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 A SIMULATIONS PLUS COMPANY

Applying in silico-in vitro-in vivo extrapolation (IS-IVIVE) techniques to predict exposure and guide risk assessment

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Evolving relationship between *in silico* tools and R&D

- **Model “supported”** (first questions 20 years ago): Do you think modeling and simulation might help?
- **Model “based”** (current questions today): How can I maximize the value of modeling and simulation in my development program?
- **Model “driven”** (future questions): How do I change the R&D process to reflect the availability of *in silico* tools and techniques?

How FDA Plans to Help Consumers Capitalize on Advances in Science

Posted on [July 7, 2017](#) by [FDA Voice](#)



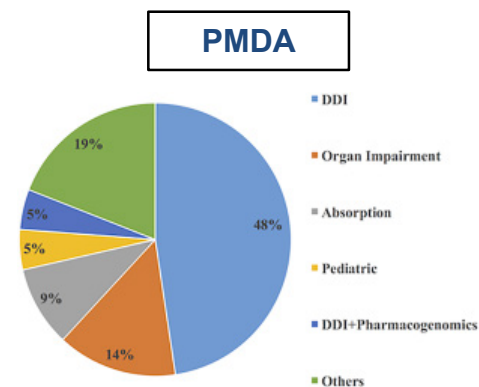
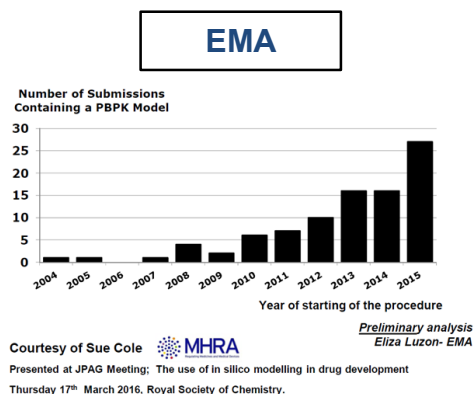
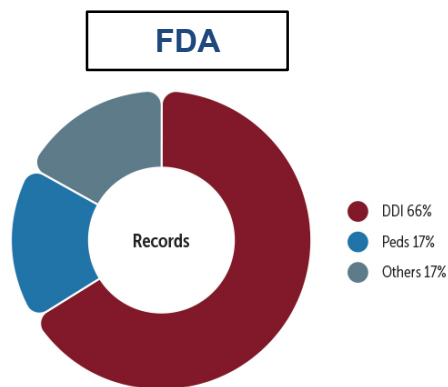
To build upon such opportunities, FDA will soon unveil a comprehensive Innovation Initiative. It will be aimed at making sure our regulatory processes are modern and efficient, so that safe and effective new technologies can reach patients in a timely fashion. We need to make sure that our regulatory principles are efficient and informed by the most up to date science. We don't want to present regulatory barriers to beneficial new medical innovations that add to the time, cost, and uncertainty of bringing these technologies forward if they don't add to our understanding of the product's safety and benefits.

Today we announced our detailed work plan for the steps we're taking to implement different aspects of Cures. I want to highlight one example of these steps, which we're investing in, and will be expanding on, as part of our broader Innovation Initiative. It's the use of in silico tools in clinical trials for improving drug development and making regulation more efficient.

FDA's Center for Drug Evaluation and Research (CDER) is currently using modeling and simulation to predict clinical outcomes, inform clinical trial designs, support evidence of effectiveness, optimize dosing, predict product safety, and evaluate potential adverse event mechanisms. We'll be putting out additional, updated guidance on how aspects of these in silico tools can be advanced and incorporated into different aspects of drug development.

Recent PBPK Modeling Trends: Regulatory Information

- 180 PBPK modeling citations in the FDA's Office of Clinical Pharmacology database (2008-15)
- 60 submissions received by EMA containing PBPK models (2013-15)
- 17 PBPK modeling citations at Japan PMDA (2014-16)



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SimulationsPlus | ***Cognigen*** | ***DILIsym Services***

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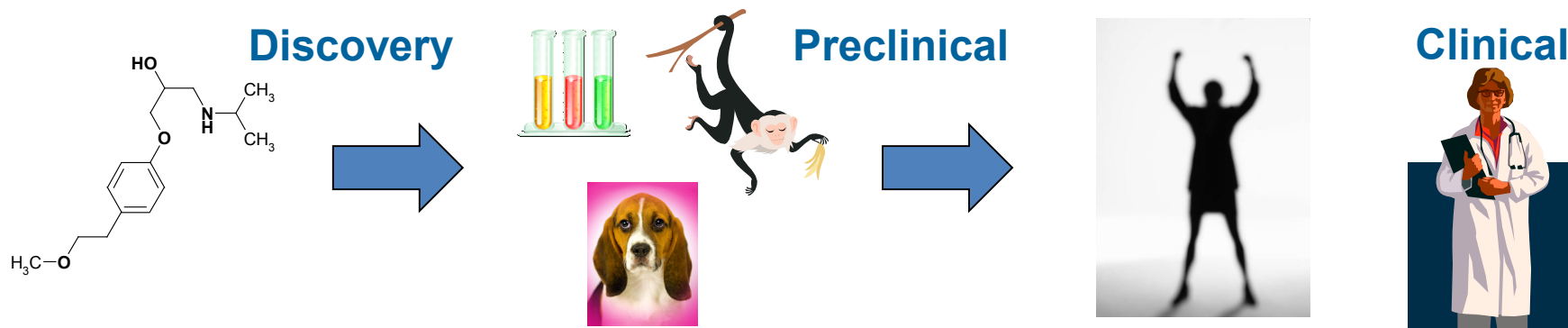
(NASDAQ: SLP); total employees ~100

- Simulations Plus, Inc.
 - Software development, PBPK modeling & simulation, and QSAR modeling
- Cognigen Corporation, a Simulations Plus company
 - Software development, pharmacometric services, and population PK/PD data analyses
- DILIsym Services, a Simulations Plus company
 - Software development, systems pharmacology/toxicology modeling

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Simulations Plus: Your end-to-end M&S solutions provider



ADMET Predictor™

GastroPlus™

**MedChem Studio™
MedChem Designer™**

**DDDPlus™
MembranePlus™**

PKPlus™

KIWI™

DILIsym®

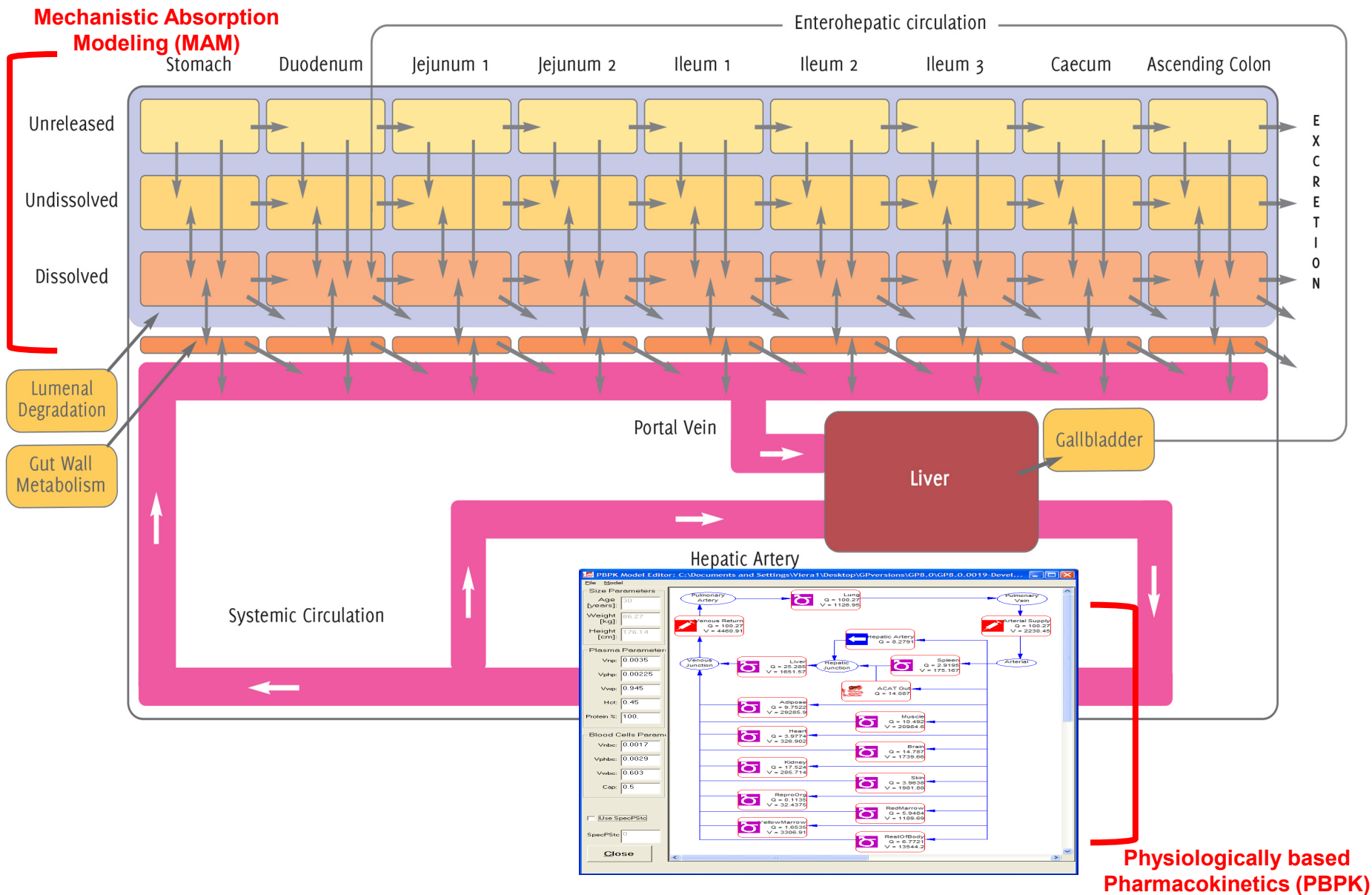
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Consulting Services and Collaborations

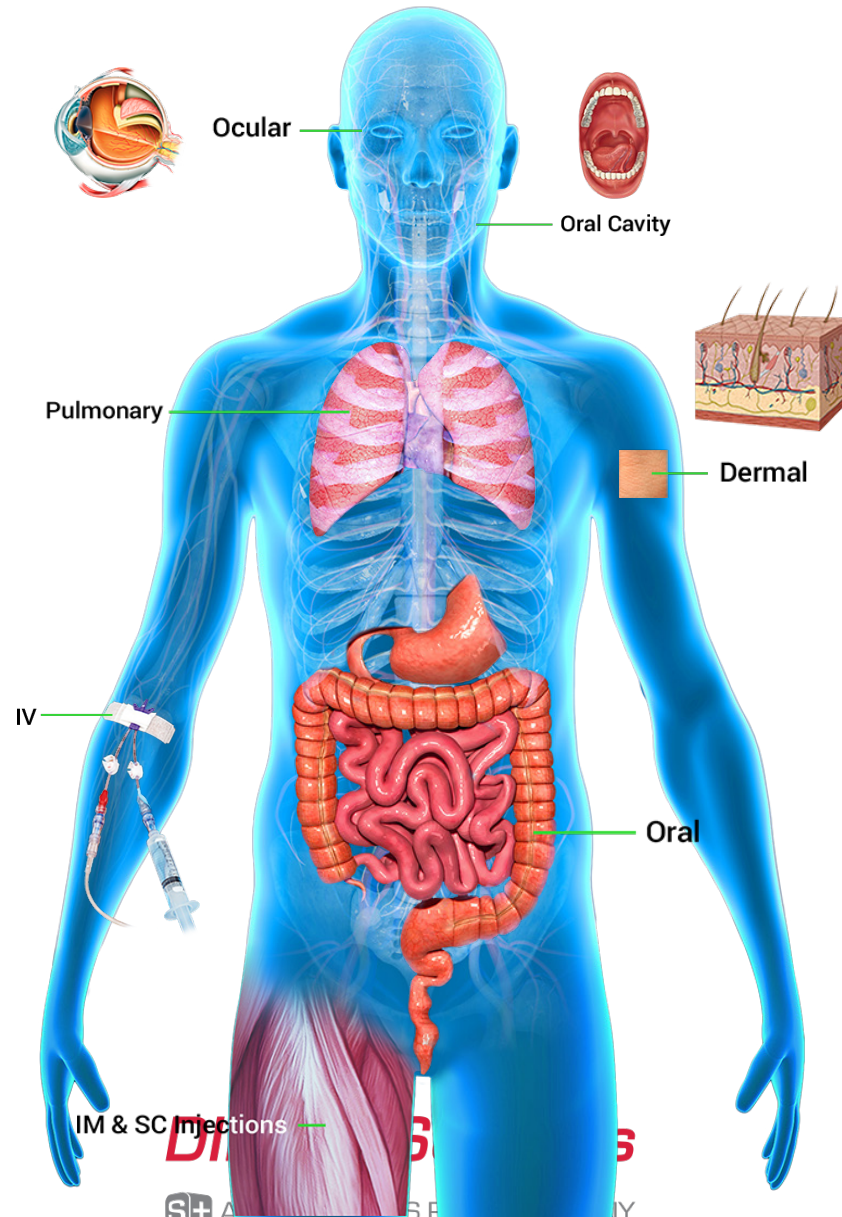
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Advanced Compartmental Absorption and Transit Model (ACAT™)

