

St Simulations Plus

Machine Learning ADMET Predictor Workflow for DILIsym Use Enables Earlier, Higher Throughput Use

ACT 2024 Lunch and Learn!

Monday, November 18, 2024

Session Topics

- Simulations Plus: Who We Are, and What We Do
 - Overview
 - News and Events
 - DILIsym Case Studies
- NEW! Liver Safety+ Package
 - Machine Learning ADMET Predictor Workflow for DILIsym Use Enables Earlier, Higher Throughput Use
- Questions Please!



Who We Are

NASDAQ: SLP



Physiologically Based Pharmacokinetics (PBPK) Software & Services

Clinical Pharmacology & Pharmacometrics (CPP) Software & Services

Quantitative Systems Pharmacology (QSP) Software & Services

Adaptive Learning & Insights (ALI) Software & Services

Medical Communications (MC) Software & Services



>280 Pharmaceutical, biotechnology, chemicals, cosmetics, & consumer goods companies in the U.S., Europe, Asia, and South America

200+ Employees Worldwide

>25 Established yrs. In 1996





Complementary Solutions

Confidence Level on Clinical Performance

Clinical Pharmacology & Pharmacometrics

Medical Communications

Simulations Plus Has the World's Largest Library of Platform QSP and QST Models to Predict Disease and Injury Outcomes

QST: Liver and Kidney Safety

- Drug induced acute kidney injury
- Drug induced liver injury (DILI)



QSP: Metabolic Diseases

- Non-alcoholic fatty liver disease / steatohepatitis (NAFLD/NASH or MASH)
- Obesity



QSP: Immuno-Oncology

- Acute myeloid leukemia (AML)
- Diffuse large B-cell lymphoma (DLBCL)
- Multiple myeloma (MM)
- Myelofibrosis
- Solid tumor (NSCLC, melanoma, prostate cancer, colorectal cancer, ovarian cancer, endometrial cancer)

QSP: Inflammation and Immunology (including fibrotic diseases)

- Asthma/COPD (in development)
- Atopic dermatitis (AD)
- Crohn's disease (CD)
- Dermatomyositis
- Dysregulation of alternative and terminal pathways (AP, TP) of complement
- Idiopathic pulmonary fibrosis (IPF)
- Interstitial lung disease (ILD) associated with systemic sclerosis

S + SimulationsPlus

- Multiple sclerosis (MS, in development)
- Psoriatic arthritis (PSA)
- Psoriasis (PSO)
- Rheumatoid arthritis (RA)
- Systemic lupus erythematosus (SLE including CLE)
- Ulcerative colitis (UC)
- Uric acid disposition in gout
- Wound healing after myocardial infarction (MI)

What's It Like to Work With Us?

We believe the relationships we build with our clients are critical for mutual success

A highly interactive collaboration not only allows us to deliver results as quickly as possible, but also ensures a higher quality deliverable

- Regular interactions ensure the relevancy of results as the knowledge-base continues to evolve Transparency provided by progress updates eliminates surprises
- Synergies come from a shared knowledge-base of expertise and experience
- Involvement, participation, and input from stakeholders outside of M&S is welcome







