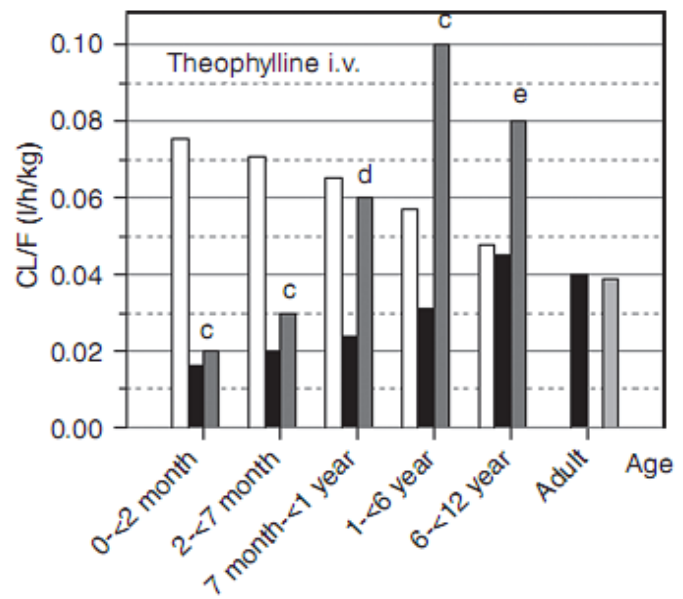
$$HC = f(\text{age}) + f(\text{sex})$$

$$0.00718 \times Ht^{0.725} \times Wt^{0.425}$$

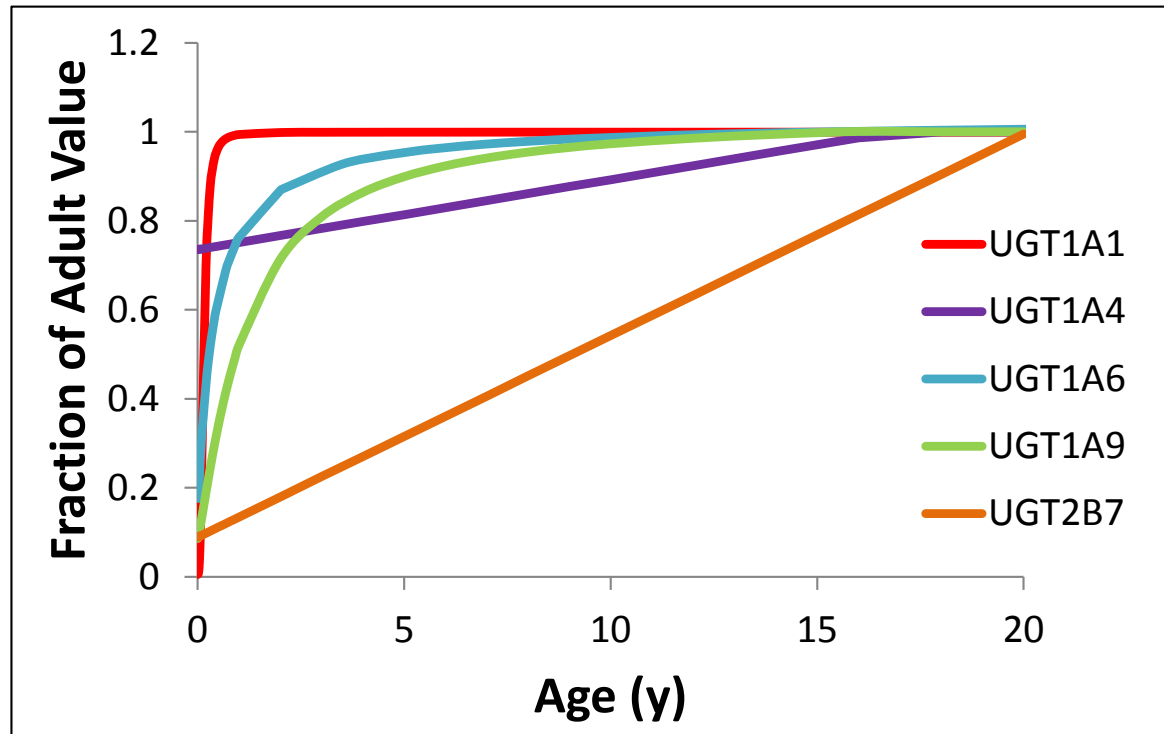
$$f(\text{age}) \times x$$

Revised *in vivo* ontogeny functions for CYP1A2 and 3A4

(Leong *et al.*, CPT 2012; 91: 926-931)