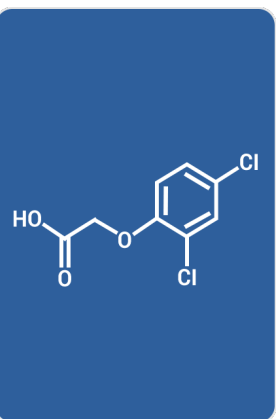


Mechanistic Absorption Models in GastroPlus™

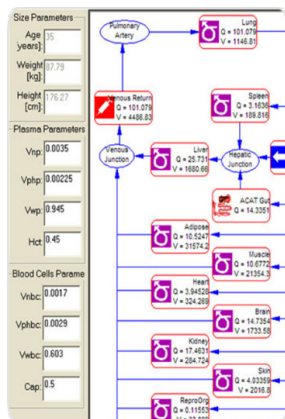


Note: all developed through funded collaborations with industry companies and/or the U.S. FDA

Saying "I do" to the QSAR/PBPK marriage...



Permeability, solubility vs. pH, pKa(s), logD vs. pH, Fup, blood:plasma ratio, tissue Kps, CLint, CLfilt

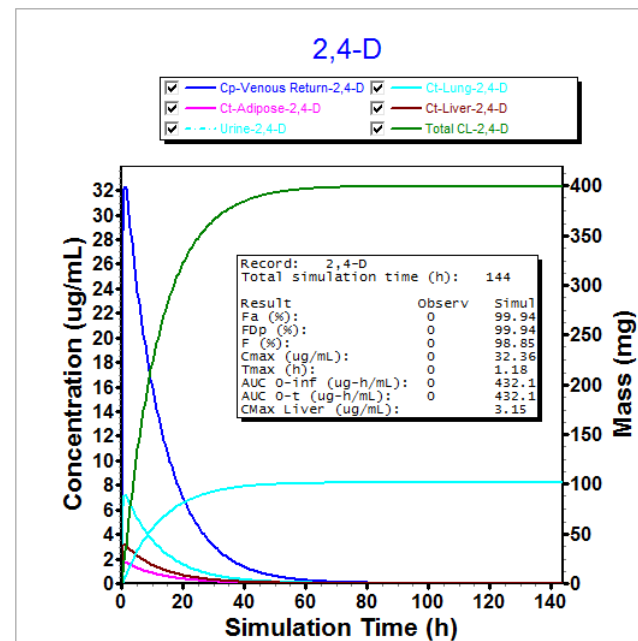


Quantitative Structure Activity Relationships (QSAR)

Physiologically-Based Pharmacokinetics (PBPK)

ADMET Predictor™

GastroPlus™



Goal: reliably and efficiently utilize PBPK modeling to reduce animal/human testing

Are the models any good?

Independent comparison of aqueous solubility predictors
(Dearden JC. Exper. Opin. Drug Discovery 2006 1:31)

Table 4. Predictive abilities of some commercially available software for aqueous solubility prediction, based on 122-compound test-set of drugs.

Software	% Compounds predicted within		r ²	q ²	s	Ref.
	± 0.5 log unit	± 1.0 log unit				
SimulationsPlus	64.8	91.0	0.82	0.82	0.47	[203]
Admensa	72.1	88.9	0.70	0.74	0.65	[205]
Pharma Algorithms ADME Boxes	59.0	86.9	0.74	0.73	0.62	[206]
ChemSilico	59.8	86.0	0.67	0.65	0.73	[202]
ACDLabs	59.0	85.2	0.73	0.72	0.66	[204]
AlogS	51.6	81.1	0.67	0.66	0.73	[207]
PredictionBase	46.7	81.1	0.48	0.46	1.07	[208]
ESOL	54.9	78.7	0.60	0.59	0.84	[209]
MOLPRO	62.3	77.9	0.44	0.42	1.22	[210]
Absolv 2	44.3	74.6	0.53	0.51	0.95	[206]
QikProp	47.6	73.8	0.57	0.57	0.97	[201]
SPARC*	42.9	73.1	0.73	0.72	0.96	[211]
Cerius ² ADME	37.7	72.9	0.61	0.60	1.02	[212]
WSKOWWIN	41.0	67.2	0.51	0.49	1.17	[213]
ADMEWORKS Predictor	34.4	66.4	0.42	0.39	1.24	[214]
AlogP98	38.5	62.3	0.42	0.40	0.77	[85,212]
CHEMICALC [†]	23.3	45.7	0.35	0.34	1.96	[215]

*Based on 119 compounds; SPARC could not calculate solubilities of 3 compounds.

[†]Based on 116 compounds, using log P method with calculated melting point, which was not available for 6 compounds; kindly calculated by Prof. G. Schüürmann.

Table 2. Performance of algorithms

Method	Star (234)		Nostar (50)		Zwitterions (18)		Other (266)
	MAE	Rank	MAE	Rank	MAE	AE	MAE
A _s -logP	0.33	I	0.7	I	0.4	-0.01	0.4
ALOGPS ³	0.39	I	0.7	I	0.64	-0.51	0.44
VLOGP ⁴	0.50(0.41)	II	0.95(0.84)	I,III	0.87(0.69)	-0.8(-0.62)	0.56(0.47)
SLIPPER	0.58	II	0.91	I,III	1.2	-1.14	0.6
QikProp	0.58	II	1.01	III	0.83	-0.48	0.64
CSlogP	0.61	II	0.95	I,III	0.54	-0.06	0.68
TLOGP ⁵	0.64	II	1.01	III	1.26	-0.97	0.69
Absolv	0.65	II	0.94	I,III	1.98	-1.97	0.61
QuantlogP ³	0.7	II	1.03	III	1.91	-1.9	0.68
QLOGP							
VEGA ⁶							
CLIP ⁷							
LISER							
MLOGP							
SPARC ^{8,9}	0.93	III	1.17	III	0.72	0.06	0.99
COSMOFrag ³	1.13	III	1.38	IV	2.48	-2.47	1.09
LISER UFZ ⁸	1.19	IV	2.15	IV	2.32	-1.75	1.29
GBLOGP ⁷	1.25	IV	1.76	IV	2.51	2.46	1.26
HINT	1.38	IV	2.14	IV	3.25	-3.24	1.39
AAM	1.37	IV	1.87	IV	2.96		1.36

Independent comparison of logP predictors
(Tetko & Poda, 2007)

Comparison of first-in-human (FIH) PBPK prediction accuracy in a 2-year study of 21 compounds
(Cole et al., ISSX 2008)

Summary of IV profile prediction accuracy

APPROACH	PROFILE	Vss		CL	
	Weighted sum of squares (RANK)	AFE	% within 2-fold error (3-fold error)	AFE	% within 2-fold error (3-fold error)
GastroPlus	-11.7 (1)	1.4	90 (100)	1.6	80 (85)
PKSim	-6.4 (2)	1.7	70 (90)	1.6	80 (85)
Current Pfizer Approach	-3.8 (3)	1.6	75 (85)	1.6	80 (85)
SimCYP - hlm	5.6 (4)*	1.5	80 (95)	2.5	58 (74)
SimCYP - rhCYP	7.8 (5)*	1.5	80 (95)	2.4	55 (65)
ChloePK	8.5 (6)*	-	-	1.7	70 (80)

Summary of Oral profile prediction accuracy

AFE → Average Fold Error

APPROACH	PROFILE	AUC		Cmax	
	Weighted sum of squares (RANK)	AFE	% within 2-fold error (3-fold error)	AFE	% within 2-fold error (3-fold error)
GastroPlus	-9.8 (1)	2.7	50 (72)	2.0	67 (72)
Current Pfizer Approach	-5.3 (2)	3.9	33 (56)	2.5	44 (61)
SimCYP - rhCYP	-3.7 (3)	3.0	56 (67)	2.2	61 (72)
SimCYP - hlm	5.7 (4)*	3.6	41 (53)	2.7	53 (59)
PKSim	6.1 (5)*	4.7	22 (39)	5.0	17 (33)
ChloePK	7.0 (6)*	2.8	39 (50)	2.4	50 (61)

Predicted by	Trained with	MAE	RMSE	R ²
ACD/Percepta v. 12	15932 lit pK _a	0.77	1.05	0.84
ADMET Predictor v. 6.1	14147 lit pK _a	0.73	0.95	0.86
ADMET Predictor v. 7.0	14149 lit pK _a + 19467 Bayer pK _a	0.51	0.67	0.93

Fraczkiewicz et al. (2015) J. Chem. Inf. Model. 55(2):389

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Recent Validation

Daga et al. (2018) Mol. Pharm.
 >75% of compounds predicted within 2-fold

Lawless et al. (2016) ASCPT Annual Meeting
 Using QSAR and PBPK to predict human F%:
 70% of compounds predicted within 2-fold

Physiologically Based Pharmacokinetic Modeling in Lead Optimization. 1. Evaluation and Adaptation of GastroPlus To Predict Bioavailability of Medchem Series

Pankaj R. Daga,^{†*} Michael B. Bolger,[§] Ian S. Haworth,^{||} Robert D. Clark,[§] and Eric J. Martin^{*†}

[†]Novartis Institute of Biomedical Research, Emeryville, California 94608, United States

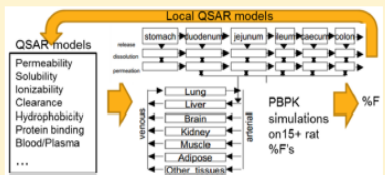
[§]Simulations Plus, Inc., 42505 10th Street West, Lancaster, California 93534, United States

^{||}Department of Pharmacology and Pharmaceutical Sciences, University of Southern California, Los Angeles, California 90089, United States

Supporting Information

ABSTRACT: When medicinal chemists need to improve bioavailability (%F) within a chemical series during lead optimization, they synthesize new series members with systematically modified properties mainly by following experience and general rules of thumb. More quantitative models that predict %F of proposed compounds from chemical structure alone have proven elusive. Global empirical %F quantitative structure–property (QSPR) models perform poorly, and projects have too little data to train local %F QSPR models. Mechanistic oral absorption and physiologically based pharmacokinetic (PBPK) models simulate the dissolution, absorption, systemic distribution, and clearance of a drug in preclinical species and humans. Attempts to build global PBPK models based purely on calculated inputs have not achieved the <2-fold average error needed to guide lead optimization. In this work, local GastroPlus PBPK models are instead customized for individual medchem series. The key innovation was building a local QSPR for a numerically fitted effective intrinsic clearance ($CL_{int,eff}$). All inputs are subsequently computed from structure alone, so the models can be applied in advance of synthesis. Training $CL_{int,eff}$ on the first 15–18 rat %F measurements gave adequate predictions, with clear improvements up to about 30 measurements, and incremental improvements beyond that.