	CONGRESS.GOV Advanced Searches Browse Legislation Examples: hr5. sres9. "health care"	Search Tools Support • Sign In • Congressional Record Committees Members	
Physiologic	MORE OPTIONS ~ Home > Legislation > 117th Congress > S.5002 S.5002 - FDA Modernization Act 2.0 117th Congress (2021-2022)	ation 🖸 Subscribe 🖪 Share/Save 🗩 Site Feedback	
Analyses — Cont Guidance fo	BILL Hide Overview X Sponsor: Sen. Paul. Rand [R-KY] (Introduced 09/29/2022) Latest Action: House - 09/29/2022 Held at the desk. (All Actions) Tracker: Introduced Passed Senate	More on This Bill <u>CBO Cost Estimates (9)</u> Subject — Policy Area: Health <u>View subjects</u> »	e document on the ation, validation and Physiologically Based pmodels for regulatory purposes
2018	Summary (2) Text (2) Actions (4) Titles (3) Amendments (0) Cosponsors (11) Committees (0) Related Bills (2) Summary: S.5002 — 117th Congress (2021-2022) Image: I	<u>All Information</u> (Except Text)	
996	There are 2 summaries for S.5002. Passed Senate (09/29/2022) Bill summaries are authored by CRS. Shown Here: Passed Senate (09/29/2022) FDA Modernization Act 2.0 This bill authorizes the use of certain alternatives to animal testing, including cell-based assays and computer models, to obtain an exemption from the For and effectiveness of a drug. The bill also removes a requirement to use animal studies as part of the process to obtain a license for a biological product that is biosimilar or interchangement.	ood and Drug Administration to investigate the safety geable with another biological product.	ance Document

Breaking News...

June 20, 2023 8:00 AM

Simulations Plus Acquires Immunetrics to Expand its Immunology and Oncology Drug Development Capabilities

Acquisition increases breadth and depth of QSP expertise and range of therapeutic applications

Simulations Plus, Inc. (Nasdaq: SLP) ("Simulations Plus"), a leading provider of modeling and simulation software and services for pharmaceutical safety and efficacy, today announced the acquisition of Immunetrics, Inc. ("Immunetrics"), a modeling and simulation company focused on

° April 4, 2024 8:30 AM

U.S. FDA Renews DILlsym® Software Licenses for 7th Year

Predicting DILI risk supports informed decision-making regarding drug evaluations and approvals

June 12, 2024 8:00 AM

Simulations Plus Acquires Pro-ficiency, Creating One-of-a-Kind Platform Spanning the Drug Development Continuum

Acquisition doubles the Company's TAM to \$8 billion

July 30, 2024 8:30 AM

Simulations Plus Releases ADMET Predictor® Version 12

Enhancements in key models power HT-PBPK simulations and Al-driven drug design with unprecedented performance and accuracy

Simulations Plus. Inc. (Nasdag SLP) ("Simulations Plus"), a leading provider of biosimulation, simulation-enabled performance and intelligence solutions, and medical communications to the biopharma industry, today announced the release of version 12.0 of ADMET Predictor® (AP12), its flagship machine learning (ML) modeling platform for the discovery, design, and optimization of new molecules.

The con AP12 includes:

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- Enhanced Models: New and expanded models offer greater predictive accuracy, with an average 30% increase in training set sizes, for microsome and hepatocyte clearance, protein binding, biorelevant solubilities, MDCK-LE/PAMPA permeability, and more.
- High-Throughput Pharmacokinetics (HTPK): New options for solution dosing, adjusted free fraction outputs, and species-specific simulations enhance the flexibility and precision of HTPK studies.
- Artificial Intelligence-Driven Drug Design (AIDD): Integration of 3D shape matching and tissue sensitivities (based on tissue Kp values) as
 new objectives, facilitating innovative lead optimization processes.
- New DILI Module : Introduction of the first drug-induced liver injury (DILI) endpoint models to support high-throughput (HT) DILIsym®
 predictions in early drug development.
- Boosted ANN Regression Models and added 37 new descriptors in ADMET Modeler™.
- General Usability and Informatics Improvements.

