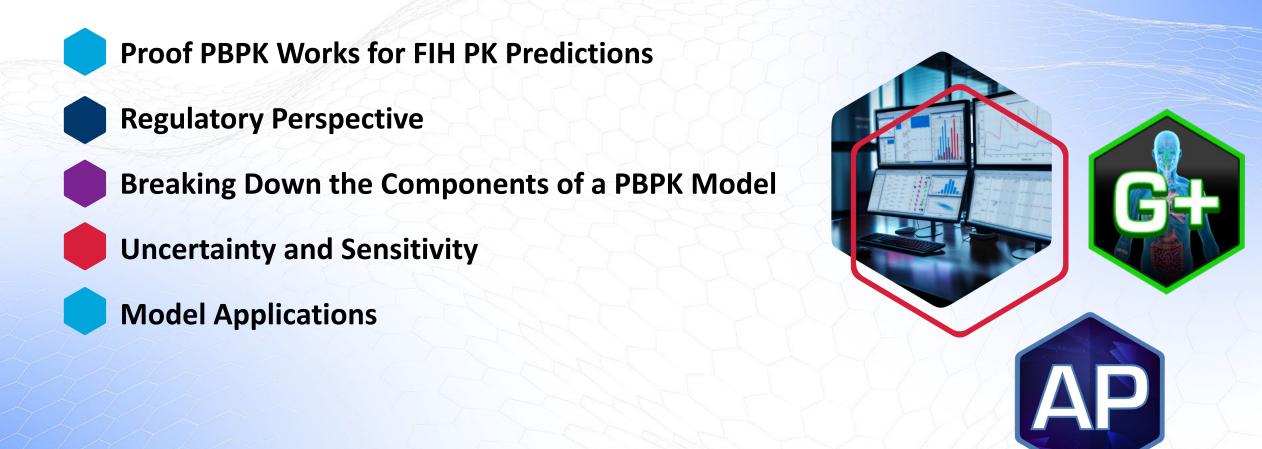


## St SimulationsPlus

Unlocking the Power of PBPK Modeling: PBPK for First-in-Human and Beyond March 20, 2024

#### Outline





#### **Considerations**

## Use everything that you know, but keep the model as simple as possible

- Aim to use PBPK not empirical modelling
- Combine compound specific data with physiology as they are intimately interconnected and explicable only by reference to the whole

Preclinical verification can increase the accuracy of the human prediction but will always lead to learning





#### Considerations

#### Always include uncertainty

Never give single point estimates

#### Be realistic with your expectations

 A successful prediction gets the category right or within 2-fold, but not necessarily matching values exactly

Sensitivity analysis and model application can inform future decisions

> Apply model to gain insight into potential development questions





#### **Proof PBPK Works for FIH PK Predictions**

ORIGINAL RESEARCH ARTICLE

#### A Novel Strategy for Physiologically Based Predictions of Human Pharmacokinetics

Hannah M. Jones,1 Neil Parrott,1 Karin Jorga2 and Thierry Lavé1

1 Drug Metabolism and Pharmacokinetics, F. Hoffmann-La Roche Ltd, Basel, Switzerland 2 Clinical Pharmacology, F. Hoffmann-La Roche Ltd, Basel, Switzerland "In the majority of cases, PBPK gave more accurate predictions of pharmacokinetic parameters and plasma concentration-time profiles than the Dedrick approach."

> DRUG MITAROLISM AND DISPOSITION Copyright © 2007 by The American Society for Pharmacology and Experimental Therapeut DMD 35:1766–1780, 2007

Vol. 35, No. 10 15644/3252832 Printed in U.S.A

Prediction of Human Pharmacokinetics Using Physiologically Based Modeling: A Retrospective Analysis of 26 Clinically Tested Drugs

Stefan S. De Buck, Vikash K. Sinha, Luca A. Fenu, Marjoleen J. Nijsen, Claire E. Mackie, and Ron A. H. J. Gilissen

Johnson & Johnson Pharmaceutical Research and Development, Discovery ADME-Tox Department, Beerse, Belgium

Received March 5, 2007; accepted July 3, 2007

"This evaluation demonstrates that PBPK models can lead to reasonable predictions of human pharmacokinetics."



#### **Proof PBPK Works for FIH PK Predictions**

ORIGINAL RESEARCH ARTICLE

03 12-5963/11/0005-0331/5-© 2011 Adls Data Information BV. All rights res

#### Simulation of Human Intravenous and Oral Pharmacokinetics of 21 Diverse Compounds Using Physiologically Based Pharmacokinetic Modelling

Hannah M. Jones,<sup>1</sup> Iain B. Gardner,<sup>1</sup> Wendy T. Collard,<sup>2</sup> Phil J. Stanley,<sup>3</sup> Penny Oxley,<sup>3</sup> Natilie A. Hosea,<sup>4</sup> David Plowchalk,<sup>5</sup> Steve Gernhardt,<sup>6</sup> Jing Lin,<sup>6</sup> Maurice Dickins,<sup>1</sup> S. Ravi Rahavendran,<sup>4</sup> Barry C. Jones,<sup>1</sup> Kenny J. Watson,<sup>1</sup> Henry Pertinez,<sup>1</sup> Vikas Kumar<sup>5</sup> and Susan Cole<sup>1</sup>

- 1 Department of Pharmacokinetics, Dynamics and Metabolism, Pfizer Worldwide R&D, Sandwich, UK
- 2 Department of Metabolism and Safety, Pfizer Animal Health, Kalamazoo, Michigan, USA
- 3 Department of Research Statistics, Pfizer Worldwide R&D, Sandwich, UK
- 4 Department of Pharmacokinetics, Dynamics and Metabolism, Pfizer Worldwide R&D, La Jolla, California, USA
- 5 Department of Clinical Pharmacology, Pfizer Worldwide R&D, Groton, Connecticut, USA
- 6 Department of Pharmacokinetics, Dynamics and Metabolism, Pfizer Worldwide R&D, Groton, Connecticut, USA

atriana, et Esseach Sutistics, Prize Worldwide R&D, Sundwich, UK artment of Pharmacolimetics, Dynamics and Metabolism, Pitzer Worldwide R&D, La Jolla, California, USA artment of Clinical Pharmacology, Pfizer Worldwide R&D, Groton, Connecticut, USA

> "Our prospective human PK prediction methods yielded good prediction results."

"The simulation results using PBPK were shown to be superior to those obtained via traditional one compartment analyses. In many cases, this difference was statistically significant."