KEYWORDS: PBPK, lead optimization, lead series, local model, intrinsic clearance



PREDICTION OF ORAL BIOAVAILABILITY *in silico*

Michael Lawless, John DiBella, Michael B. Bolger, Robert D. Clark, Eva Huehn, Marvin Waldman, Jinhua Zhang, and Viera Lukacova
 Simulations Plus, Inc. (www.simulations-plus.com)

Abstract	Results																		
<ul style="list-style-type: none"> A database of 42 drugs including oral bioavailability (F%) and dose was constructed All compounds' reported major clearance pathways (MCP) were CYP-mediated For 42 drugs with more than one reported value of F%, the average experimental F% was 20% Reported F% values varied from 3% (sphenacetin) to 99% (nitroglycerin, galantamine, glimepiride, indinavir, and tramadol), with an average of 46% F% was predicted by comparing quantitative structure activity relationship (QSAR) model predictions* and physiologically based pharmacokinetic (PBPK) simulations** A 15-year-old American male physiology was used for all PBPK simulations All molecules were predicted to be substrates of the CYP associated with their MCP In 42 of 42 molecules, the CYP substrate with highest predicted intrinsic clearance (CL_{int}) was the same as the MCP Overall, 68% of the molecules were predicted within 2-fold of their reported F% Scaling V_{max} by the CYP substrate model's confidence estimate resulted in lower underprediction 	<table border="1"> <thead> <tr> <th>CYP substrate</th> <th>Observed F%</th> <th>Observed F%</th> <th>Observed F%</th> <th>Observed F%</th> <th>Observed F%</th> </tr> <tr> <th>(arbitrarily)</th> <th>1</th> <th>2</th> <th>3</th> <th>4</th> <th>5</th> </tr> </thead> <tbody> <tr> <td></td> <td>12</td> <td>4.6</td> <td>3.1</td> <td>3.6</td> <td>3.6</td> </tr> </tbody> </table>	CYP substrate	Observed F%	Observed F%	Observed F%	Observed F%	Observed F%	(arbitrarily)	1	2	3	4	5		12	4.6	3.1	3.6	3.6
CYP substrate	Observed F%	Observed F%	Observed F%	Observed F%	Observed F%														
(arbitrarily)	1	2	3	4	5														
	12	4.6	3.1	3.6	3.6														

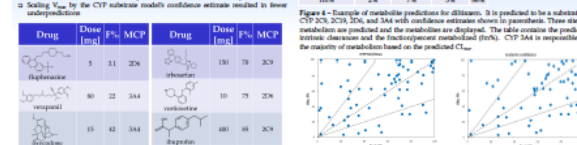


Figure 3 – Examples of drugs in the data set along with their dose, F%, and MCP.

Methodology
 Figure 2 = CYP inhibition models for 5 CYP isoforms (2C9, 2C19, 2C18, 2D6, and 3A4). The first model predicted whether a molecule is a substrate for each CYP isoform. These predictions included confidence estimates** (red, size of metabolites are predicted; blue, metabolites are not predicted; finally, kinetic parameters are predicted and metabolites are depicted).

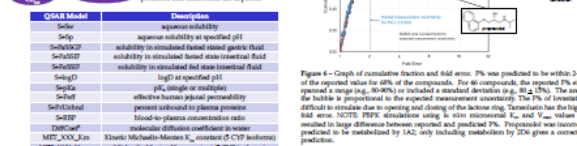


Figure 4 – Graph of cumulative fraction and AUC error. F% was predicted to be within 2-fold of the reported value for 48% of the compounds. For the 46 compounds, the reported F% either spanned a major log (e.g., 0.0010 to 1.0) or included a standard deviation (e.g., 0.1g, 10%). The size of the bubble is proportional to the reported measurement uncertainty. The F% of levamisole is difficult to compare due to spacing and dosing of the active drug. Tramadol has the highest AUC error. NOTE: PBPK simulations using in vitro measured K_m and V_{max} values also resulted in large differences between reported and predicted F%. Preparation was incorrectly predicted to be metabolized by 1A2, only including metabolites by 2D6 gave a correct F% prediction.

Conclusions
 A database of 42 drugs along with dosage and F% was compiled. Each compound's reported MCP was CYP-mediated. F%, F₁₂, and F₁₈ were extracted with PBPK simulations using physicochemical and CYP kinetic parameters predicted entirely from QSAR models. The CYP isoform associated with the MCP was correctly predicted in 42 of 42 of the CYP substrates. Additionally, 68% of the predicted oral bioavailability values were within 2-fold of the observed oral bioavailability. Scaling V_{max} by confidence estimates from our CYP substrate model reduced the number of underpredictions.

Table 1 – QSAR models used in PBPK simulations.

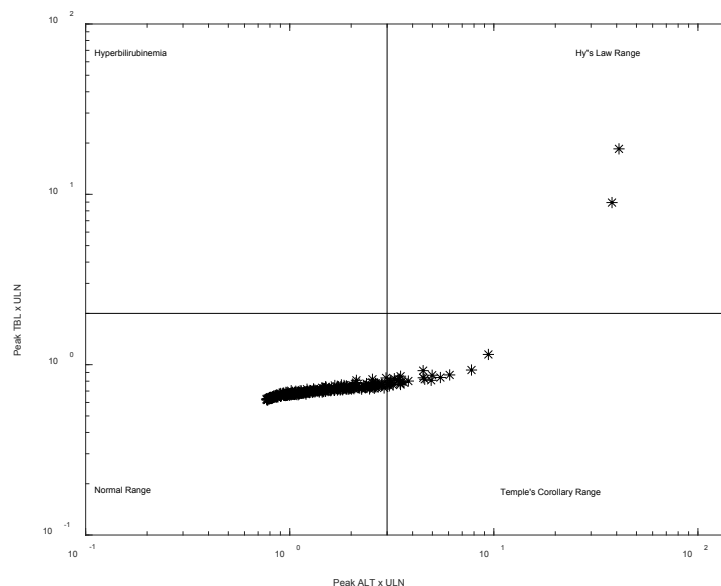
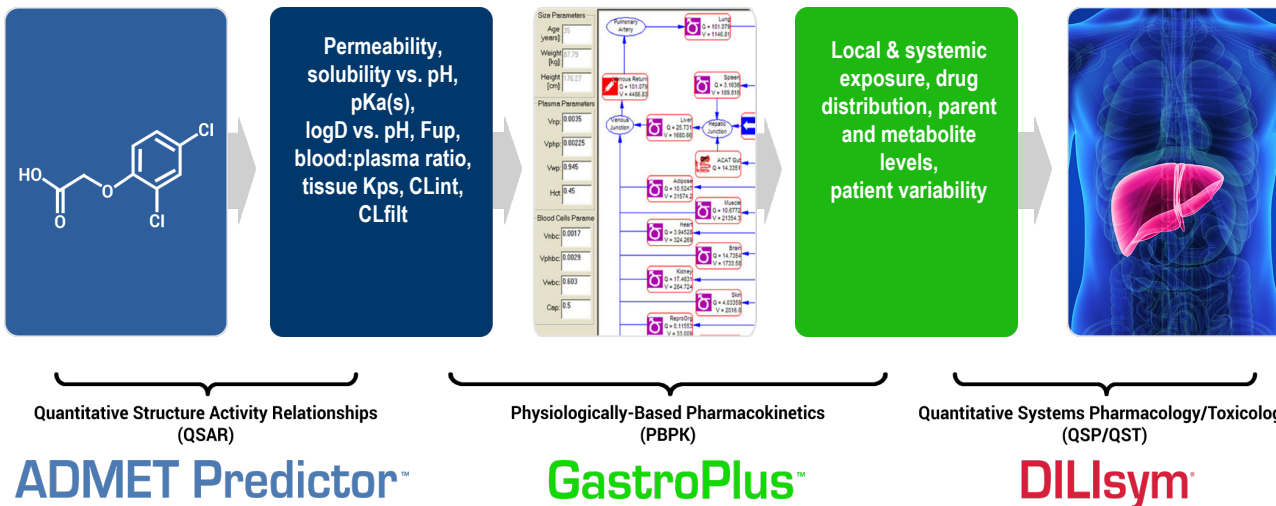
QSAR Model	Descriptor
Selfp	aqueous solubility at specified pH
SelfpH2CF	solubility in simulated human gastric fluid
SelfpH2CF	solubility in simulated human intestinal fluid
SelfpH2CF	solubility in simulated fed state intestinal fluid
SelfpD	logP at specified pH
SelfpA	logP (single or multiple)
SelfpH	effective human jejunal permeability
SelfpH2CF	permeability to plasma proteins
SelfpH2CF	blood-to-plasma concentration ratio
DiffCoeff	molecular diffusion coefficient in water
MET_XXX_XXX	kinetic Michaelis-Menten K_m constant of CYP isoform X
MET_XXX_Vmax	kinetic Michaelis-Menten V_{max} constant of CYP isoform X

References
 *Tobinico E et al, Drug Model. Clin. Post Forward. Published on August 14, 2014.
 **Thurston ET et al., In: Brunton LL, Chabner BA, Knollman BC, editors. Goodman & Gilman's pharmacological basis of therapeutics. 12th ed. New York: McGraw-Hill; 2011. ADAMPT Predictions version 7.2. Simulations Plus, Inc., Lancaster, CA 93534 USA.
 ***CognisPlus version 10. Simulations Plus, Inc., Lancaster, CA 93534 USA.
 †Clark RD et al., J. Chromatogr B 2014, 624.
 †Huehn E and Lander JL, American Institute of Chemical Engineers J. 2014, 2011.
 †Modified from van de Wallebeert H and Offord E. ADMET in Drug Modeling: Trends Prediction. Trends Pharm. New Drug Des. 2002, 2:182-200.



Figure 2 – Orally dosed drugs typically dissolve in the stomach and transit into the intestine, where they can be absorbed into the gut wall. F% (fraction absorbed) is the fraction of dose that is absorbed into the apical membrane of the gut epithelium. CYP enzymes metabolize some compounds in the intestine. F₁₂ and F₁₈ are the fraction (percent) of dose that makes it to the portal vein. F% is the fraction (percent) of dose that enters systemic circulation. F₁₂, F₁₈, and F% were predicted by our GastroPlus PBPK simulations.

Saying "I do" to the QSAR/PBPK marriage...



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DILIsym Services, Inc. – Our Vision

“Our vision is safer, effective, more affordable medicines for patients through modeling and simulation.”