February 15, 2024 8:30 AM

Simulations Plus Extends Collaboration with Major Toxicology Research Agency

Research project with NIEHS includes focus on qualification of in silico methods for prioritization, assessment of risk, and identification of safety margins for chemical use

May 15, 2024 8:30 AM

Simulations Plus Releases GastroPlus® X, The Next Generation PBPK/PBBM Modeling & Simulation Software

Redesigned platform offers ease-of-use, enhanced software engineering, and significant productivity gains for users

July 11, 2024 8:30 AM

Simulations Plus Announces New Research Project with the International Collaboration on Cosmetics Safety

bjective to define best practices for the use of novel PBK modeling strategies to support animal-free safety assessment of new chemicals

August 1, 2024 8:30 AM

Simulations Plus Releases DILIsym® X

Updated quantitative systems toxicology (QST) software investigates and predicts drug-induced liver injury (DILI)

Simulations Plus, Inc. (Nasdaq: SLP) ("Simulations Plus"), a leading provider of biosimulation, simulation-enabled performance and intelligence solutions, and medical communications to the biopharma industry, has released the latest version of its flagship quantitative systems toxicology (QST) platform, DILIsym® version X.



Upcoming Events



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DILIsym: By the Numbers...

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Number of biopharmaceutical Sponsors that have engaged DILIsym* Mechanistic projects simulating liver injury based on in vitro data on mechanisms of toxicity Biomarker fitting projects simulating hepatocyte loss consistent with measured transaminase profiles 50+

Known regulatory submissions Marketed drugs in which it has been publicly disclosed that DILIsym contributed to liver safety decisions

8

* Total DILIsym clients, some with multiple projects / compounds



QST Predicts Tox via the Intersection Between Exposure, Mechanisms, and Inter-Patient Variability





Highlights of DILIsym Version X (DSX)

- Completely <u>NEW</u> software platform!
 - Much faster and more user-friendly design
 - Command line and GUI options
 - No reliance on MATLAB base or runtime
 - Server/cloud computing capability (HPGL)
- 4 <u>NEW</u> exemplar compounds included with varying clinical presentations
 - PF-04895162 (Generaux 2019)
 - <u>Efavirenz</u>
 - Anastrozole
 - <u>Tamoxifen</u>
- 2 <u>NEW</u> SimCohorts that include variability in susceptibility to liver injury and biomarker-related parameters (ALT and bilirubin)





Recent DILIsym (and BIOLOGXsym) Publications Showcase Various QST Model Applications and R&D

CLINICAL PHARMACOLOGY & THERAPEUTICS | VOLUME 114 NUMBER 5 | November 2023

Assessing Liver Effects of Cannabidiol and Valproate Alone and in Combination Using Quantitative Systems Toxicology

Vinal V. Lakhani¹, Grant Generaux¹, Brett A. Howell¹, Diane M. Longo¹ and Paul B. Watkins^{2,3,*}

CLINICAL PHARMACOLOGY & THERAPEUTICS | VOLUME 115 NUMBER 3 | March 2024

Quantitative Systems Toxicology Modeling Informed Safe Dose Selection of Emvododstat in Acute Myeloid Leukemia Patients

Kyunghee Yang^{1,*} , Ronald Kong², Robert Spiegel², John D. Baird², Kylie O'Keefe², Brett A. Howell¹

CLINICAL PHARMACOLOGY & THERAPEUTICS | VOLUME 114 NUMBER 5 | November 2023

Quantitative Systems Toxicology Identifies Independent Mechanisms for Hepatotoxicity and Bilirubin Elevations Due to AKR1C3 Inhibitor BAY1128688

Christina Battista^{1,}• [©], Lisl K.M. Shoda¹, Paul B. Watkins², Esther Groettrup-Wolfers³, Antje Rottmann⁴, Marian Raschke⁴ and Grant T. Generaux⁵

Frontiers | Frontiers in Pharmacology

Investigating bile acid-mediated cholestatic drug-induced liver injury using a mechanistic model of multidrug resistance protein 3 (MDR3) inhibition

PUBLISHED 17 January 2023

o: 10.3389/fphar 2022 1085621

James J. Beaudoin¹, Kyunghee Yang¹, Jeffry Adiwidjaja^{1,2}, Guncha Taneja¹¹, Paul B. Watkins², Scott Q. Siler¹, Brett A. Howell¹ and Jeffrey L. Woodhead^{1*}