RESEARCH ARTICLE – Drug Discovery-Development Interface

#### Prospective Predictions of Human Pharmacokinetics for Eighteen Compounds

#### TAO ZHANG, TYCHO HEIMBACH, WEN LIN, JIN ZHANG, HANDAN HE

Drug Metabolism and Pharmacokinetics, Novartis Institutes for Biomedical Research, East Hanover, New Jersey 07936

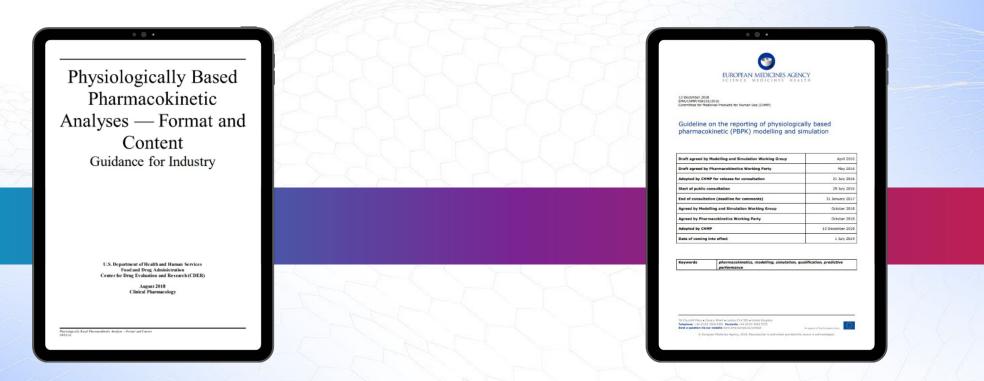
Received 26 August 2014; revised 2 January 2015; accepted 8 January 2015

Published online 17 February 2015 in Wiley Online Library (wileyonlinelibrary.com). DOI 10.1002/jps.24373

Received: 20 August 2014; revises 2 January 2015, accepted 8 January 2015 Published online 17 February 2015 in Wiley Online Library (wileyonlinelibrary.com). DOI 10.1002/jps.24373



#### **Regulatory Guidance for PBPK Modelling**



"Primarily for drug developers, FIH prediction using PBPK is important for decision-making and allows additional learning of the molecule and coping with situations when other methods may not be adequate." "PBPK modelling is a state-of-the-art modelling tool for assessing an appropriate starting dose for healthy volunteers.[...] The methods used and calculations on how doses and estimated exposure levels are determined, including methods for modelling such as PBPK modelling, should be included in the protocol and may be summarised in the investigator's brochure"

#### **An Industry Defined FIH PBPK Strategy**

Clinical Pharmacokinetics (2019) 58:727–746 https://doi.org/10.1007/s40262-019-00741-9

Physiologically Based Pharmacokinetic Modelling for First-In-Human Predictions: An Updated Model Building Strategy Illustrated with Challenging Industry Case Studies

• • •

Neil A. Miller<sup>1</sup> · Micaela B. Reddy<sup>2</sup> · Aki T. Heikkinen<sup>3</sup> · Viera Lukacova<sup>4</sup> · Neil Parrott<sup>5</sup>

Published online: 7 February 2019 © The Author(s) 2019

#### Abstract

Physiologically based pharmacokinetic modelling is well established in the pharmaceutical industry and is accepted by regulatory agencies for the prediction of drug-drug interactions. However, physiologically based pharmacokinetic modelling is valuable to address a much wider range of pharmaceutical applications, and new regulatory impact is expected as its full power is leveraged. As one example, physiologically based pharmacokinetic modelling is already routinely used during drug discovery for in-vitro to rive tor tastiant and pharmacokinetic modelling in preclinical species, and this leads to the application of verified models for first-in-human pharmacokinetic predictions. A consistent cross-industry strategy in this application area would increase confidence in the approach and facilitate further learning. With this in mind, this article aims to enhance a previously published first-in-human physiologically based pharmacokinetic modelling the strategy. Based on the experience of scientists from multiple companies participating in the GastroPlus<sup>TM</sup> User Group Steering Committee, new Absorption, Distribution, Metabolism and Exercion knowledge is integrated and decision trees proposed for each essential component of a first-in-human prediction. We have reviewed many relevant scientific publications to identify new findings and highlight key components of the strategy.

#### 1 Introduction

Physiologically based pharmacokinetic (PBPK) models represent the body as compartments parameterised based on physiology of tissues and organs including composition, volumes and blood flows [1]. Physiologically based pharmacokinetic models interpart this physiological description with compound-specific data to predict the pharmacokinetics of drugs, allowing simulation of the time course of drug concentrations in plasma and tissues.

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   Department of Clinical Pharmacology, Array BioPharma,
- Boulder, CO, USA <sup>3</sup> Admescope Ltd, Oulu, Finland
- Simulations Plus, Inc., Lancaster, CA, USA
   Pharmaceutical Sciences, Roche Pharma Research and Early
- Pharmaceutical Sciences, Roche Pharma Research and Early Development, Roche Innovation Centre Basel, Basel, Switzerland

ev Points

racy.

Linking of in-silic quantitative structure-property relationship models with physiologically based pharmacokinetic (PBPK) modelling is a powerful emerging technique, which is already being employed during early drug discovery. Combined with parameter sensitivity analyses, this can identify the compound properties most influencing systemic exposure and thus guide lead optimisation.

The quality of first-in-human PBPK predictions is greatly improved when measured inputs are available for the most critical parameters. PBPK model verification in preclinical species, which has not always been included in assessments of first-in-human pharmacokinetic predictions, is critical to build confidence and improve accu-

Uncertainty analysis is a key consideration to obtain maximal value from first-in-human PBPK predictions.

 $\triangle$  Adis

Check for

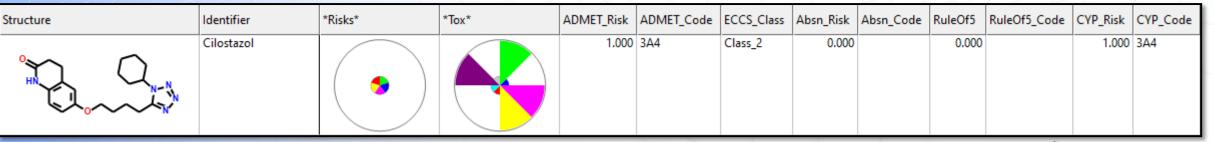
Flow diagrams for each essential component of a FIH prediction using PBPK modelling

### **Use Everything You Know**

FIH Simulator includes the major preclinical species to enable you to move seamlessly from mouse, rat, dog, minipig, rabbit, cyno through to human in one platform

- The overall risk profile can be assessed in ADMET Predictor
- Inputs without measured data can be predicted using ADMET Predictor
- NCA and compartmental analysis can be performed within the software
- Measured or predicted metabolism data for individual enzymes can be incorporated