- DILIsym Services, Inc. offers comprehensive program services:
 - **DILIsym** software licensing, training, development (DILI-sim Initiative)
 - **DILIsym** and **NAFLDsym** simulation consulting projects
 - Consulting and data interpretation
 - *in vitro* assay experimental design and management

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The DILI-sim Initiative is a Partnership between DILIsym Services and Pharmaceutical Companies to Minimize DILI

- Overall Goals

- Improve patient safety through QST
- Reduce the need for animal testing
- Reduce the costs and time necessary to develop new drugs

- History

- Officially started in 2011
- 19 major pharmaceutical companies have participated
- Members have provided compounds, data, and conducted experiments to support effort
- Over \$8 million total invested in project

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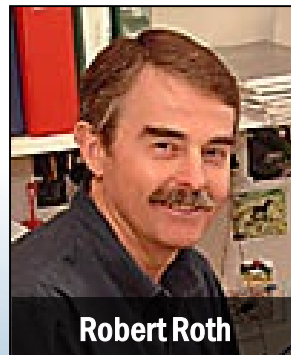
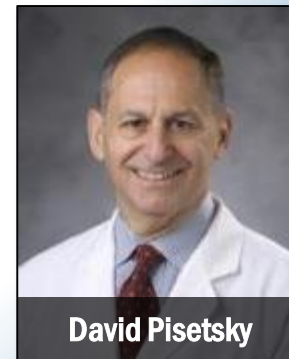
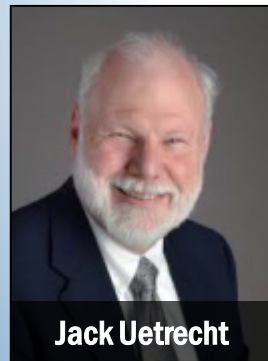
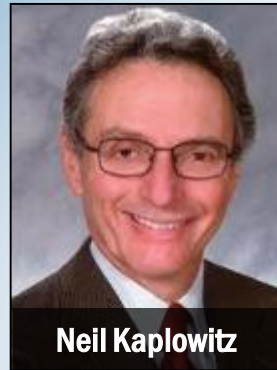
Sample of Some Current Consortium Members



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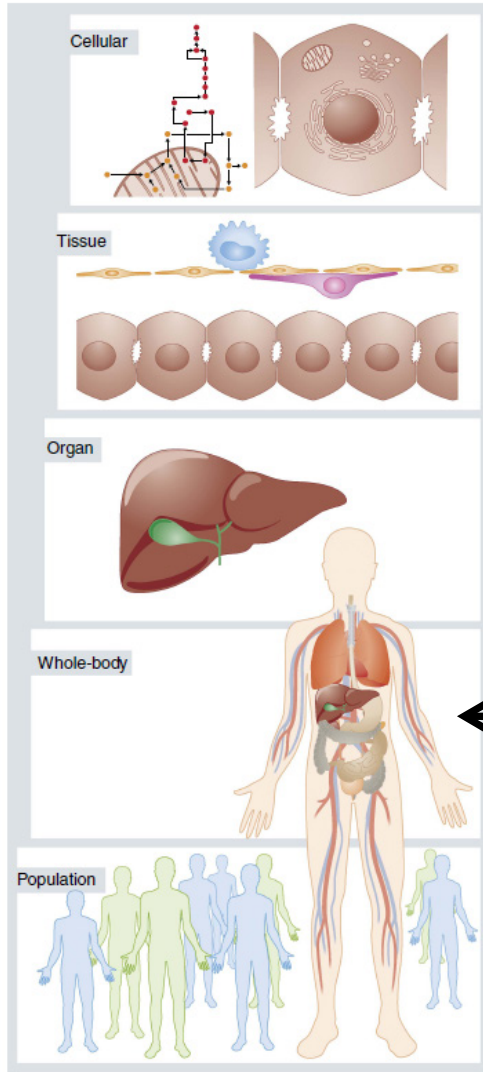
The DILI-sim Scientific Advisory Board Includes World Class Scientists



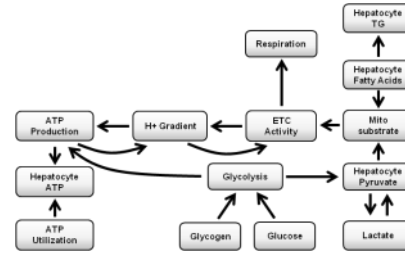
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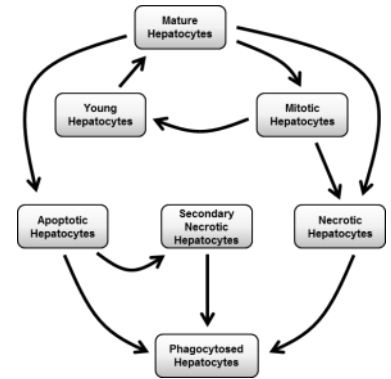
DILIsym: Quantitative Systems Toxicology



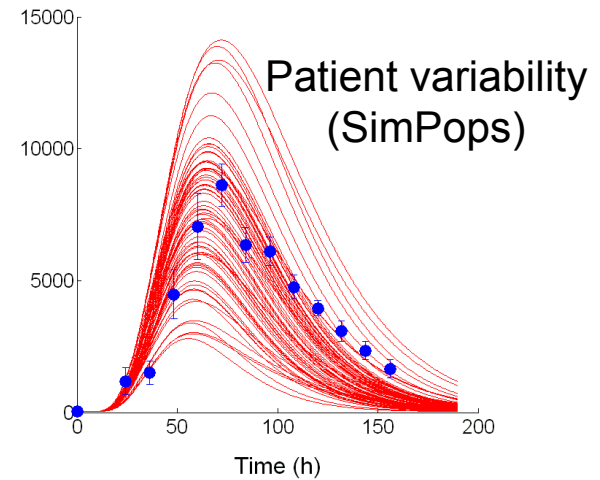
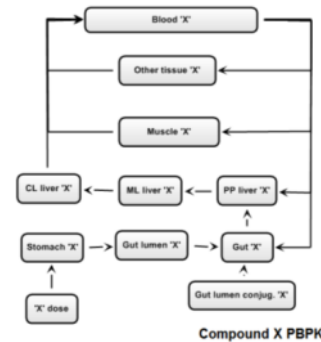
Mitochondrial dysfunction



Cellular life-cycle



Drug distribution & metabolism

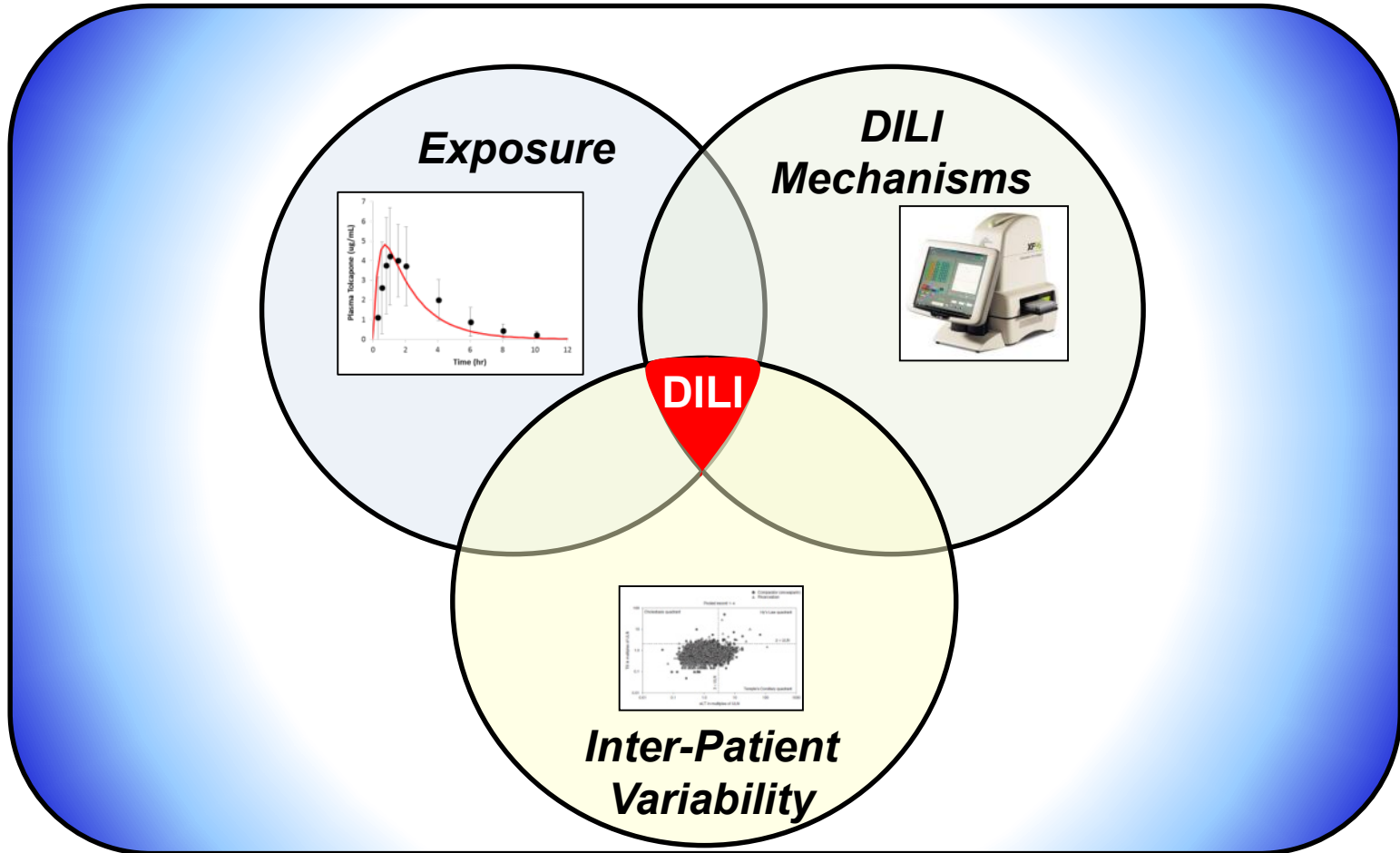


Kuefer 2010, Molecular Systems Biology

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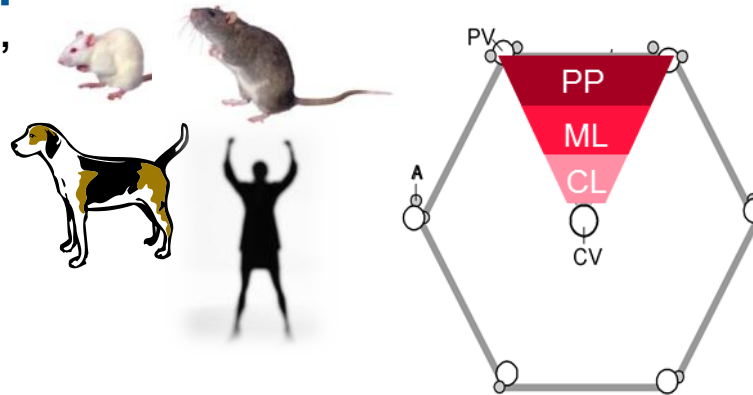
DILIsym Predicts DILI via the Intersection Between Exposure, Mechanisms, and Inter-Patient Variability



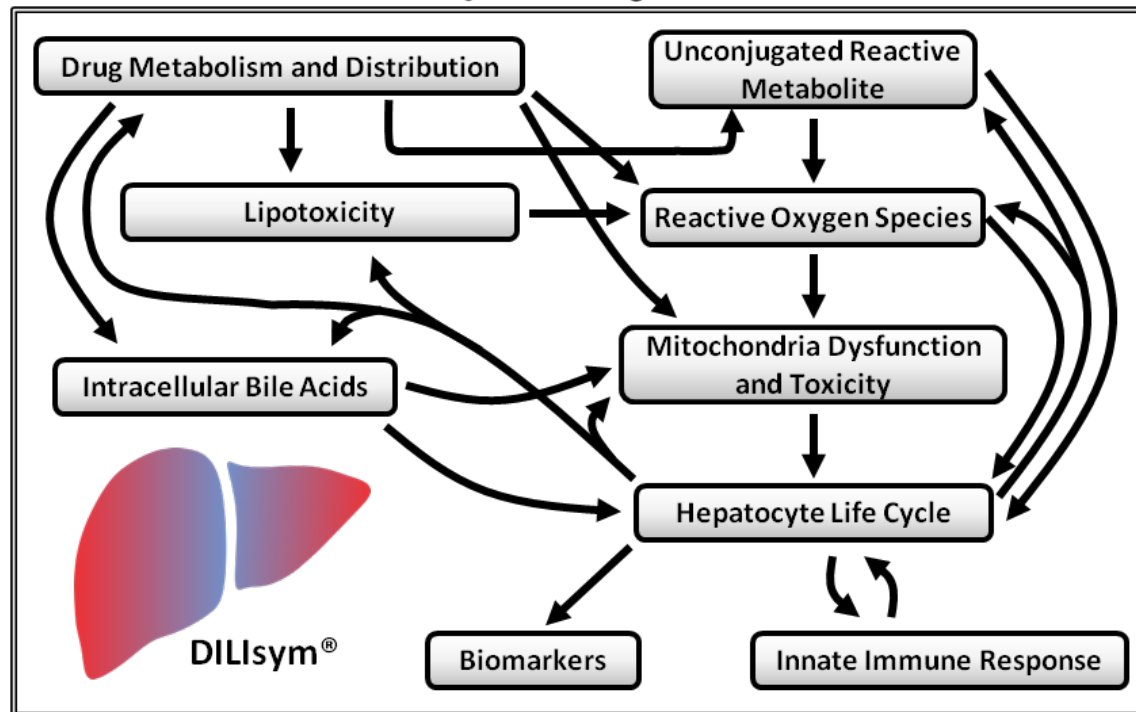
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DILIsym Overview: Multiple Sub-models Combined