UGT Ontogeny

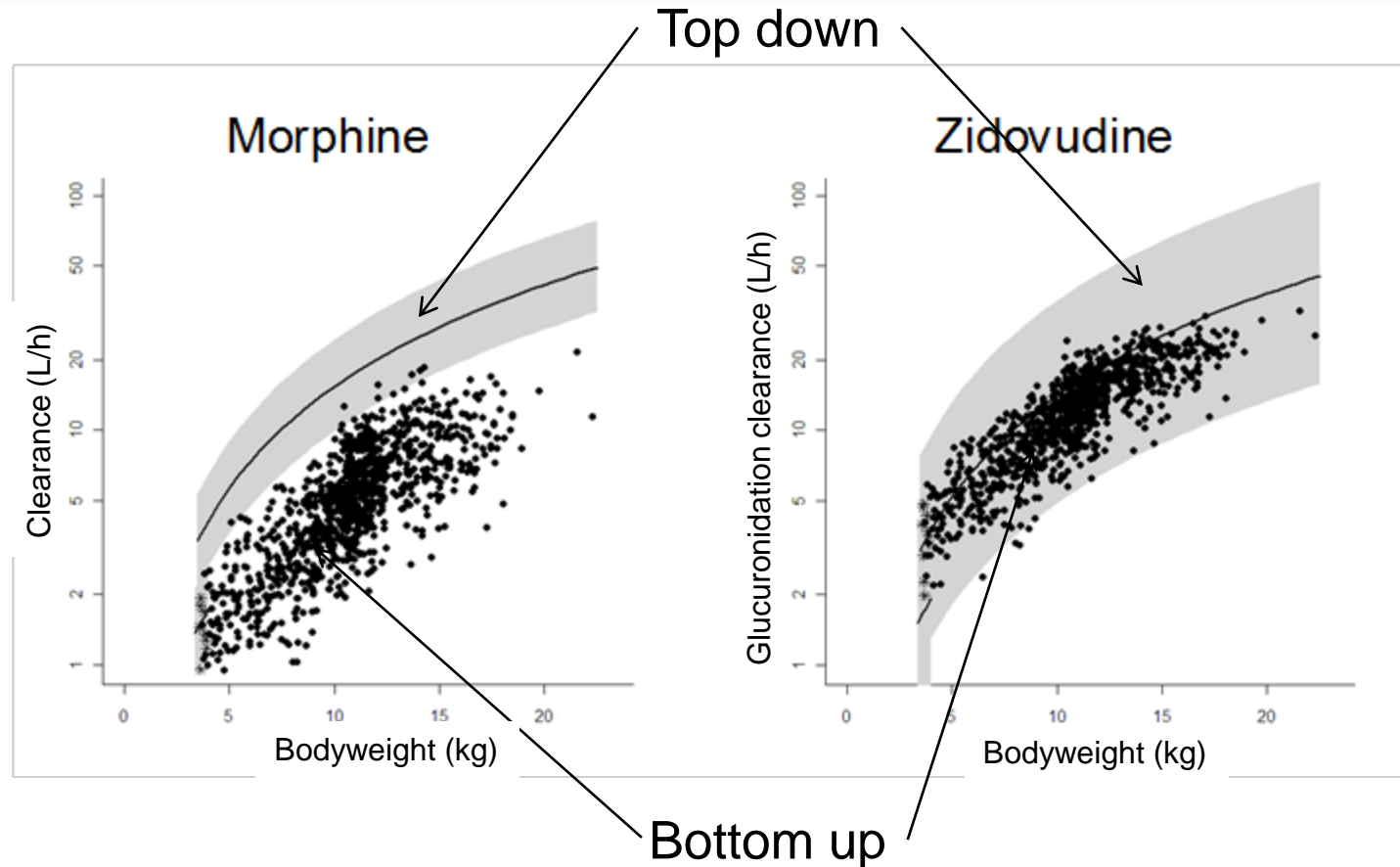


Strassburg *et al* 2002
Burchell *et al* 1989
Onishi *et al* 1997
Leakey *et al* 1987
Coughtrie *et al* 1988
Miyagi and Collier 2007
Zaya *et al* 2006
Pacifci *et al* 1990
Pacifci *et al* 1982
Choonara *et al* 1989

Leiden Collaboration – Top down vs bottom up ontogeny for UGT2B7

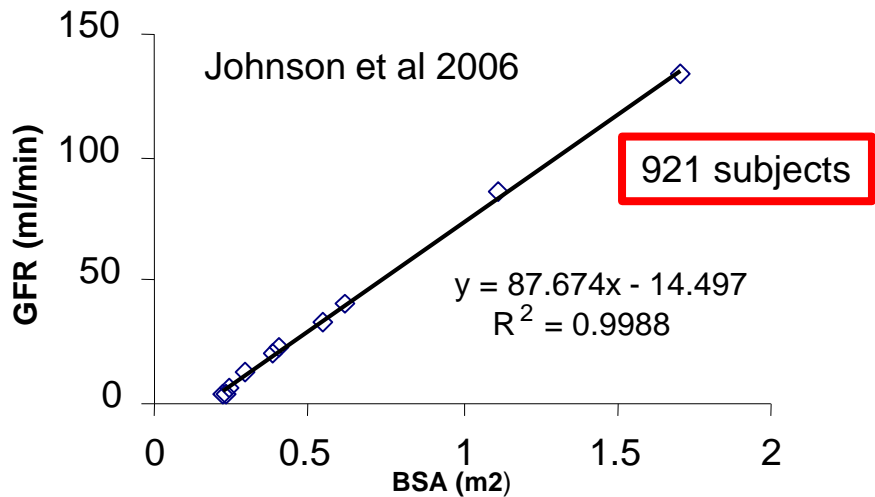
- Morphine
- Zidovudine

UGT2B7 ontogeny 'Top down' vs 'Bottom up'



- Take home message is that pattern of ontogeny appears to be reasonable except for early neonates
- But under-prediction of CL across age band with morphine.

Maturation of Renal Clearance



Pediatr Nephrol (2009) 24:67-76
DOI 10.1007/s00467-008-0997-5

ORIGINAL ARTICLE

Human renal function maturation: a quantitative description using weight and postmenstrual age

Malin M. Rhodin · Brian J. Anderson · A. Michael Peters · Malcolm G. Coulthard · Barry Wilkins · Michael Cole · Etienne Chatelut · Anders Grubb · Gareth J. Veal · Michael J. Keir · Nick H. G. Holford

923 subjects

Pharm Res (2014) 31:2643-2654
DOI 10.1007/s1095-014-1361-z

RESEARCH PAPER

Simultaneous Pharmacokinetic Modeling of Gentamicin, Tobramycin and Vancomycin Clearance from Neonates to Adults: Towards a Semi-physiological Function for Maturation in Glomerular Filtration

Rozsmanjn F.W., De Cock · Karel Allegaert · Jannelle M. Brussee · Catherine A. J. van den Broek · Johannes N. van den Anker · Merdort Danhof · Catherine A. J. Jibbe