#### which in turn allows an assessment of potential gut metabolism and phenotypic variability





#### Use Everything You Know: Start with QSPR + PBPK

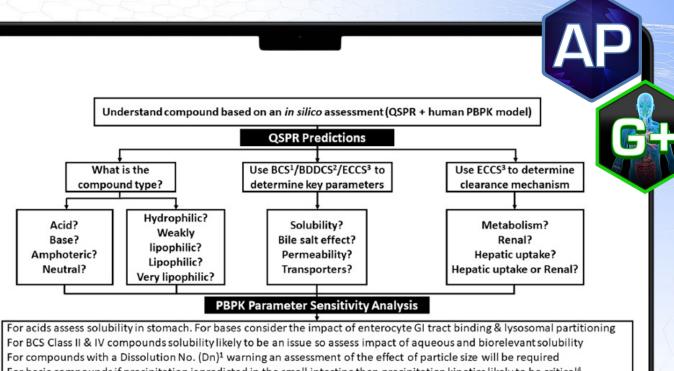


BDDCS – Biopharmaceutics Drug Disposition Classification System

ECCS – Extended Clearance Classification System

QSPR – Quantitative Structure-Property Relationship

BCS – Biopharmaceutic Classification System



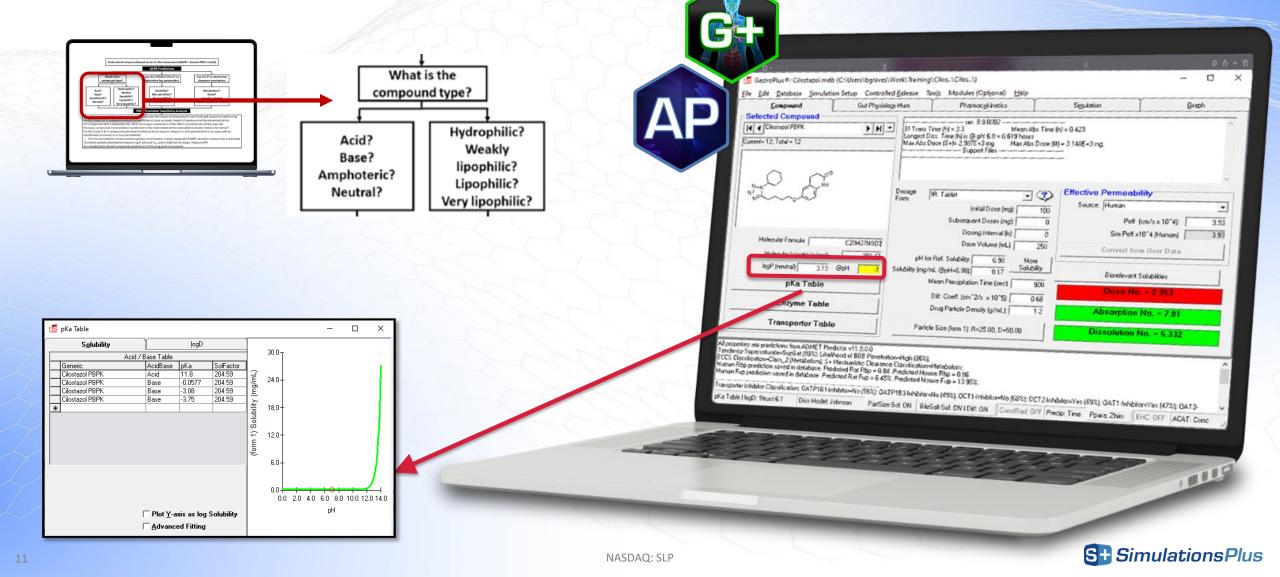
For basic compounds if precipitation is predicted in the small intestine then precipitation kinetics likely to be critical<sup>4</sup> For BCS Class III & IV compounds permeability likely to be an issue so measure *in vitro* permeability in an assay with an established conversion to *in vivo* permeability

 For low permeability compounds transporters could have an impact, especially if QSPR classifies compound as a substrate To predict systemic distribution measure log P, pKa and Fu<sub>p</sub>, and in addition, for bases, measure BPR For metabolically cleared compounds establish an IVIVE using preclinical species



NASDAQ: SLP

## Use Everything You Know: Compound Assessment Using QSPR + PBPK



## Use Everything You Know: Critical Thinking Using QSPR + PBPK

Edit Database Simulation Setu	up Controlled <u>R</u> elease Gut Physiology-Hum	Y	Circulation	Graph	determine key parameters		
<u>Compound</u> elected Compound  Cilostazol PBPK arrent= 12; Total = 12		ver. 9.9.0002	Si <u>m</u> ulation  ne (h) = 0.423 e (ik) = 3.148E+3 mg.	Graph	Solubility? Bile salt effect? Permeability? Transporters?	Metabolism? Renal? Hepatic uptake? Hepatic uptake or Re	
N-N N/N-N-O-CJ	C	Dosage Form: IR: Tablet	CFfective Permeab Source: Human	lity		een low and hi x 10^4 based o	gh permeability is on metoprolol
Molecular Formula: Molecular Weight (g/mol): logP (neutral): 3.13		Dosing Interval (h): Dose Volume (mL): pH for Ref. Solubility: 6.98 Solubility (mg/mL @pH=6.98): 0.17	250 Convert	eff x10^4 (Human) 3.93 from User Data	Bile Salt Effect ② ☑ Adjust solubility for bile s ☑ Adjust diff coeff for bile s	alt effect Use the biorel At least one o	Vitro Solubilities evant solubilities of form 1 f the FaSSIF, FeSSIF, or User solubilities must l alculate solubilization ratio
pKa Table		Mean Precipitation Time (sec): Diff. Coeff. (cm^2/s ×10^5):	900 Dose	No. = 2.353	· · ·		lues of biorelevant solubilites that are not availa re not used in SR calculation SGF FaSSIF FeSSIF FeSSIF V2
Enzyme Table		Drug Particle Density (g/mL):		tion No. = 7.81	Use theoretical solubili		pH: 1.2 6.5 5 5.8
Transporter Tab Classification=Class_2 (Metabolism); S- in Rbp prediction saved in database. Pr	+ Mechanistic Clearance Cl		Dissolut	ion No. = 5.332	Duodenal solubility at bile salt o 2.8mM will be 759.0 mg/mL	concentration Exp. Sol. (m	ıg/mL): 17.04 189.85 0 48.16



G

## Use Everything You Know: Pragmatic Assessment Using QSPR + PBPK

	Simulation	Graph				Cilostazol HLM	
arameters           PBPK         PK Model: HumAmeMalHithy30Y0_85.53kg_27.488MI           PBPK         Body Weight (kg):         65.53           Orat:         0         Liver:           Blood/plasma Conc Ratio:         0.7           C         Use Exp Plasma Fup [2]:         6.25           C         Use Adj Plasma Fup [2]:         6.25           Fisue         C         Use Adj Plasma Fup [2]:         6.25           Fisue         CL         CL int Fut/Fulnt           Hepatic Artey         0.000         0.000         0.000           Tissue         K         C         Use Adj Plasma Fup [2]:         6.25           Tissue         K          C		x (µg/mL) 0 TMax (h): 0 FPE g-h/mL): 0 ice (L/h): 0 M	(if fixed) [%] ral: 0 Intestinal: 0 Blood/ C Use Exp C Use Adj	■       ■         Body Weight (kg):       70         0       Liver:       12.67         /plasma Conc Ratio:       0.7         p Plasma Fup [%]:       6.25         i plasma Fup [%]:       6.25         i plasma Fup [%]:       6.25         or       (L/h/kg):       0         vc (L/h/kg):       0         Vc (L/kg):       0.89         T 1/2 (h):       5.38         K13 (1/h):       0         V3 (L/kg):       0	Concentration (ug/mL)		20 2