- **Multiple species: human, rat, mouse, and dog**
 - Population variability
- **The three primary acinar zones of liver represented**
- **Essential cellular processes represented to multiple scales in interacting sub-models**



- **Over 30 detailed representations of optimization or validation compounds**
- **Single and combination drug therapies**



- Pharmacokinetics
- Dosing (IP, IV, Oral)
- Transporter Inhibition
- Drug metabolism
- GSH depletion
- Injury progression
- Mitochondrial dysfunction, toxicity, DNA depletion
- Bile acid mediated toxicity
- Steatosis and lipotoxicity
- Cellular energy balance
- Hepatocyte apoptosis and necrosis, and proliferation
- Macrophage, LSEC life cycles
- Immune mediators
- Caloric intake
- Biomarkers

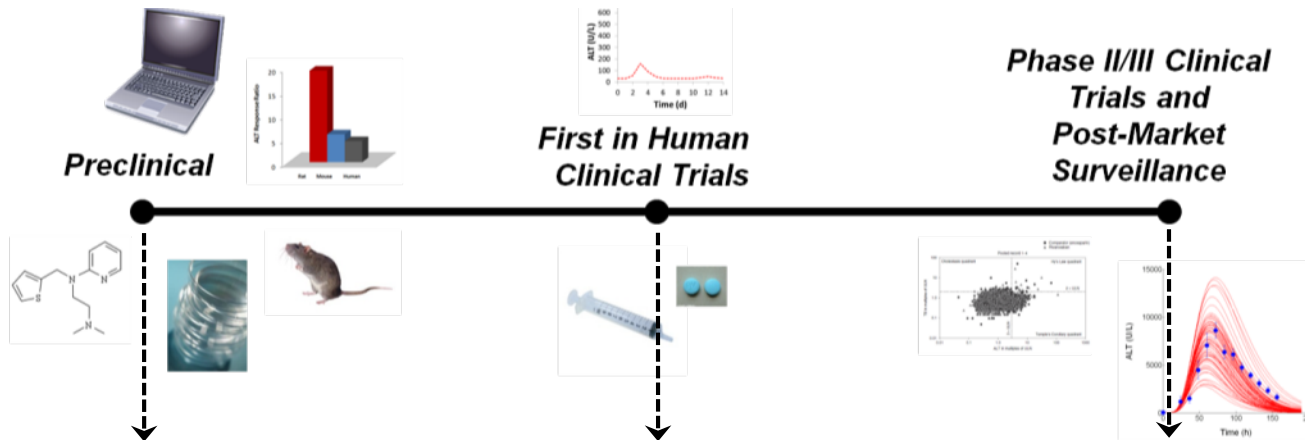
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Applications of DILIsym Along the Drug Development Pipeline



Predictions of hepatotoxicity for humans and preclinical animal models



- Mechanism exploration
- Rank candidates for DILI potential
- Extrapolation from animal and in vitro findings to humans

- Dose optimization (risk versus presumed benefit)
- Infer magnitude of injury based on measured biomarkers
- Extrapolation from healthy volunteers to patient groups
- Guide incorporation of emerging biomarker measurements in clinical trials
- Analysis of mechanisms underlying observed liver signals

- Inform choice and timing of biomarker measurement
- Aid identification of risk factors leading to personalized medicine approaches
- Analysis of mechanisms underlying observed liver signals

DILIsym Services

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Known DILsym Applications Submitted to or Intended for Regulatory Agencies

N	Agency	Context	Scenario	Simulation Type	Presented/ Submitted By
1	FDA	Simulation results included in formal, written correspondence to agency	Sponsor responding to concerns over liver safety signals	Hepatocyte loss (biomarker fitting)	Sponsor
2	FDA	Simulation results included in formal, written correspondence to agency	Sponsor responding to concerns over liver safety signals	Hepatocyte loss (biomarker fitting)	Sponsor
3	FDA	Simulation results included in formal, written correspondence to agency and presented during meeting	Sponsor responding to concerns over liver safety signals	Hepatocyte loss (biomarker fitting)	Sponsor and DSS
4	BARDA*	Simulation results presented to sponsor group at BARDA	Sponsor responding to concerns over liver safety signals	Mechanistic liver injury (predictive)	DSS and Sponsor
5	FDA and Japanese FDA	Simulation results included in formal, written correspondence to agency and presented during meeting	Sponsor addressing concerns over liver safety in NDA submission	Mechanistic liver injury (predictive)	Sponsor and DSS
6	FDA	Simulation results included in formal, written correspondence to agency and presented during meeting	Sponsor repurposing compound that failed due to hepatotoxicity in IND submission	Mechanistic liver injury (predictive)	Sponsor and DILsym Services
7	FDA	Simulation results included in formal, written correspondence to agency and presented during meeting	Sponsor addressing concerns over liver signals from other drug in same class with same indication	Mechanistic liver injury (predictive)	Sponsor
8	FDA	Simulation results included in formal, written correspondence to agency	Sponsor addressing concerns over liver safety in NDA submission	Mechanistic liver injury (predictive)	Sponsor
9	FDA	Simulation results included in formal, written correspondence to agency and discussed during call with FDA	Sponsor responding to concerns over liver safety signals	Hepatocyte loss (biomarker fitting)	Sponsor
10	FDA and global regulators	Sponsor intended to submit simulation results	Sponsor addressing concerns over liver safety signals	Hepatocyte loss (biomarker fitting) Mechanistic liver injury (predictive)	Sponsor
11	FDA	Sponsor intended to submit simulation results	Sponsor addressing concerns over liver signals from other drug in same class with same indication	Mechanistic liver injury (predictive)	Sponsor
12	FDA	Sponsor intended to submit simulation results	Sponsor reformulating existing compound on the market	Mechanistic liver injury (predictive)	Sponsor
13	FDA	Sponsor intended to submit simulation results and present at meeting	Sponsor addressing concerns over liver safety signals	Mechanistic bilirubin (predictive)	Sponsor

*Not a direct regulatory agency, but affiliated closely with NIH and FDA

**Several additional sponsors have declared intent to include results in regulatory communications in the future

***Additional drug development teams have implied that regulators have informally requested or recommended DILsym simulations

Scientists at the FDA Have Expressed a Strong Interest in DILIsym Results

PERSPECTIVES

See ARTICLE page 589

Application of Systems Pharmacology to Explore Mechanisms of Hepatotoxicity

J Shon¹ and DR Abernethy¹

Advances in systems biology have allowed the development of a highly characterized systems pharmacology model to study mechanisms of drug-induced hepatotoxicity. In this issue of *CPT*, Yang *et al.* describe a model, DILIsym, used to characterize mechanisms of hepatotoxicity of troglitazone. Their modeling approach has provided new insight into troglitazone-induced hepatotoxicity in humans but is not associated with hepatotoxicity in rats, consistent with preclinical data for this drug.

“We look forward to future efforts to apply this model for prediction of hepatotoxicity that has not been clinically observed.”

FDA Office of Clinical Pharmacology

The views expressed here are those of the author and do not necessarily represent the opinions of the author's institution, the United States Food and Drug Administration, the Department of Health and Human Services, the Department of Defense, or the Department of Veterans Affairs.

CONFLICT OF INTEREST

The author declared no conflict of interest.

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1. January, C.T. & Riddle, J.M. Early after depolarizations: mechanism of induction and block: a role for L-type Ca^{2+} current. *Circ. Res.* 64: 877-890 (1989).

¹Office of Clinical Pharmacology, Office of Translational Sciences, Center for Drug Evaluation and Research, US Food and Drug Administration, Silver Spring, Maryland, USA. Correspondence: DR Abernethy (Damell.Abernethy@fda.hhs.gov)

doi:10.1038/cpt.2014.167

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DILIsym Services

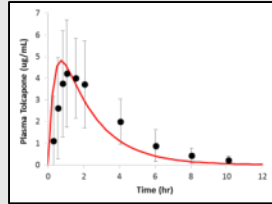
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DILIsym Utilizes Various Data Types to Inform Decisions

Exposure Data

PBPK Modeling

- **Compound Properties**
 - Tissue partition coefficients
- **Tissue penetration studies**
 - Liver to blood ratio
- **Pharmacokinetic data**
 - Absorption, extra-hepatic clearance, metabolites
- **in vitro data**
 - Metabolite synthesis, active uptake



Modeling & Simulation

Simulations and Assays inform:

- Prediction of DILI risk
- Participating DILI mechanisms
- Characteristics of patients at risk for DILI
- Drug dosing paradigms
- DILI monitoring strategies



In vitro Mechanistic DILI Data

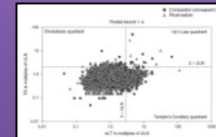
Assays performed to determine quantitative aspects of DILI mechanisms

- **Oxidative stress**
 - Direct and reactive metabolite-mediated
- **Mitochondrial toxicity**
 - ETC inhibition
 - Uncoupling
- **Bile acid transporter inhibition**
 - BSEP, MRP3 and 4, NTCP
- **Bilirubin transport/metabolism**
 - OATP1B1, OATP1B3, UGT1A1, MRP2, MRP3



Clinical Data

- **Biomarkers**
 - Timing and magnitude of injury
- **Anthropometric data**
 - Body weight, age, ethnicity
- **Pharmacokinetic data**
 - Absorption, extra-hepatic clearance, metabolites



DILIsym Services

Example Project Executive Summary

- GastroPlus™ software, along with *in vitro* data, was used to construct PBPK representations to predict liver exposures for both compounds
- DILIsym parameters were successfully calculated from *in vitro* data for both compounds
- SimPops results show Compound X and Compound Y to be safe at projected clinical doses
- ALT elevations predicted within DILIsym at higher doses for both compounds
- SimPops results suggest that neither compound is likely to cause severe liver injury

Example Project Outline

- Goals
- Summary of mechanistic DILI *in vitro* assay results and DILIsym parameters for Compound Y and Compound X
 - Mitochondrial dysfunction
 - Oxidative stress
 - Bile acid transport inhibition
- Physiologically-based pharmacokinetic (PBPK) modeling in GastroPlus to represent *in vivo* exposure for Compound X and Compound Y
- Simulation results in the full simulated population (SimPops) for Compound X and Compound Y
- Analysis of DILI mechanisms for Compound X and Compound Y

Example Project Goal – Assess Compound X and Compound Y

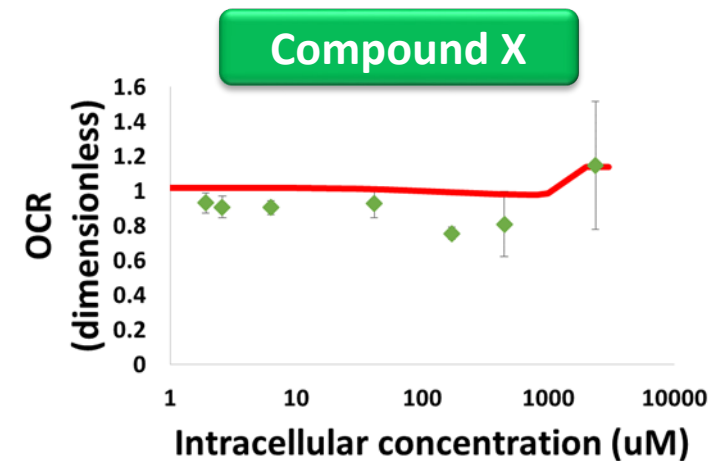
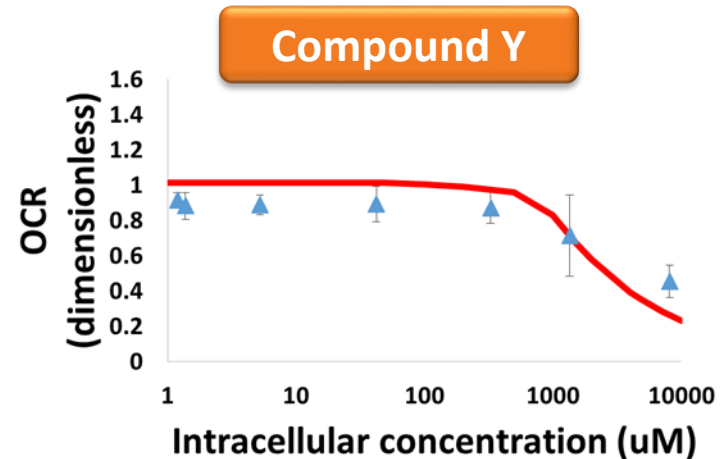
- The primary goal of this simulation work within the DILIsym software was to:
 - quantitatively and mechanistically assess the liver toxicity potential of Compound X and Compound Y combining clinical and mechanistic *in vitro* data with DILIsym and GastroPlus software simulations of previous or prospective clinical dosing paradigms.