1760 subjects

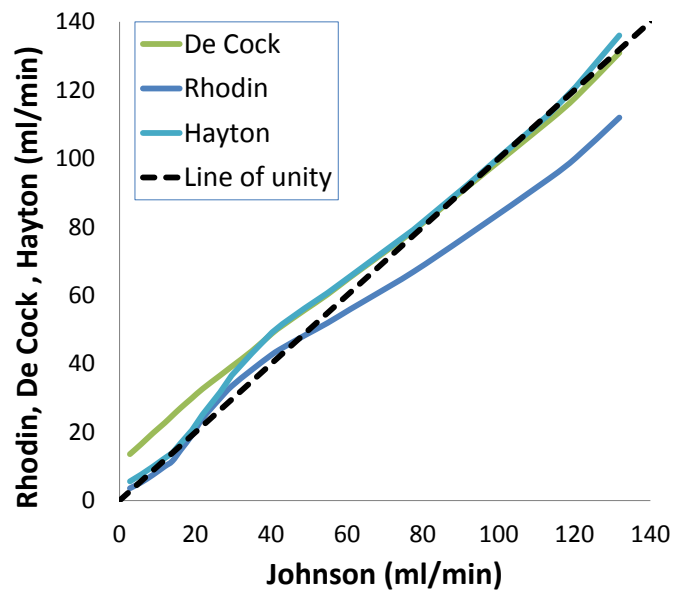
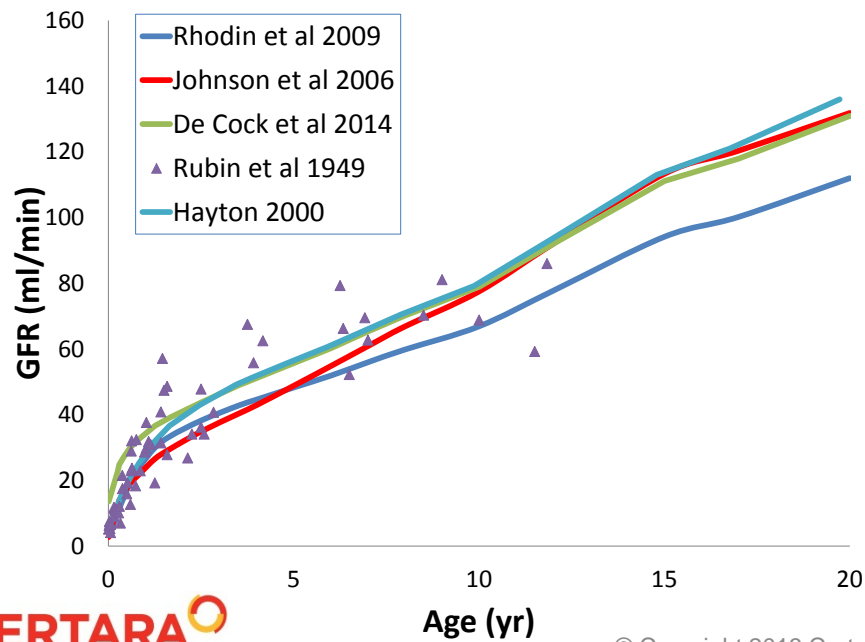
AAPS PharmSci 2002; 2 (1) article 3 (<http://www.aapspharmsci.org>).

Maturation and Growth of Renal Function: Dosing Renally Cleared Drugs in Children

Submitted: September 6, 1999; Accepted: February 15, 2000; Published: March 3, 2000