Multiple options to predict clearance: models for HLM, hepatocytes, enzyme kinetics



C

#### **Preclinical Verification to Better Predict Human PK** Distribution

Is the V<sub>ss.u</sub>

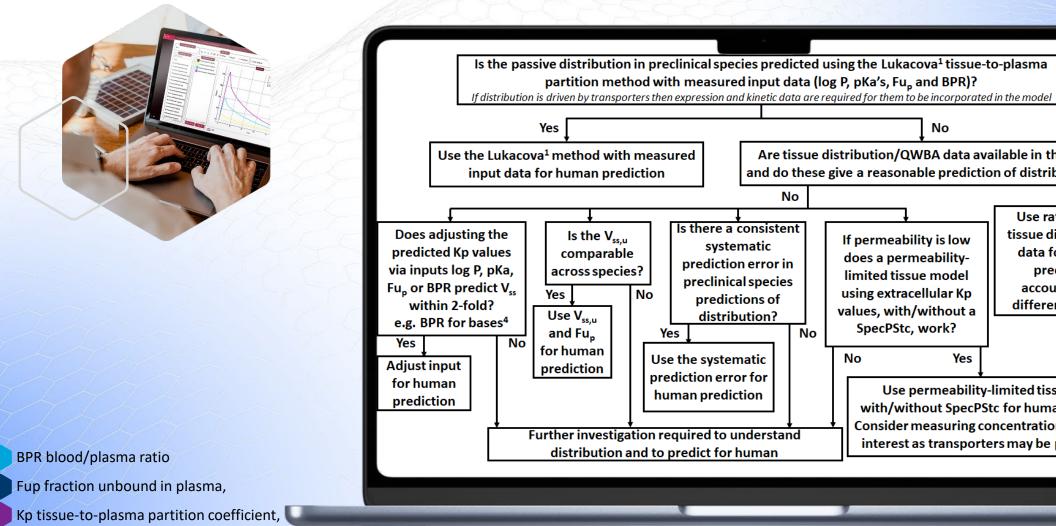
comparable

across species?

No

Yes

distribution and to predict for human



QWBA Quantitative Whole Body Autoradiography,

SpecPStc specific in-vivo diffusional clearance per millilitre of tissue cell volume.

NASDAQ: SLP



Yes

Use rat QWBA<sup>2</sup>/

tissue distribution<sup>3</sup>

data for human

prediction

accounting for

differences in Fu<sub>n</sub>

No

Are tissue distribution/QWBA data available in the rat and do these give a reasonable prediction of distribution?

If permeability is low

does a permeability-

limited tissue model

using extracellular Kp

values, with/without a

SpecPStc, work?

Yes

Use permeability-limited tissue model

with/without SpecPStc for human prediction. Consider measuring concentrations in tissues of

interest as transporters may be playing a role

No

No

No

Is there a consistent

systematic

prediction error in

preclinical species

predictions of

distribution?

Use the systematic

prediction error for

human prediction



## Preclinical Verification: Assessing the Pieces of the Jigsaw Puzzle FIH prediction for IV dosing of Tobramycin : Distribution



- Preclinical IV Cp vs Time profiles
- LogP/D
- 🌒 рКа
  - Fraction unbound in plasma (Fup)
  - Blood to plasma ratio (RBP)



Use the Lukacova Kp equation
 Input available measured data
 PBPK record per preclinical species
 PKPlus: calculate the NCA CL and V<sub>ss</sub>
 Liver CL = NCA CL - CL<sub>R</sub>



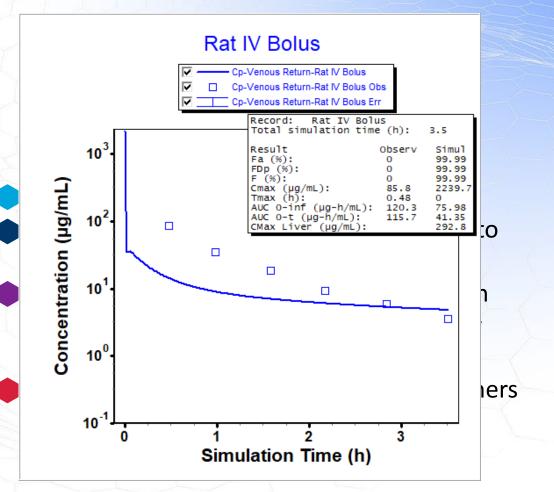


## **Preclinical Verification: Applying Critical Thinking**

FIH prediction for IV dosing of Tobramycin : Distribution

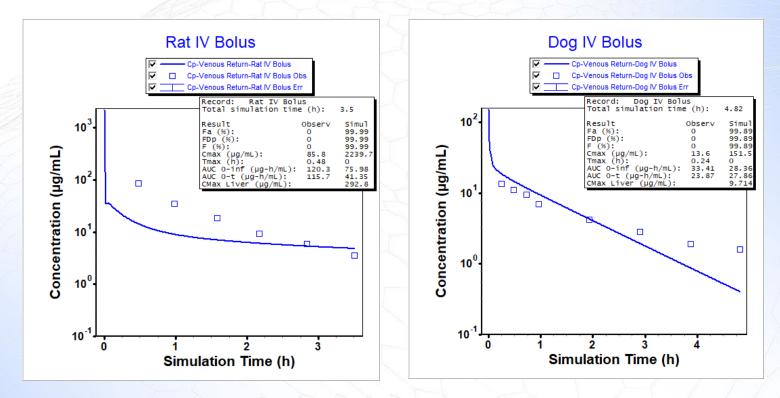
#### Assess Model

- Predicted Vss is too large causing a rapid distribution to tissues and a slow return to systemic circulation
- Knowing your compound can guide the modelling strategy
- Tobramycin is a low permeability compound so...