Example Project Outline

- Goals
- Summary of mechanistic DILI *in vitro* assay results and DILIsym parameters for Compound Y and Compound X
 - Mitochondrial dysfunction
 - Oxidative stress
 - Bile acid transport inhibition
- Physiologically-based pharmacokinetic (PBPK) modeling in GastroPlus to represent *in vivo* exposure for Compound X and Compound Y
- Simulation results in the full simulated population (SimPops) for Compound X and Compound Y
- Analysis of DILI mechanisms for Compound X and Compound Y

Mitochondrial Toxicity Parameters Determined for Compound Y and Compound X

- Parameter values were fit to mitochondrial data for Compound Y and Compound X
 - Electron transport chain inhibition for Compound Y
 - Both electron transport chain inhibition and uncoupling for Compound X
 - 24 hour data used
- MITOsym and DILIsym used to parameterize both compounds



DILIsym Parameter	Compound Y Value	Compound X Value	Units
Coefficient for ETC inhibition 1	38,000	Not used	μM
Coefficient for ETC Inhibition 3	0.1	4,200	μM
Max inhibitory effect for ETC inhibition 3	0.2	0.4 (max effect)	dimensionless
Uncoupler 1 effect Km	No effect	15,000	μM
Uncoupler 1 effect Vmax	No effect	22	dimensionless
Uncoupler 1 effect Hill	No effect	4	dimensionless

Preclinical Data and Simulation Results

DILIsym Services

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Oxidative Stress Parameters Determined for Compound Y and Compound X

- Parameter values were fit to 24-hour ROS data for Compound Y and Compound X
- DILIsym representation of *in vitro* environment used to parameterize both compounds
- Saturable model explored but did not lead to better fit

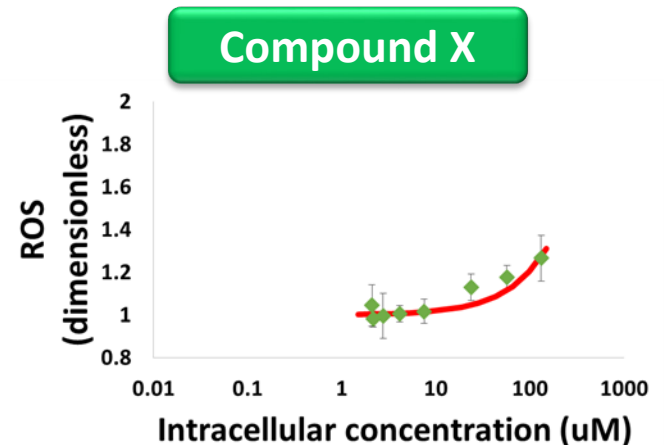
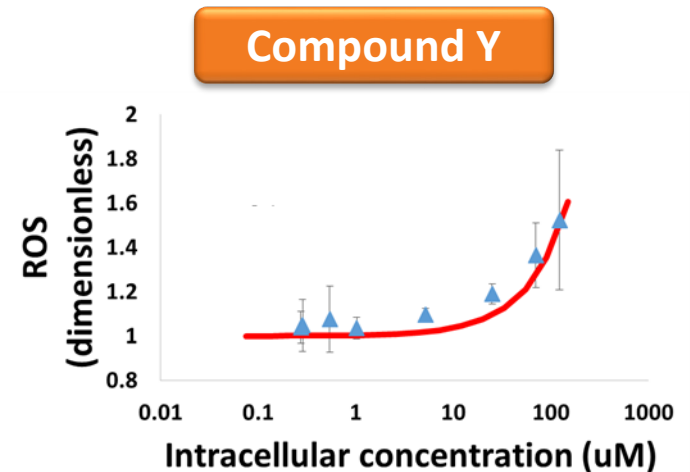
DILIsym Parameter	Compound Y Value	Compound X Value	Units
RNS/ROS production rate constant 1	3.4×10^{-4}	1.7×10^{-4}	mL/nmol/hr



Preclinical Data and Simulation Results

DILIsym Services

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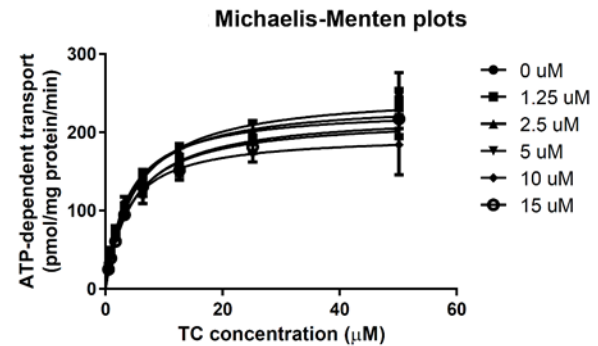


Compound Y Weakly Inhibits BSEP; Compound X Does Not

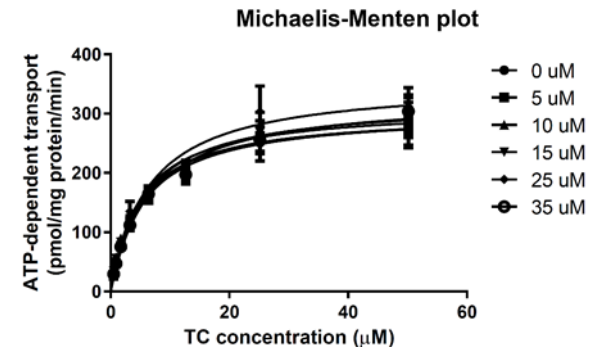
- Compound Y is a weak but noncompetitive/uncompetitive inhibitor of BSEP
- Compound X does not inhibit BSEP
 - No changes to V_{max} or K_m of transporters observed over course of assay



Compound Y; $K_i = 140 \mu\text{M}$, $\alpha = 0.6$



Compound X; no inhibition



DILIsym Toxicity Parameters for Compound Y and X

Mechanism	Parameter	Unit	DILIsym Parameter Value*	
			Compound Y	Compound X
Mitochondrial Dysfunction	Coefficient for ETC inhibition 1	μM	38,000	Not used
	Coefficient for ETC Inhibition 3	μM	0.1	4,200
	Max inhibitory effect for ETC inhibition 3	dimensionless	0.2	0.4
	Uncoupler 1 effect Km	μM	No effect	15,000
	Uncoupler 1 effect Vmax	dimensionless	No effect	22
	Uncoupler 1 effect Hill	dimensionless	No effect	4
Oxidative Stress	RNS/ROS production rate constant 1	mL/nmol/hr	3.4×10^{-4}	1.7×10^{-4}
Bile Acid Transporter Inhibition	BSEP inhibition constant	μM	140	No inhibition
	BSEP inhibition alpha value	dimensionless	0.6	No inhibition
	NTCP inhibition constant	μM	No inhibition	No inhibition
	MRP4 inhibition constant	μM	40	75

*Values shown in the table for DILIsym input parameters should not be interpreted in isolation with respect to clinical implications, but rather, should be combined with exposure in DILIsym to produce simulations that have predictive and insightful value

DILIsym Services

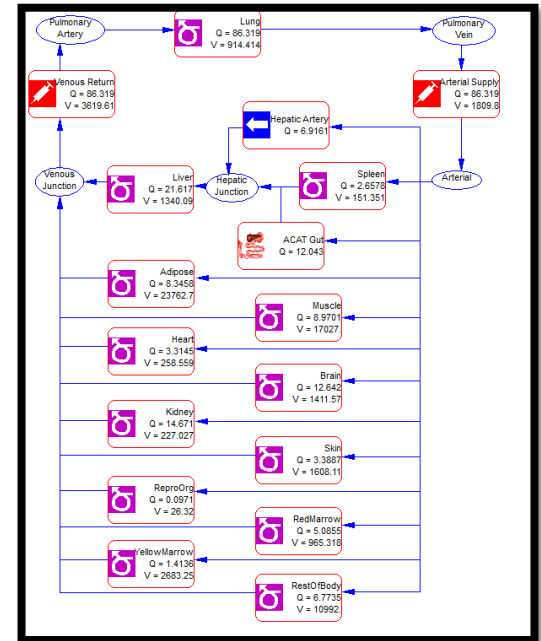
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Example Project Outline

- Goals
- Summary of mechanistic DILI *in vitro* assay results and DILIsym parameters for Compound Y and Compound X
 - Mitochondrial dysfunction
 - Oxidative stress
 - Bile acid transport inhibition
- Physiologically-based pharmacokinetic (PBPK) modeling in GastroPlus to represent *in vivo* exposure for Compound X and Compound Y
- Simulation results in the full simulated population (SimPops) for Compound X and Compound Y
- Analysis of DILI mechanisms for Compound X and Compound Y

GastroPlus PBPK Model Used to Predict Liver Exposure of Compound Y and Compound X

- Data on Compound Y and Compound X pharmacokinetics not available in the literature
 - No plasma time courses available; no *in vitro* or animal studies available either
 - Data on T_{max} , Compound Y $f_{u,plasma}$ available
 - *In vitro* data on liver distribution available from intracellular data collected for this project
- Structure of each compound available online
 - QSAR modeling using ADMET Predictor and GastroPlus provided the best possible estimate of Compound Y and Compound X distribution and pharmacokinetics
- Plasma time course was estimated in GastroPlus and translated into DILIsym using “specified data” option
 - Liver:plasma partition coefficient was calculated from the cell:media ratio in the *in vitro* data and used as input into GastroPlus; the remainder of the parameters were calculated by ADMET Predictor
- Both compounds distribute significantly into the liver
 - Compound Y average cell:media was 18; Compound X average cell:media was 9



Compound Y

Compound X

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Compound Y PBPK Representation Calculated at Clinical Dose

- GastroPlus predictions for liver and plasma at clinical dose shown at right
 - PBPK model specific predictions shown below
 - Dose escalation was simulated