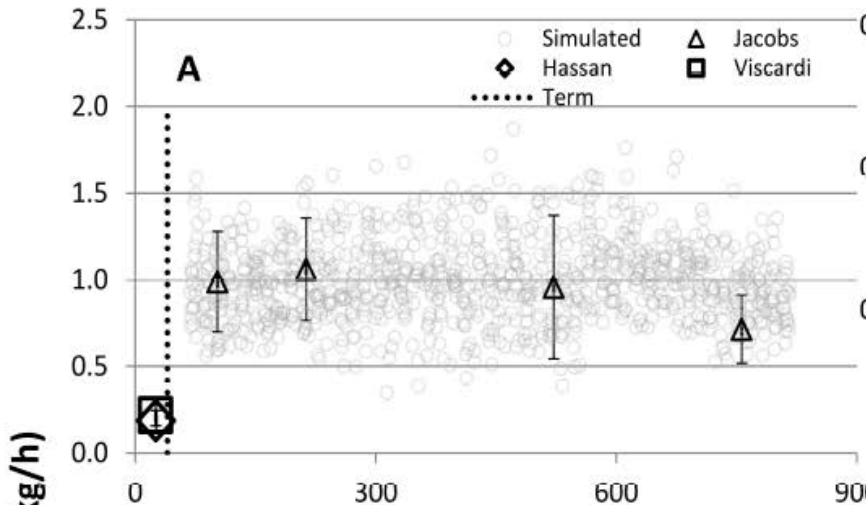
William L. Hayton¹

63 subjects

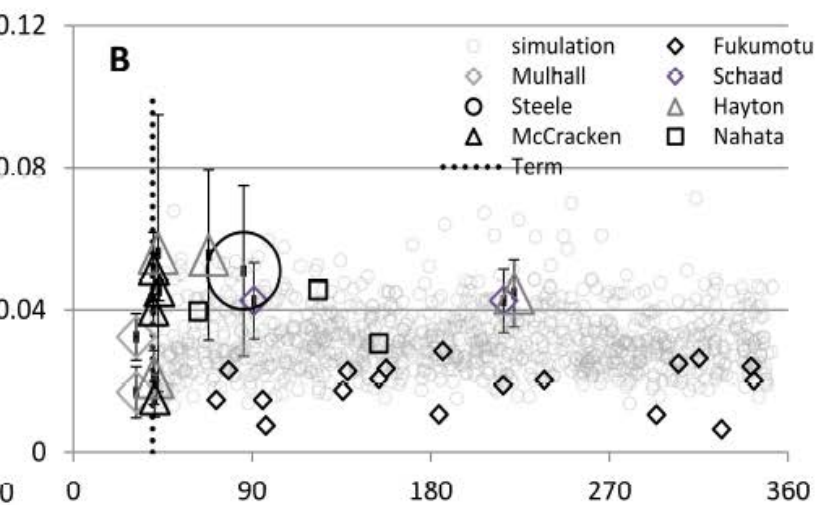


Maturation of Biliary Clearance Appears to be Rapid

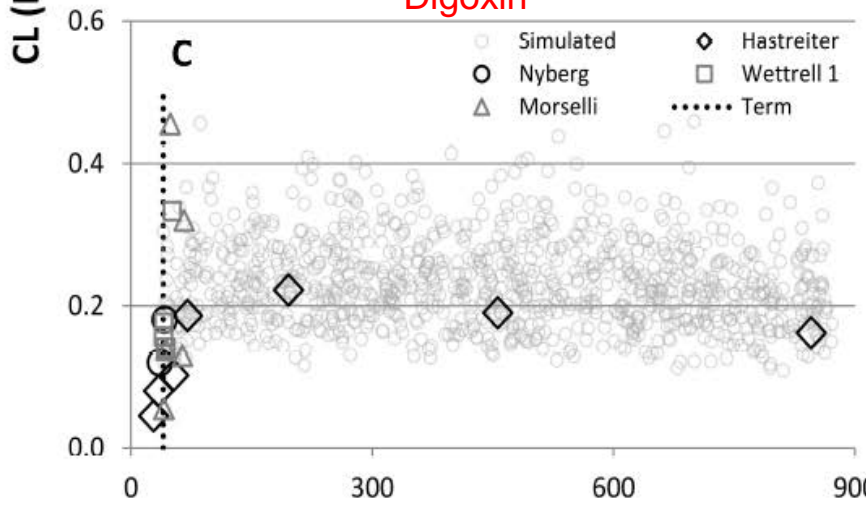
Azithromycin



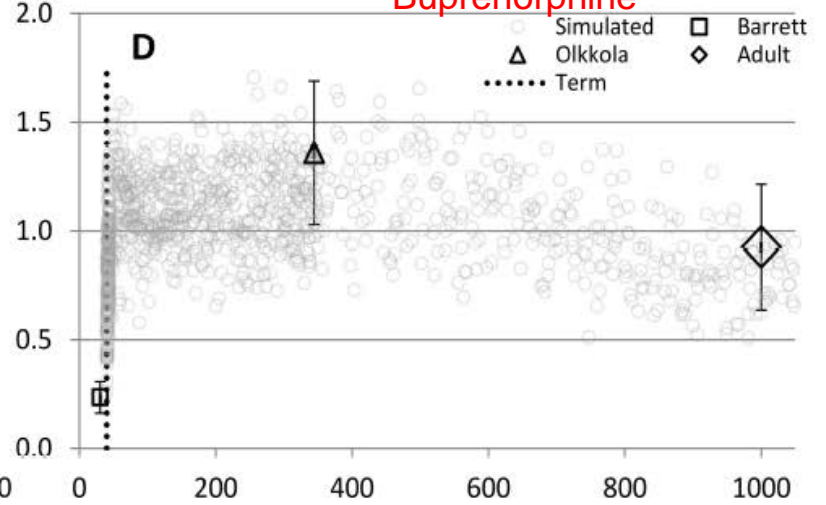
Ceftriaxone



Digoxin



Buprenorphine

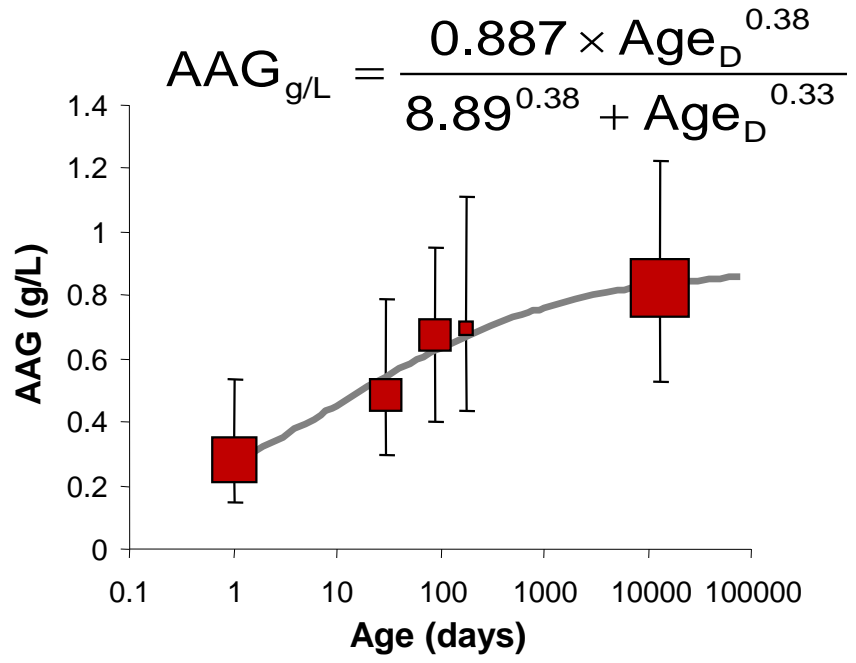
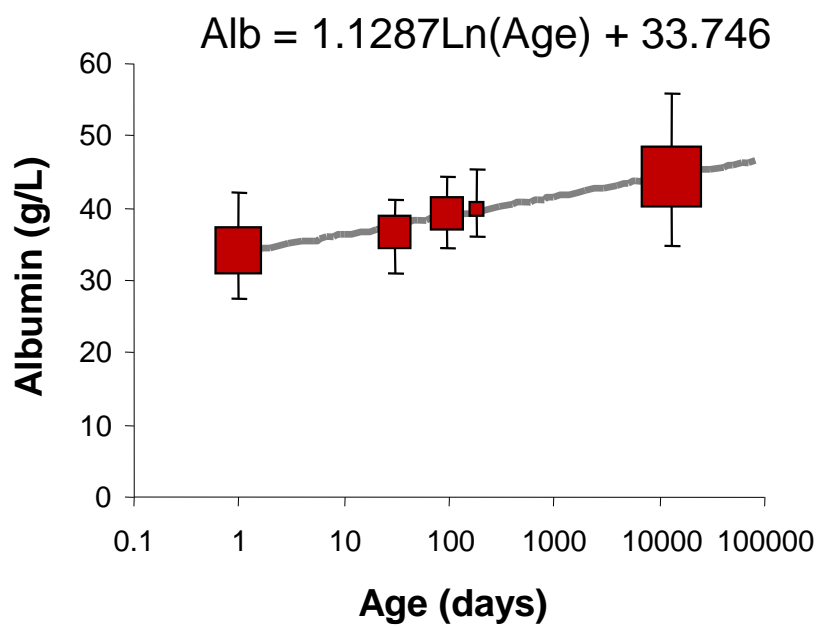


PMA (weeks)

Johnson et al Drug Metab Dispos. 2016



Variation in Protein Binding (fu)



$$f_u = \frac{1}{1 + \frac{[P]}{K_D}}$$

K_D = Dissociation Constant
 [P] = Serum Protein Concentration

$$K_D = \frac{[P]}{\frac{1}{f_u} - 1}$$

In absence of changes in dynamics of binding:

$$f_u = \frac{1}{1 + \left[\frac{[P]}{[P]_{pop^*}} \times \frac{(1 - f_{u_{pop^*}})}{f_{u_{pop^*}}} \right]}$$