## Preclinical Verification: Applying Critical Thinking

#### FIH prediction for IV dosing of Tobramycin : Distribution

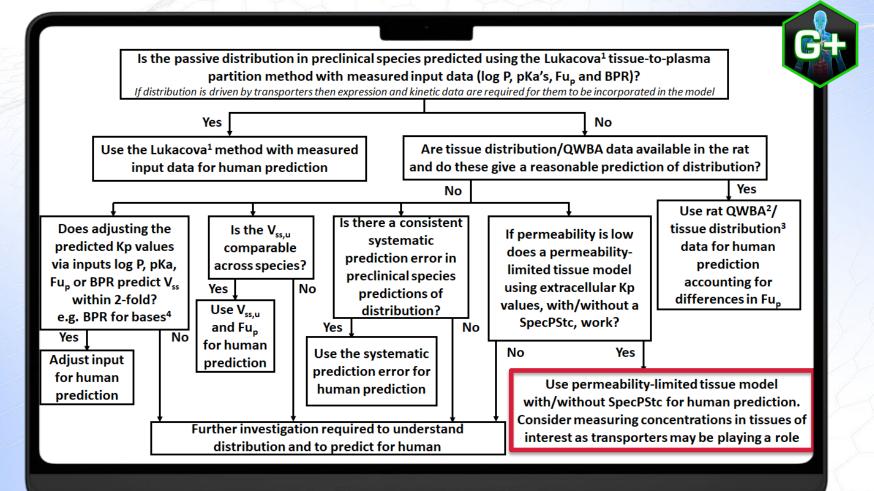


Species	Observed Vss	Predicted Vss	Pred/Obs
Rat	0.093 L	0.190 L	2.0
Dog	12.24 L	16.63 L	1.4



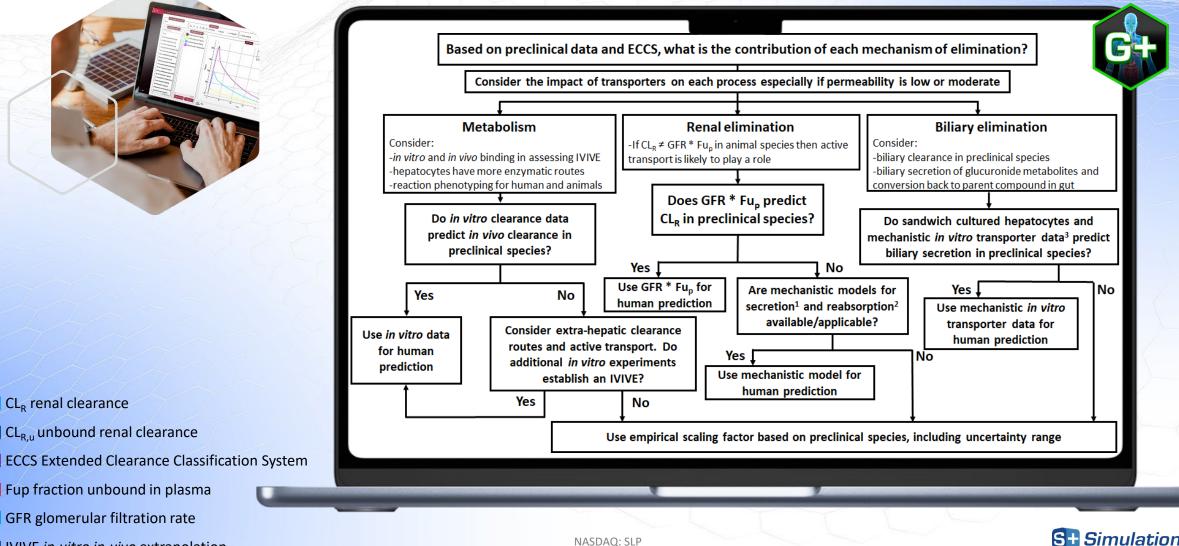
## Preclinical Verification Defines Strategy for Human Prediction

#### **Distribution - Tobramycin**



**S**+ SimulationsPlus

### **Preclinical Verification to Better Predict Human PK Metabolism and Elimination**



IVIVE in-vitro in-vivo extrapolation.

S+ SimulationsPlus

#### Preclinical Verification: Combining Measured Data with Predicted Properties Metabolism and Elimination

## Data

- Renal clearance from IV studies
   Observed or in vitro biliary clearance data
- Microsomal, hepatocyte or enzyme kinetic data

Tissue Parame	Tissue Parameters for: Kidney										
Basic	<u>⊳</u> d	vanced <u>E</u> nzyr	nes <u>I</u>	ransporters							
Name: Ki	idney	Volume (mL):	3.4336	fup*GFR	-						
Кр:	0.9341	Blood Flow (mL/s):	0.145	QUR (mL/s):	0.0004						
Fu Int:	1.4598	Lymph Flow (% PF):	0.4	GFR (mL/s):	0.0217						
Fu Ext:	0.9895	CLint (L/h):	0	fup:	0.9242						
		Renal CLsys (L/h):	0.0722	CLfilt (L/h):	0.0722						
		Basolateral:	Apical:	1	Set De <u>f</u> aults						
	PStc (mL	/s): 0.0116	0.0116								



- Validate CL<sub>R</sub> prediction Adjust CL<sub>R</sub> settings in the kidney tissue to account for active processes Parameterise and validate biliary
  - Parameterise and validate biliary clearance
- Use Metabolism and Transporter module to assess IVIVC