Blood/plasma Conc Ratio: 0.72

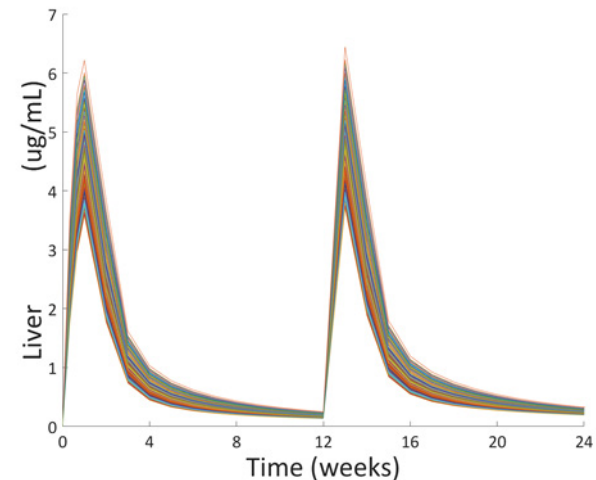
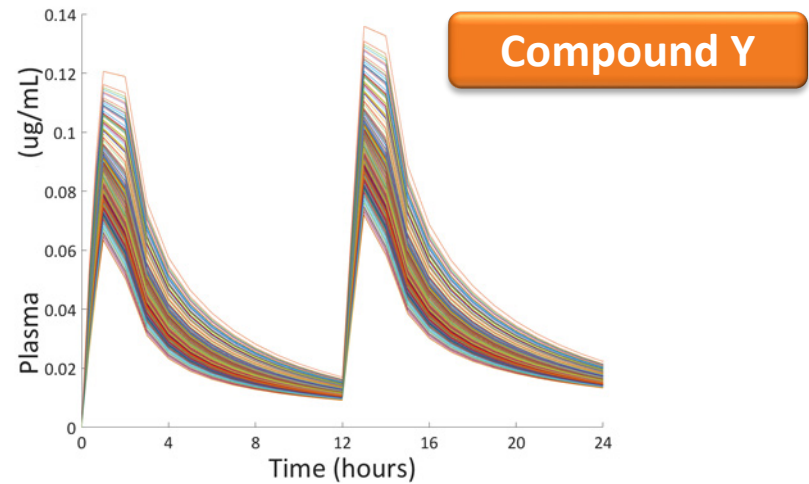
Scale Pediatric Fup & Rbp

Use Exp Plasma Fup [%]: 4.3

Use Adj Plasma Fup [%]: 2.6893

PBPK Summary

Tissue	Kp	CL	CLint	Fut/Fulnt
Hepatic Artery	0.00	0.000	0.000	0.000
Lung	0.51	0.000	0.000	0.053
Arterial Supply	0.00	0.000	0.000	0.000
Venous Return	0.00	0.000	0.000	0.000
Adipose	5.33	0.000	0.000	0.005
Muscle	1.66	0.000	0.000	0.016
Liver	18.30	0.000	0.000	0.001
ACAT Gut	0.00	0.000	0.000	0.000
Spleen	1.69	0.000	0.000	0.016
Heart	1.89	0.000	0.000	0.014
Brain	4.24	0.000	0.000	0.006
Kidney	1.69	0.318	0.000	0.016
Skin	2.17	0.000	0.000	0.012
ReproOrg	1.70	0.000	0.000	0.016
RedMarrow	4.70	0.000	0.000	0.006
YellowMarrow	5.33	0.000	0.000	0.005
RestOfBody	1.71	0.000	0.000	0.016



Compound X PBPK Representation Calculated at Clinical Dose

- GastroPlus predictions for liver and plasma at clinical dose for 25 days shown at right
 - PBPK model specific predictions below
 - Dose escalation and alternate protocols were also simulated

Blood/plasma Conc Ratio:

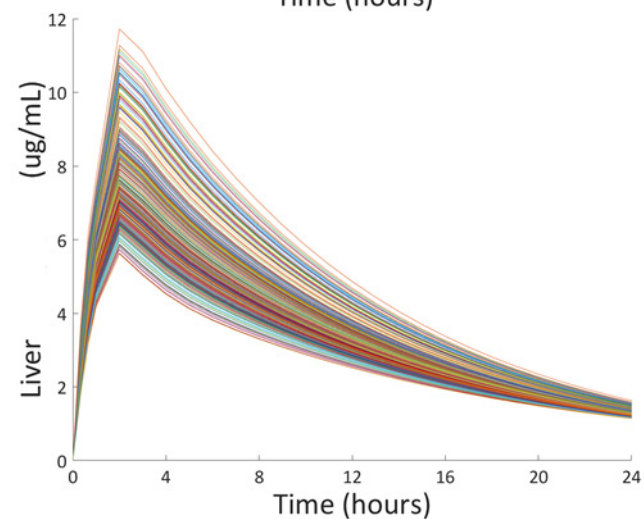
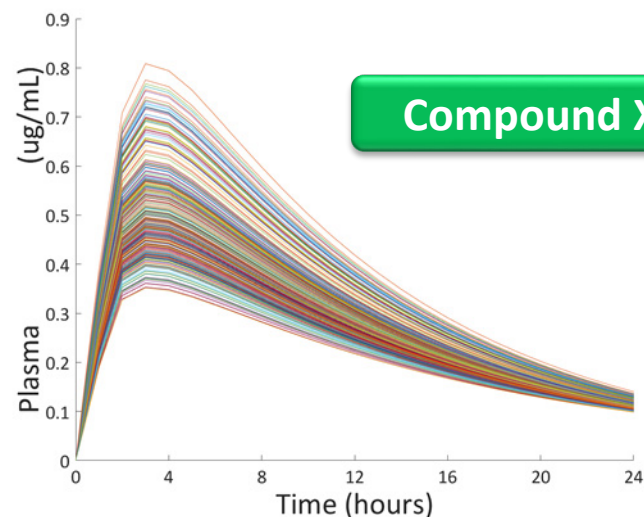
Scale Pediatric
Fup & Rbp

Use Exp Plasma Fup [%]:

Use Adj Plasma Fup [%]:

PBPK Summary

Tissue	Kp	CL	CLint	Fut/Fulnt
Hepatic Artery	0.00	0.000	0.000	0.000
Lung	0.30	0.000	0.000	0.125
Arterial Supply	0.00	0.000	0.000	0.000
Venous Return	0.00	0.000	0.000	0.000
Adipose	1.11	0.000	0.000	0.034
Muscle	0.48	0.000	0.000	0.079
Liver	9.34	0.000	0.000	0.004
ACAT Gut	0.00	0.000	0.000	0.000
Spleen	0.51	0.000	0.000	0.074
Heart	0.60	0.000	0.000	0.063
Brain	1.10	0.000	0.000	0.034
Kidney	0.53	0.309	0.000	0.071
Skin	0.75	0.000	0.000	0.050
ReproOrg	0.54	0.000	0.000	0.070
RedMarrow	1.28	0.000	0.000	0.030
YellowMarrow	1.11	0.000	0.000	0.034
RestOfBody	0.53	0.000	0.000	0.071



Simulation Results

DILIsym Services

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Example Project Outline

- Goals
- Summary of mechanistic DILI *in vitro* assay results and DILIsym parameters for Compound Y and Compound X
 - Mitochondrial dysfunction
 - Oxidative stress
 - Bile acid transport inhibition
- Physiologically-based pharmacokinetic (PBPK) modeling in GastroPlus to represent *in vivo* exposure for Compound X and Compound Y
- Simulation results in the full simulated population (SimPops) for Compound X and Compound Y
- Analysis of DILI mechanisms for Compound X and Compound Y

SimPops Results Show Compound X and Compound Y to be Safe at Clinical Doses; ALT Elevations Occur at Higher Doses for Both Compounds

Compound Y

Compound X

- Neither Compound Y nor Compound X are predicted to cause toxicity at the highest clinical dose
 - Some exposure variability included in these predictions due to GastroPlus population generation
- Both Compound Y and Compound X are predicted to cause mild ALT elevations at supratherapeutic doses
 - No bilirubin elevations or Hy's Law cases occurred in simulations with Compound X
 - 2 Hy's Law cases occurred at 10x clinical dose simulations with Compound Y

	Compound	Dosing Protocol	Simulated* ALT > 3X ULN**
Compound Y	Compound Y	1X Dose, 12 weeks	0% (0/285)
		2X Dose, 12 weeks	0% (0/285)
		5X Dose, 12 weeks	0.3% (1/285)
		10X Dose, 12 weeks	10.2% (29/285)
Compound X	Compound X	1X Dose, 15 days	0% (0/285)
		2X Dose, 15 days	0% (0/285)
		5X Dose, 15 days	1.1% (3/285)
		10X Dose, 15 days	11.6% (33/285)

*The full v4A-1 SimPops (n=285) of normal healthy volunteers was used

**Upper limit of normal (ULN) in DILIsym is 40 U/L

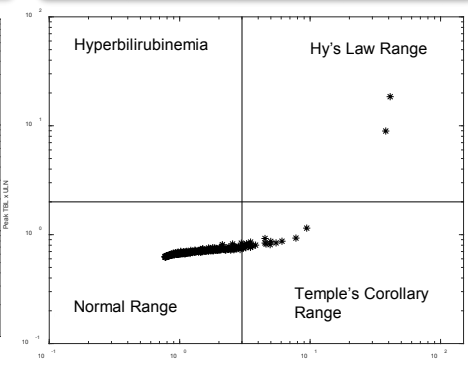
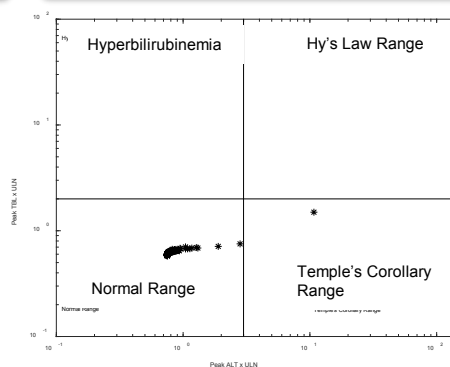
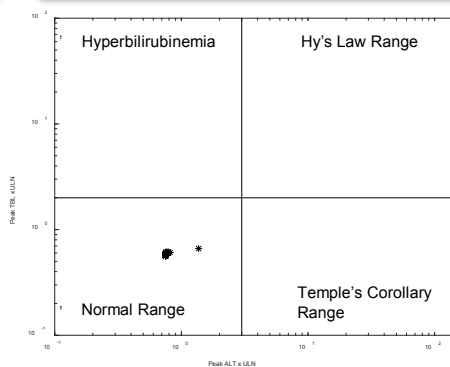
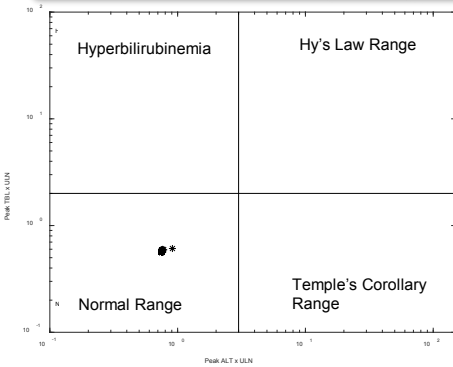
SimPops Results Show Lack of Severe Liver Injury for Both Compound Y and Compound X at Clinical Doses

Compound Y; 1X Dose, 12 weeks

Compound Y; 2X Dose, 12 weeks

Compound Y; 5X Dose, 12 weeks

Compound Y; 10X Dose, 12 weeks

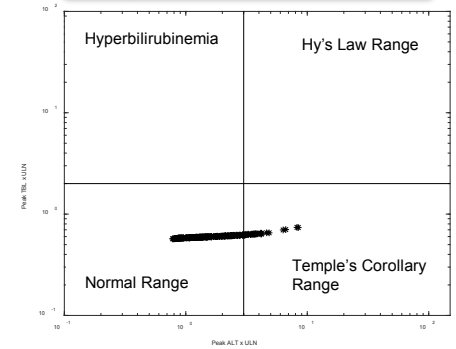
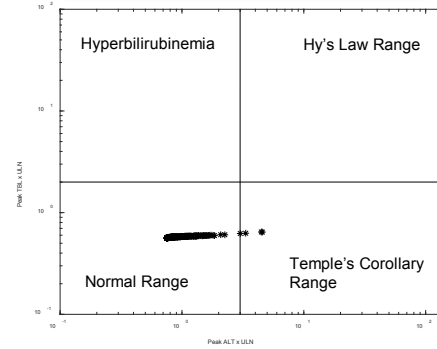
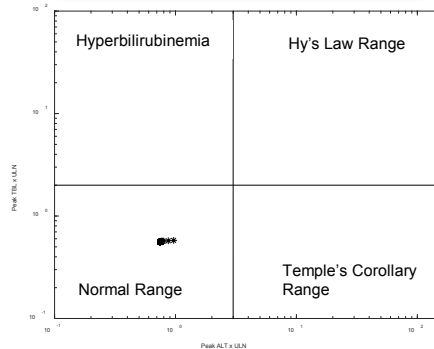
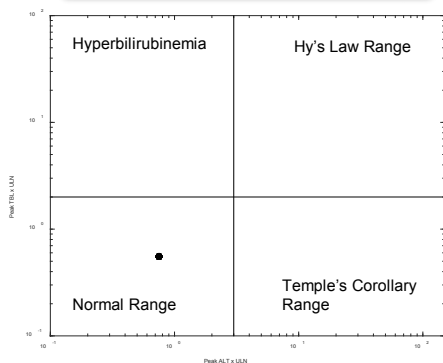