*pop is the population under investigation *i.e* paediatric

Developing and testing a Geriatric population

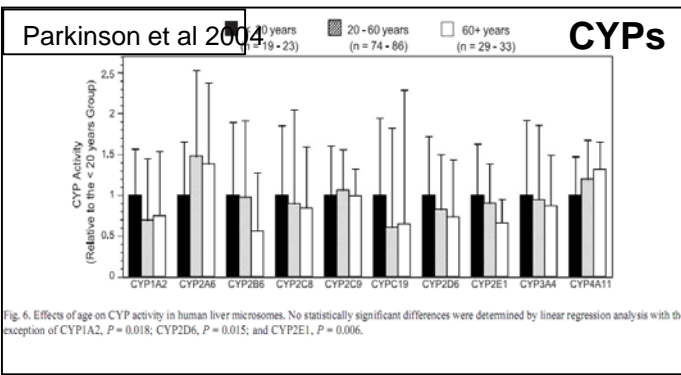
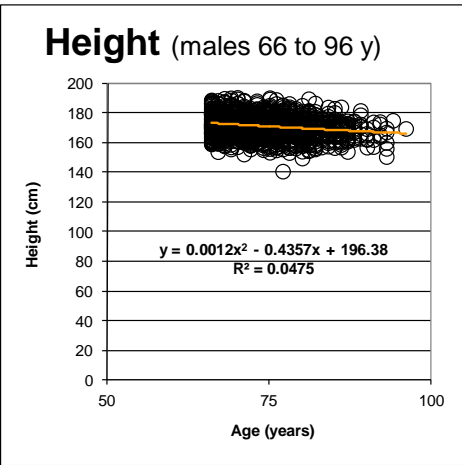
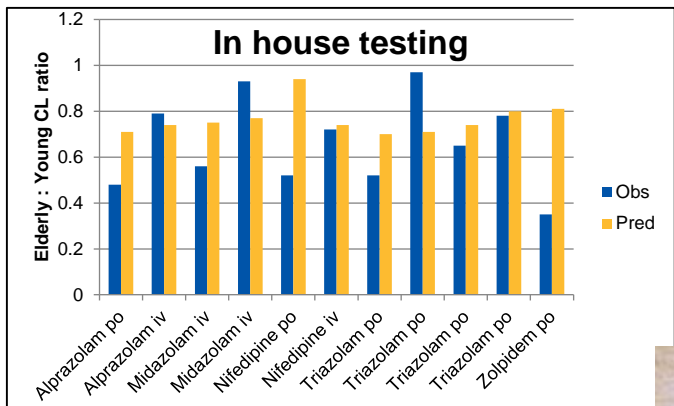
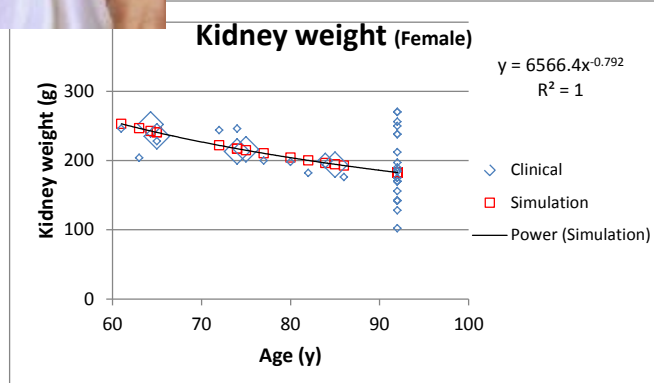