**S**+ SimulationsPlus

#### **Preclinical Verification:** Combining Measured Data with Predicted Properties **Metabolism and Elimination**

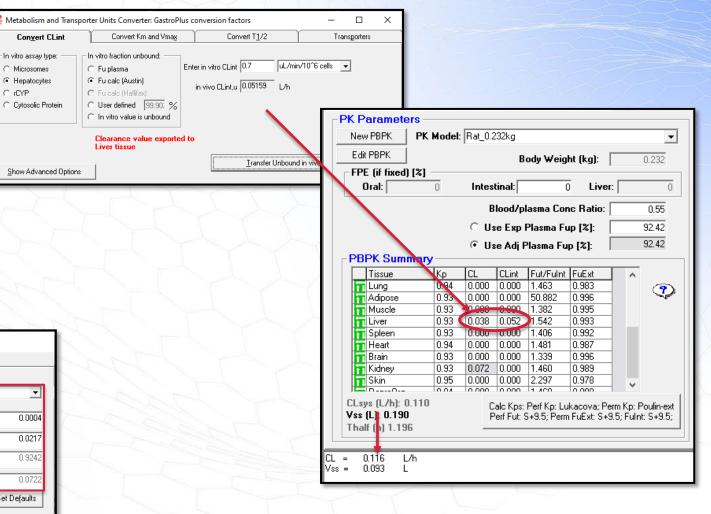
Microsomes

C rCYP

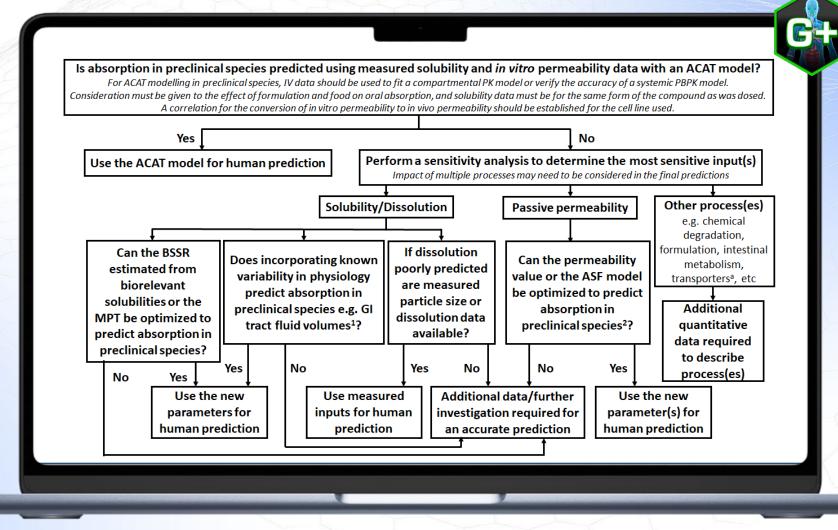
Data

- Renal clearance from IV studies Microsomal, hepatocyte or enzyme kinetic data
- Observed biliary clearance data

Tissue Parame	Tissue Parameters for: Kidney									
<u>B</u> asic	Basic Advanced		<u>E</u> nzym	nes	Īĸ	ansporters				
Name: K	idney	Volume	e (mL):	3.4336	5	up*GFR	_			
Kp:	0.9341	Blood Flow (r	nL/s):	0.145	5	QUR (mL/s):	0.0004			
Fu Int:	1.4598	Lymph Flow (% PF):		0.4	ī.	GFR (mL/s):	0.0217			
Fu Ext	0.9895	CLint (L/h):		(	ĩ	fup:	0.9242			
	, Renal CLs			0.0722	2	CLfilt (L/h):	0.0722			
Basolati			eral:	Apical:			Set De <u>f</u> aults			
	PStc (mL	/s): 0	.0116	0.0	116					



### **Preclinical Verification to Better Predict Human PK Oral Absorption**



ASF absorption scale factors BSSR bile salt solubilisation ratio MPT mean precipitation time

Efflux transporters can be incorporated in GastroPlus models with a simple method (e.g. adjusting permeability based on preclinical observations or in-vitro data) to more complex S+ SimulationsPlus

methods (e.g. specifically incorporating effects of transporters)

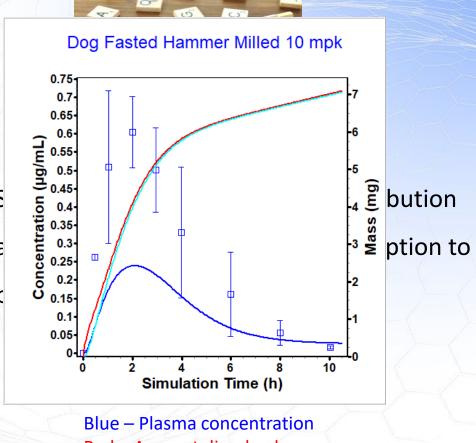
NASDAO: SLP



#### **Preclinical Verification: Building on the IV Model** FIH prediction for PO dosing of Cilostazol : Oral Absorption

Data

- Moderately lipophilic
- Essentially neutral at physiological pH
- Low to moderate solubility
- Reasonable permeability
- Low bile salt effect
- Dog PO PK



Red – Amount dissolved Cyan – Amount absorbed

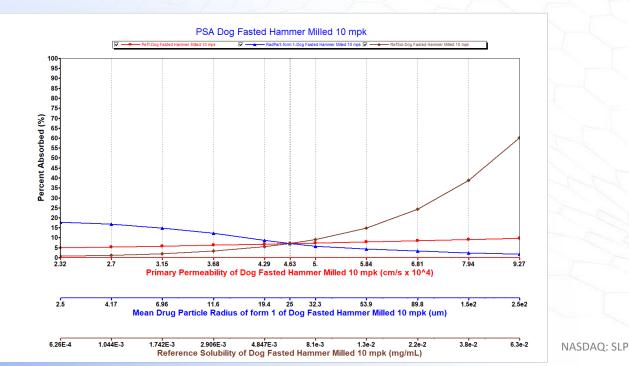


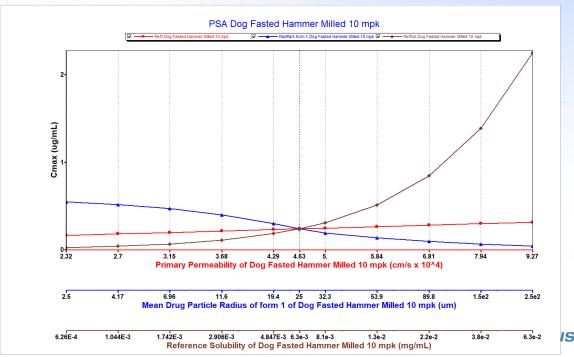


# **Preclinical Verification : Assessing the sensitive parameters**

#### FIH prediction for PO dosing of Cilostazol : Oral Absorption

	Parameter	Lower Bound	Baseline Value	Upper Bound	Number of Test	Spacing of Param Values
▲	Primary Permeability of Dog Fasted Hammer Milled 10 mpk (cm/s x 10^4)	2.316	4.633	9.266	10	Logarithmic
	Mean Drug Particle Radius of form 1 of Dog Fasted Hammer Milled 10 mpk (um)	2.5	25	250	10	Logarithmic
	Reference Solubility of Dog Fasted Hammer Milled 10 mpk (mg/mL)	0.000626	0.00626	0.063	10	Logarithmic





4



## Preclinical Verification : Additional measured data to enable prediction

#### FIH prediction for PO dosing of Cilostazol : Oral Absorption



- Dissolution is the rate limiting step for absorption
- Solubility has been fitted to measured pKas
- Particle size data is available