Compound X; 1X Dose

Compound X; 2X Dose

Compound X; 5X Dose

Compound X; 10X Dose



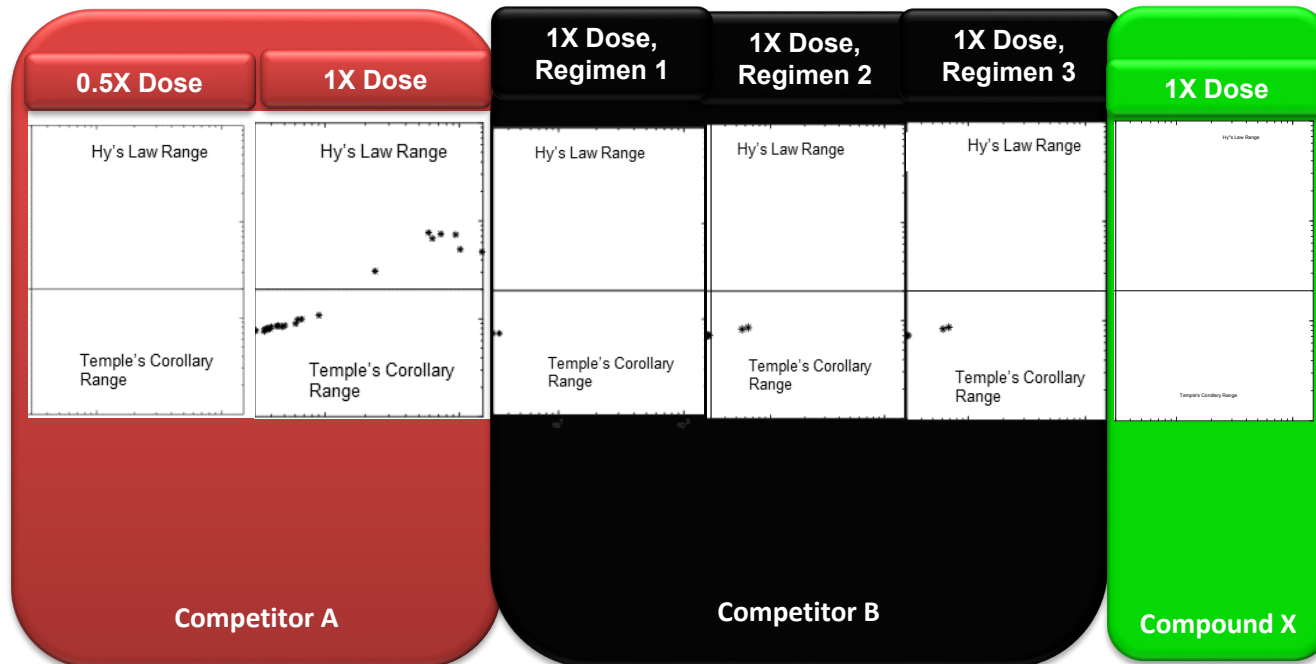
Simulation Results

DILsymS

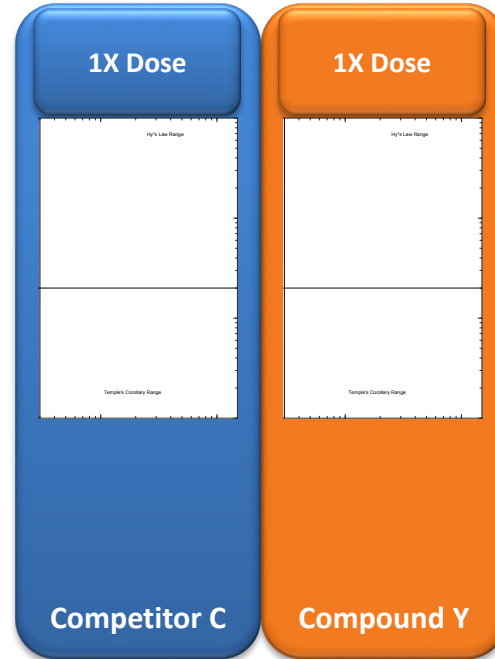
*The full v4A-1 SimPops (n=285) of normal healthy volunteers was used
**Upper limit of normal (ULN) in DILsym is 40 U/L

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Focus on Hy's Law Side of eDISH Plot – Comparison of Competitors and Compound X at Clinical Doses (285 Simulated Individuals in All Cases)



Focus on Hy's Law Side of eDISH Plot – Comparison of Competitor and Compound Y at Predicted Clinical Doses (285 Simulated Individuals in All Cases)



Simulation Results

DILIsym Services

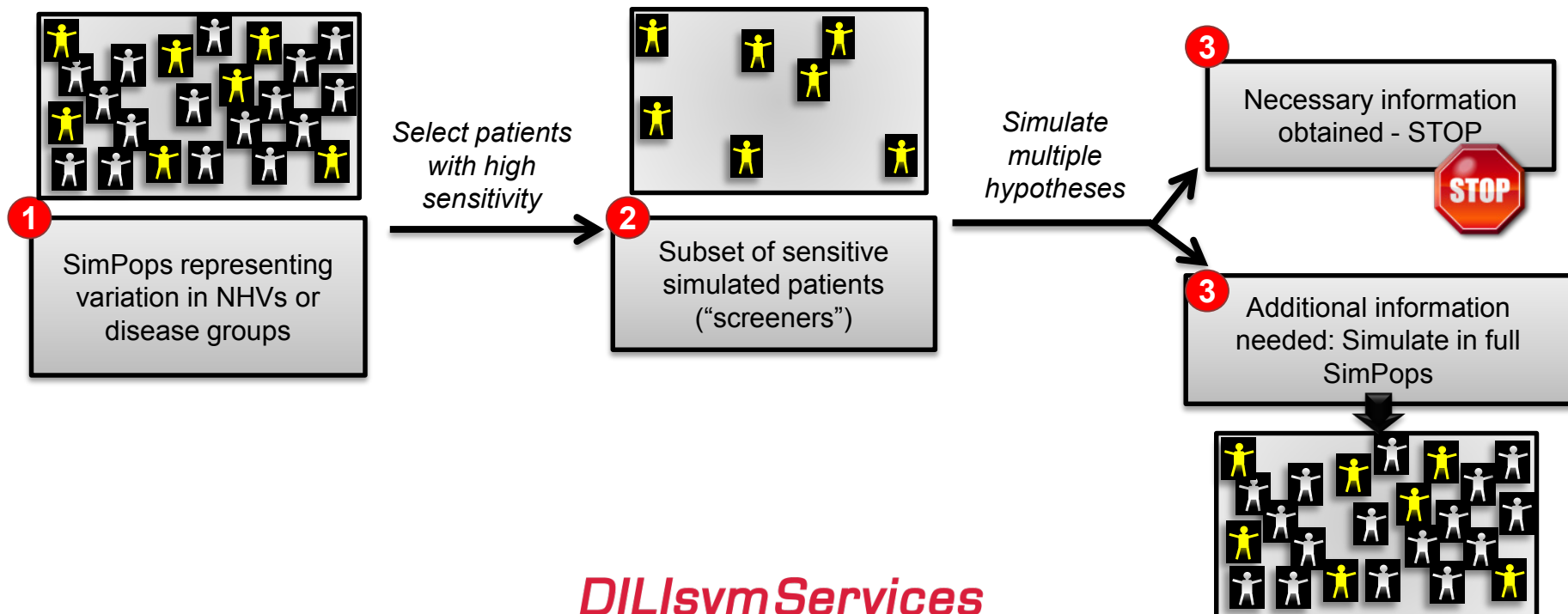
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Example Project Outline

- Goals
- Summary of mechanistic DILI *in vitro* assay results and DILIsym parameters for Compound Y and Compound X
 - Mitochondrial dysfunction
 - Oxidative stress
 - Bile acid transport inhibition
- Physiologically-based pharmacokinetic (PBPK) modeling in GastroPlus to represent *in vivo* exposure for Compound X and Compound Y
- Simulation results in the full simulated population (SimPops) for Compound X and Compound Y
- Analysis of DILI mechanisms for Compound X and Compound Y

Subsets of Simulated Humans (SimCohorts) are Used for Preliminary Simulations in DILIsym

- SimPops include variability in toxicity mechanisms for healthy human volunteers and some disease populations
- SimCohorts consisting of a subset of sensitive individuals to specific DILI mechanisms from existing SimPops are used for screening and preliminary simulation purposes
- Simulations were conducted in a SimCohorts sensitive to specific DILI mechanisms, along with the baseline human simulations to produce preliminary results
- SimCohorts used: **Human_ROS_apop_mito_BA_v4A_1_Multi16**; includes some sensitive individuals from v4A_1 SimPops for bile acid, mitochondrial, ROS mechanisms and combinations (BA + mito) plus some insensitive individuals and the baseline individual



Mechanistic Investigation Simulation Results Show ROS as Main Driver of Compound X and Compound Y ALT Elevations

- Mechanistic investigation simulations for Compound Y and Compound X on a subset of susceptible individuals at 10x clinical dose
 - One mechanism turned off at a time; if turning off a mechanism leads to lower frequency of ALT elevations, that mechanism contributes to the simulated toxicity
- Compound Y and Compound X ALT elevations at simulated supratherapeutic doses are mostly due to ROS (oxidative stress) generation

	Compound and Protocol	Mechanisms On	Mechanisms Off	Simulated* ALT > 3X ULN**
Compound Y		All	None	29/29
		BA inhibition, ETC inhibition	ROS generation	2/29
		BA inhibition, ROS generation	ETC inhibition	24/29
		ETC inhibition, ROS generation	BA inhibition	24/29
Compound X		All	None	33/33
		BA inhibition, ETC inhibition, Uncoupling	ROS generation	0/33
		BA inhibition, Uncoupling, ROS generation	ETC inhibition	33/33
		BA inhibition, ETC inhibition, ROS generation	Uncoupling	33/33
		BA inhibition, ROS generation	ETC inhibition, Uncoupling	33/33
		ETC inhibition, Uncoupling, ROS generation	BA inhibition	33/33

Compound Y

Compound X

Simulation Results

DILIsymS

*The full v4A-1 SimPops (n=285) of normal healthy volunteers was used

**Upper limit of normal (ULN) in DILIsym is 40 U/L

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Summary of Key Points

- Combining QSAR, PBPK, and QST models is a powerful approach to getting more information out of your data investments early in product development
- The Simulations Plus family has extensive experience with these approaches and can help
- ADMET Predictor / GastroPlus / DILIsym is a winning software package combination for predicting the liver safety of your molecules and those of your competitors prior to clinical trial surprises

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SCIENCE + SOFTWARE = SUCCESS

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Visit us in booth #442 for product demonstrations

SOT presentations:

Altered bile acid homeostasis and mitochondrial function: potential mechanisms for BMS-986020-induced human hepatobiliary toxicity

Monday @ 1:30 PM; Poster board #P814

In silico-in vitro extrapolation for dermal exposure

Tuesday @ 10:10 AM; Hemisfair Ballroom C3

Mechanistic modeling of mitochondrial biogenesis within DILIsym could explain clinically observed adaptation of ALT elevations

Tuesday @ 3:00 PM; Poster board #P294

Prediction of the liver toxicity of the endothelin receptor antagonists sitaxsentan and ambrisentan for the treatment of pulmonary arterial hypertension with a QST tool

Wednesday @ 1:30 PM; Poster board #P836

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