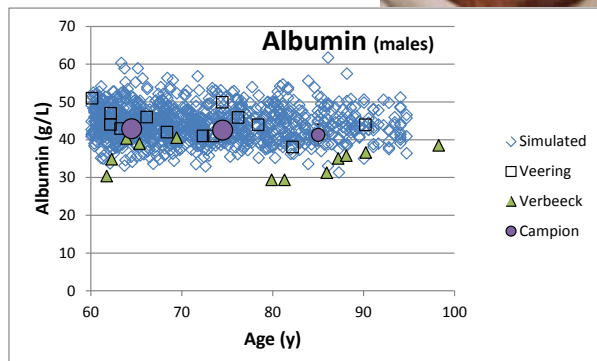
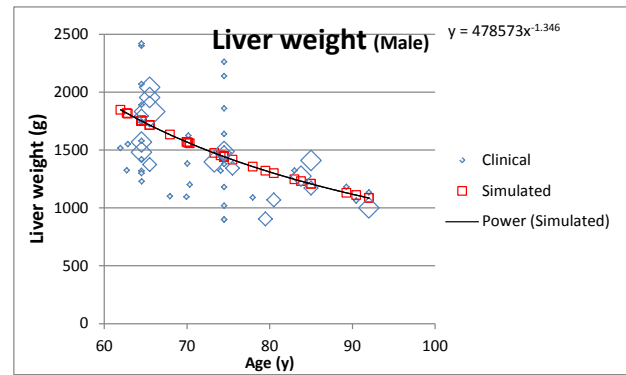
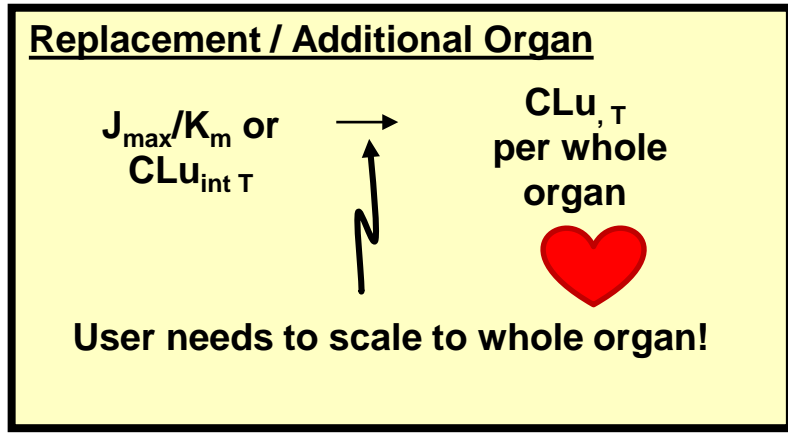
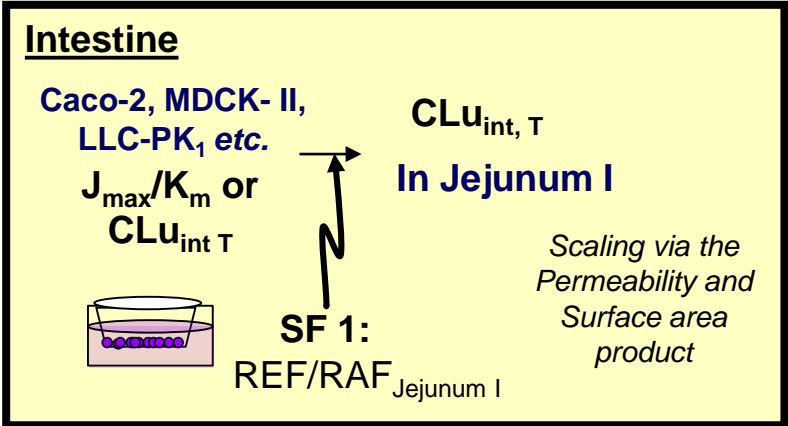
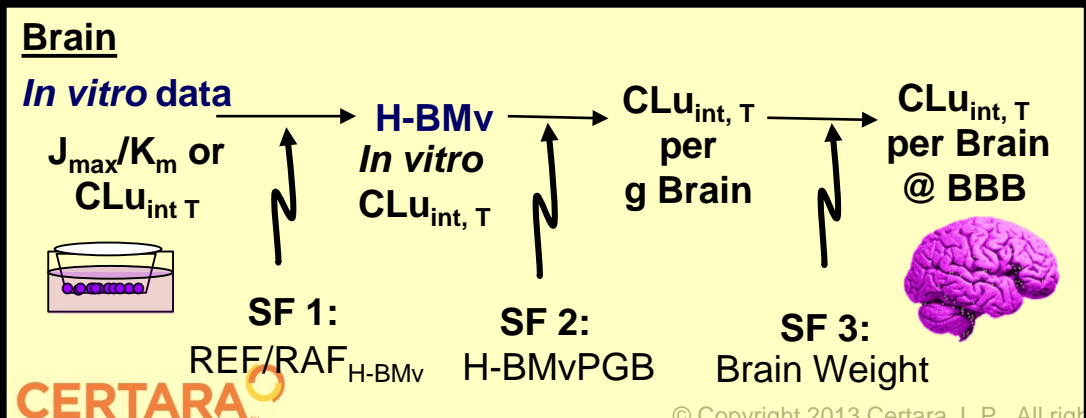
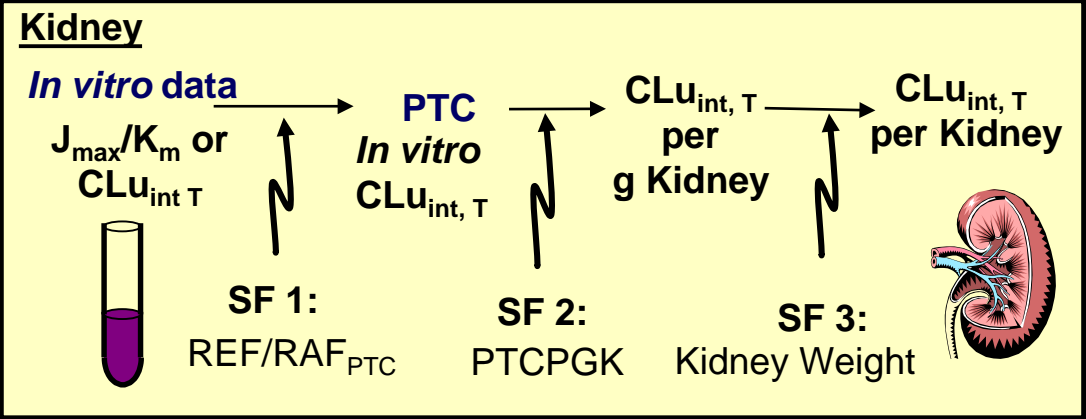
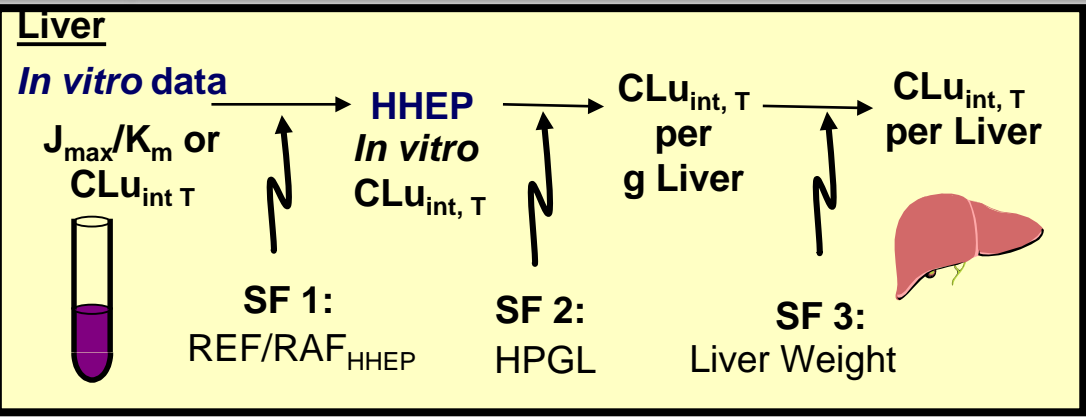