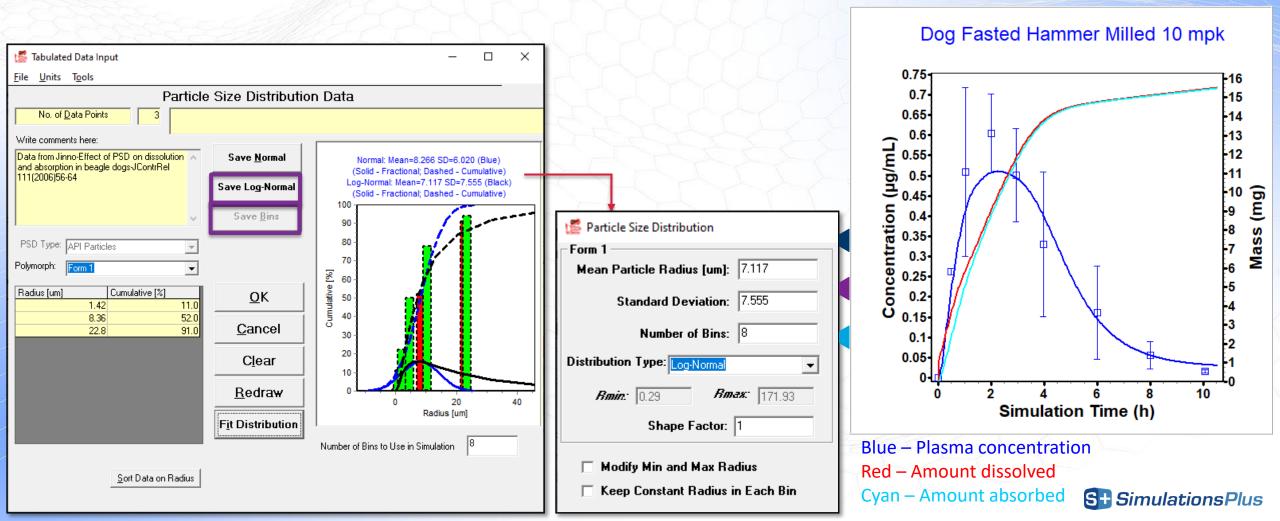
- Fit distribution to particle size data
- Apply to model
- Verify with other doses / species /

formulations



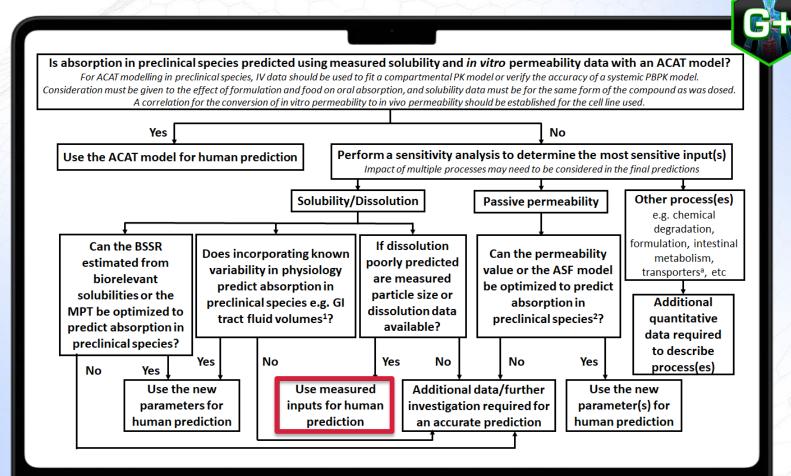
## Preclinical Verification : Additional measured data to enable prediction

#### FIH prediction for PO dosing of Cilostazol : Oral Absorption



## Preclinical Verification Defines Strategy for Human Prediction

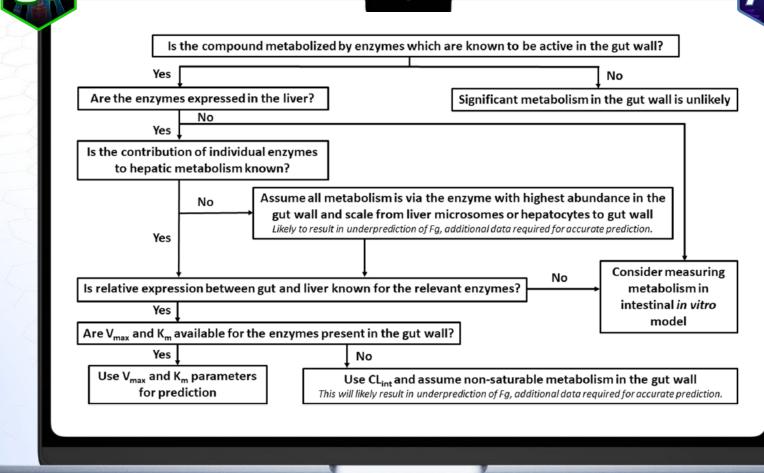
**Oral Absorption - Cilostazol** 



St SimulationsPlus

## Gt

## Preclinical Verification to Better Predict Human PK



CL<sub>int</sub> hepatic intrinsic clearance

Fg fraction of drug escaping gut wall metabolism

K<sub>m</sub> concentration of substrate at half V<sub>max</sub>

V<sub>max</sub> maximum velocity or rate of enzyme catalyzed reaction

Gut wall metabolism is often saturable, and thus if  $V_{max}$  and  $K_m$  parameters are available, evaluate saturation relative to dose





## **Preclinical Verification: Gap Analysis**



FIH prediction for PO dosing of Cilostazol : Gut Wall Metabolism



- Metabolism data
- Prediction of the human CL
- Isoforms involved in metabolism
- (from AP or in vitro data)



- CYP3A4 is predicted to metabolise Cilostazol, which is also present in the gut
- Incorporate learnings from preclinical PO modelling
- Simulate human profile using rCYP predicted clearance

Enzyme Table								
	Generic	Enzyme	Location		Vmax (mg/s) or (mg/s/mg·enz)	Km (mg/L)		
	Cilostazol	3A4	PBPK	Microsomes	0.0125	25.07		
	Cilostazol	3A4	Gut	Microsomes	5.53	25.07		
	Cilostazol	3A4	Liver	Microsomes	5.53	25.07		



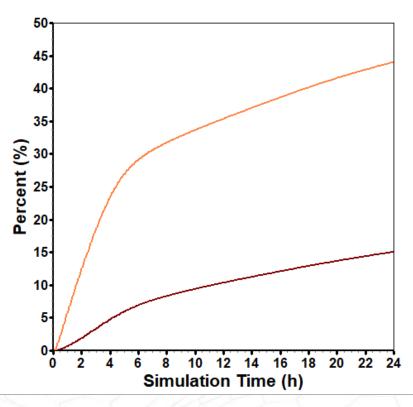
## Preclinical Verification: Extrapolation to Human



FIH prediction for PO dosing of Cilostazol : Gut Wall Metabolism



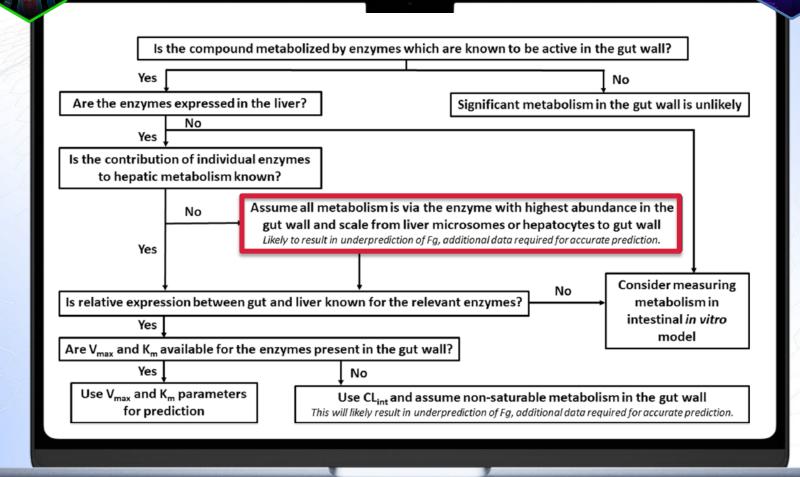
- Predicted human CLh is approximately 40% of liver blood flow
- Simulated profile using the predicted enzyme kinetics results in a Fh of ~59%
- Fg in human is predicted to be 35% indicating that gut metabolism could be a sensitive parameter



Human gut metabolism

Bhoeven Plassmætadnæksetdabio GYP3A4 in the liver RedngAm&untediasolised by CYP3A4 Cyan – Amount absorbed Purple – Amount reaching the portal vein S+ SimulationsPlus

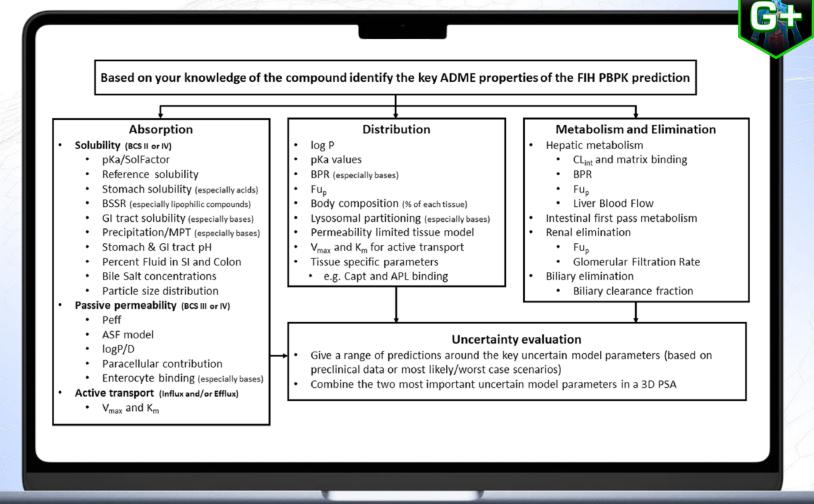
#### Preclinical Verification Defines Strategy for Human Prediction Gut Wall Metabolism





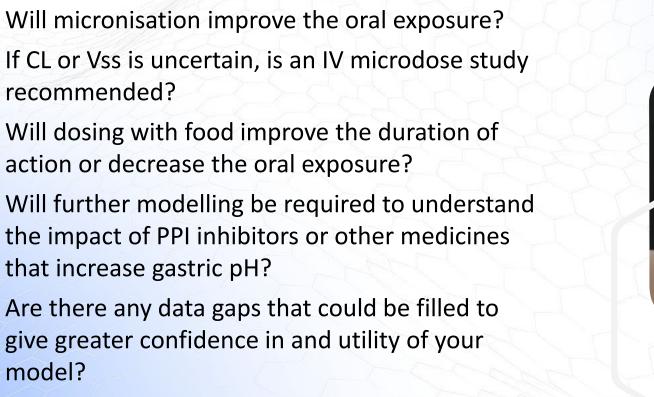
AP

#### **Always Include Uncertainty**





## Understanding the Sensitive Parameters: Planning for the Future



 For example, reaction phenotyping of metabolising enzymes, transporter kinetics





Explore

### **Model Application Beyond FIH: Additional Learning**

- Modelling high dose PK data in preclinical species to enable the prediction of Toxicology dose levels and frequency
  - To aid in the design of these pivotal supporting studies
- Predicting the PK in the target population in addition to healthy volunteers can inform the clinical plan
  - Populations available in the FIH Simulator include male and female, American and Asian, infant and paediatric, healthy, obese, hepatically and renally impaired





Explore

Learn

#### **Model Application Beyond FIH: Inform Your Clinical Plan**



- Models developed using the FIH Simulator can be used to:
  - predict the possibility of DDIs with your test compound as the victim or the perpetrator
  - Consider additional dosage routes
- This does require the purchase of additional modules





Explore

### **Regulatory Applications**

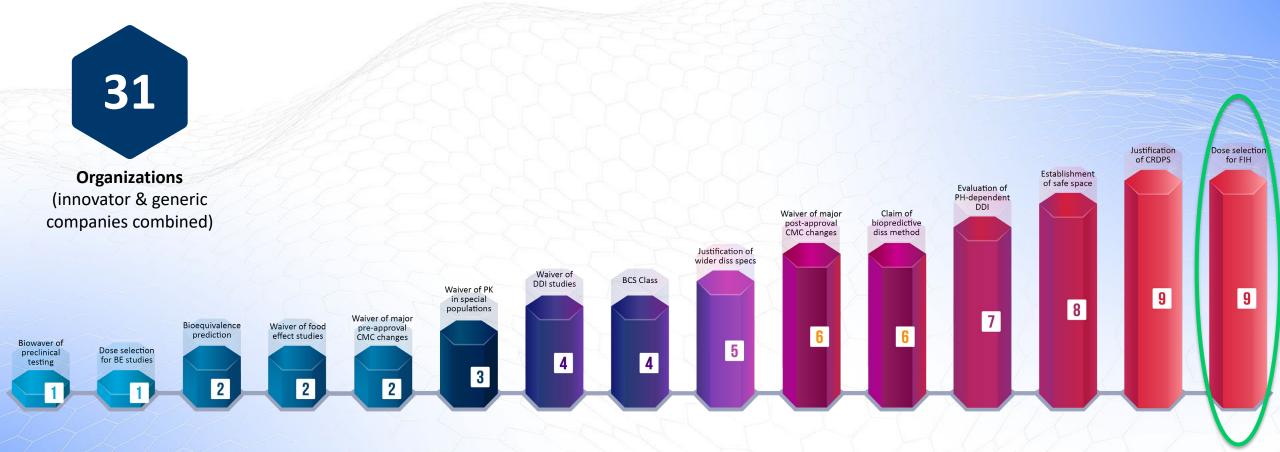


Fig. 1 Regulatory Applications of GastroPlus<sup>®</sup> (PBBM/PBPK Modeling) in the Pharmaceutical Industry (Years 2020-2022): Results from Survey



### Considerations

Use everything that you know, but keep the model as simple as possible

- Aim to use PBPK not empirical modelling
- Combine compound specific data with physiology as they are intimately interconnected and explicable only by reference to the whole

Preclinical verification can increase the accuracy of the human prediction but will always lead to learning

#### Always include uncertainty

Never give single point estimates

#### Be realistic with your expectations

 A successful prediction gets the category right or within 2-fold, but not necessarily matching values exactly

#### Sensitivity analysis and model application can inform future decisions

Apply model to gain insight into potential development questions





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# Thank You

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