Fig. 6. Effects of age on CYP activity in human liver microsomes. No statistically significant differences were determined by linear regression analysis with the exception of CYP1A2, $P = 0.018$; CYP2D6, $P = 0.015$; and CYP2E1, $P = 0.006$.



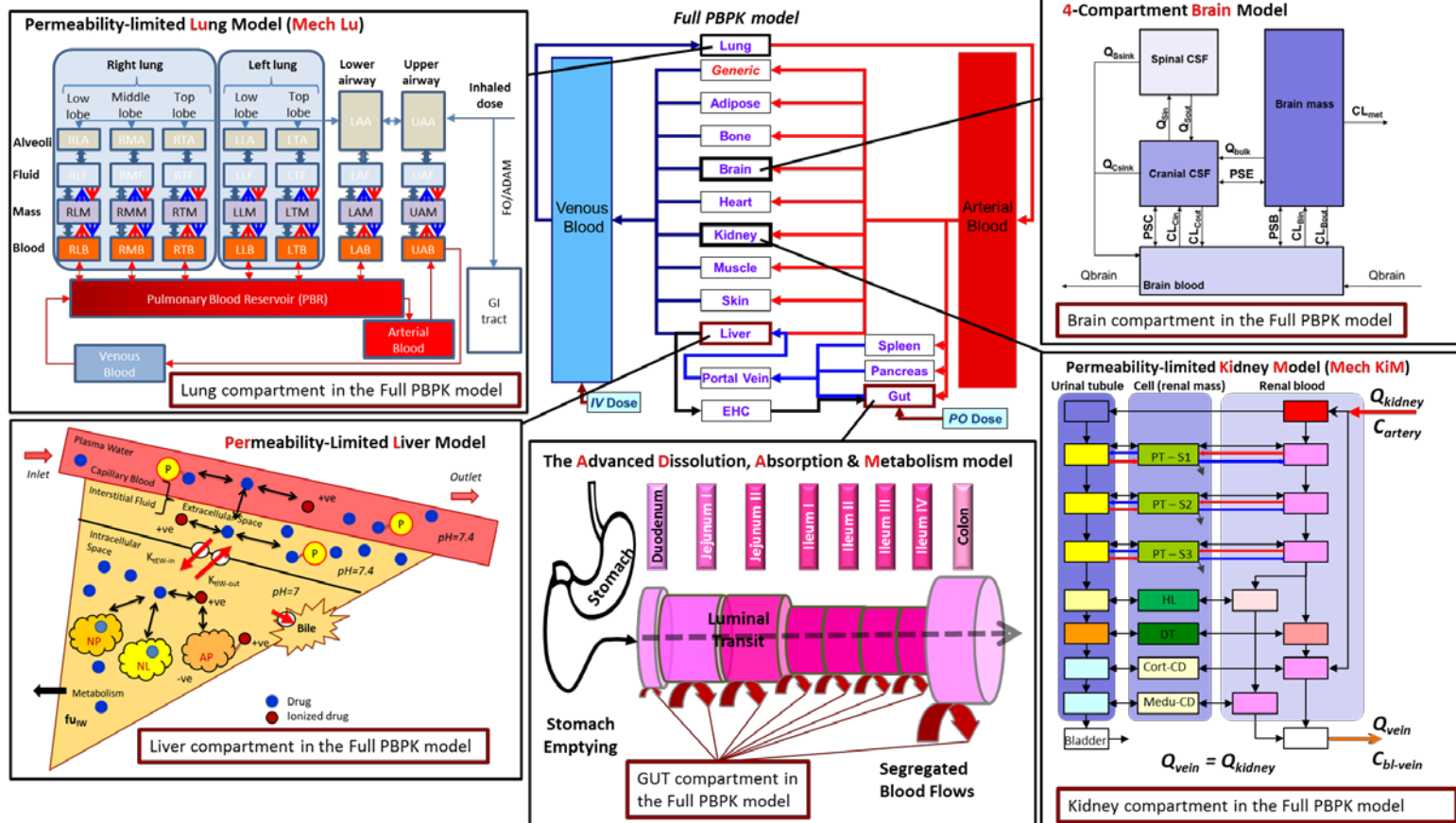
Scaling from in vitro: drug data vs systems data



SF: Scaling Factor

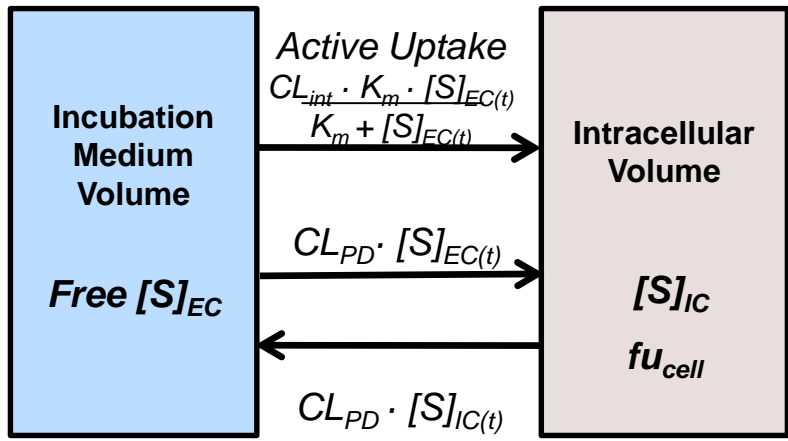
Translating *in vitro* effective concentrations to concentrations at the site of action

- Mechanistic, multi-compartmental tissue models (brain, kidney, liver, lung and intestine) are available
- Enable more reliable estimates of intracellular tissue concentrations



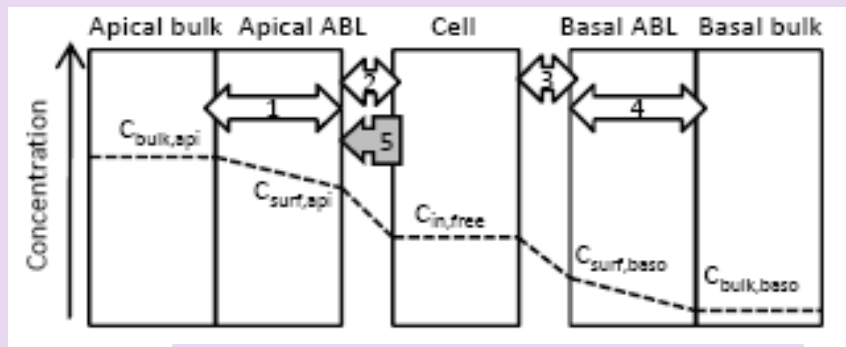
Modelling *in vitro* assays – a must to do!

2 Compartment Model

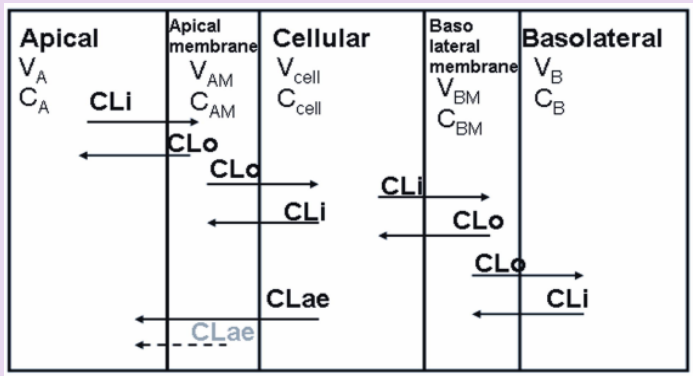


Baker et al., Xenobiotica, 2007; Soars et al., Mol Phar, 2009; Poirier et al., Mol Pharm, 2009; Menochet et al., J Pharm Exp Ther, 2012

5 Compartment Model - Transwell






















Heikkinen et al., 2010 Mol Pharmaceutics

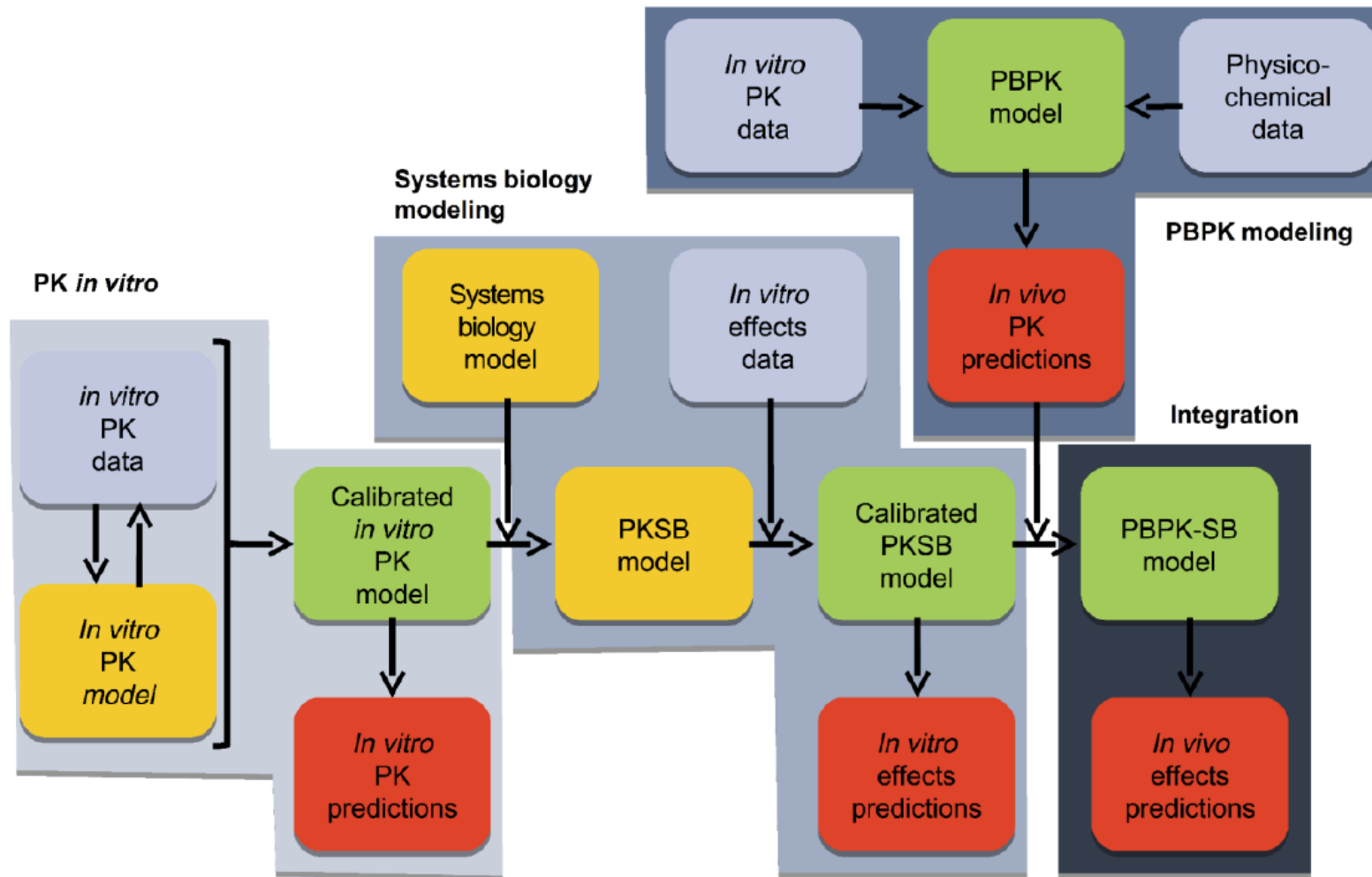


Korzekwa et al., 2012 DMD

PBPK Impact on 19 US Drug Labels in Last 2 Years

<p>Olysio (Simeprevir) Hepatitis C</p> 	<p>Xarelto (Rivaroxaban) Thrombosis & Embolism</p> 	<p>Edurant (Rilpivirine) HIV infection</p> 	<p>Imbruvia (Ibrutinib) Lymphoma and Leukemia</p> 	<p>Opsumit (Macitentan) Pulmonary Hypertension</p> 
<p>Zykadia (Ceritinib) Lung Cancer</p> 	<p>Odozmzo (Sonidegib) Basal Cell Carcinoma</p> 	<p>Farydak (Panobinostat) Multiple myeloma</p> 	<p>Revatio (Sildenafil) Pulmonary Hypertension</p> 	<p>Bosulif (Bosutinib) Myelogenous Leukemia</p> 
<p>Lynparza (Olaparib) Advanced Ovarian Cancer</p> 	<p>Movantik (Naloxegol) Opioid Induced Constipation</p> 	<p>Tagrisso (Osimertinib) Metastatic NSCLC</p> 	<p>Iclusig (Ponatinib) Chronic Myeloid Leukemia</p> 	<p>Cerdelga (Eliglustat) Gaucher Disease</p> 
<p>Jevtana (Cabazitaxel) Prostate Cancer</p> 	<p>Cotellic (Cobimetinib) Metastatic Melanoma</p> 	<p>Lenvima (Lenvatinib) Thyroid cancer</p> 	<p>Aristada (Aripiprazole) Schizophrenia</p> 	

Quantitative *IVIVE* of Tissue Toxicity Supported by European Commission 7th FP Predict-IV Grant



1

Figure 1: Components of the four integrative modeling steps followed in Predict-IV.

Summary

- **In a systems pharmacology paradigm, the bottom-up approach to modeling and simulation of the ADME processes of a chemical, is a valuable tool in integrating available prior information and improving decision making.**
- **Improvement in the in vitro systems which can act as surrogates for in vivo reactions relevant to ADME**
- **Advances in the understanding of the extrapolation factors**
- **Advances in the development of mechanistic models of the human body**
- **Facilitate predicting PK characteristics in a wide range of healthy or disease populations accounting for age, sex, ethnicity, genetic, etc variability**
- **Moving towards PBPK coupling with systems biology models to predict toxicity endpoints/biomarkers and their associated variability from in vitro data**