Int. J. Mol. Sci. 2023, 24, 9692. https://doi.org/10.3390/ijms24119692

The Combination of a Human Biomimetic Liver Microphysiology System with BIOLOGXsym, a Quantitative Systems Toxicology (QST) Modeling Platform for Macromolecules, Provides Mechanistic Understanding of Tocilizumab- and GGF2-Induced Liver Injury

James J. Beaudoin ^{1,†}¹⁰, Lara Clemens ^{1,†}, Mark T. Miedel ², Albert Gough ²¹⁰, Fatima Zaidi ³, Priya Ramamoorthy ³, Kari E. Wong ³, Rangaprasad Sarangarajan ³, Christina Battista ¹⁰, Lisl K. M. Shoda ¹, Scott Q. Siler ¹, D. Lansing Taylor ², Brett A. Howell ¹, Lawrence A. Vernetti ^{2,*}¹⁰ and Kyunghee Yang ^{1,*}

XENOBIOTICA 2024, VOL. 54, NO. 7, 401–410 https://doi.org/10.1080/00498254.2024.2361027

Prediction of the liver safety profile of a first-in-class myeloperoxidase inhibitor using quantitative systems toxicology modeling

Jeffrey L. Woodhead^a, Yeshi Gebremichael^b, Joyce Macwan^a, Irfan A. Qureshi^c, Richard Bertz^c, Victoria Wirtz^c and Brett A. Howell^a



Mechanistic Modeling and *In Vitro* Studies of Drug-induced Liver Injury Suggest a Role for Reduced Biliary Efflux in Tolvaptan-associated Hepatotoxicity



CLINICAL PHARMACOLOGY & THERAPEUTICS | VOLUME 109 NUMBER 2 | February 2021

Quantitative Systems Toxicology Modeling Predicts that Reduced Biliary Efflux Contributes to Tolvaptan Hepatotoxicity

James J. Beaudoin¹¹, William J. Brock²¹, Paul B. Watkins¹, and Kim L. R. Brouwer^{1,4}

First DILIsym publication via academic license



BSEP (Bile Salt Export Pump) NTCP (Sodium-Taurocholate Cotransporting Polypeptide) MRP (Multidrug Resistance–Associated Protein) OATP (Organic Anion-Transporting Polypeptide) P-gp (P-glycoprotein) OSTα/β (Organic Solute Transporter α/β)



QST Modeling of Otenaproxesul Liver Enzyme Elevations Leads to Prediction of Liver Safety for Acute Otenaproxesul Dosing



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<u>NEW! Liver Safety+ Package</u>

- Machine Learning ADMET Predictor Workflow for DILIsym Use Enables Earlier, Higher Throughput Use
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Simulations Plus Has Developed a Roadmap to Derive an Early Assessment of Hepatotoxic Risk

- New DILIsym module in ADMET Predictor 12 generates outputs that can be used to inform inputs within the quantitative systems toxicology (QST) modeling platform DILIsym
 - Permissive of liver safety assessment during early drug discovery efforts!
 - Predictions of the current roadmap are qualitative
 - Yes/no toxicity mechanism classifications
 - Rank ordering of a compound's toxicity assessment with other in-class compounds
 - Accuracy and use of outputs will improve iteratively, as more data become available to inform predictions
- Workflow permissive for early discovery applications
 - No need for data from typical DILIsym in vitro assays
 - Leverages ADMET Predictor informed structure-based compound properties
 - Applies ADMET Predictor Machine Learning from a library of DILI/clean compounds
 - Use of constant liver exposure based on molar concentrations OR use of ADMET Predictor High-Throughput PK (HTPK) results
 - Integration of the above in the DILIsym *in vivo* context for early insights into liver liabilities







The ADMET Predictor 12 DILIsym (APD) Module Adds Liver Safety Insights to Weight of Evidence Informing Compound Selection



- Liver safety liabilities are commonly identified when a compound enters the clinic, sometimes as late as phase 3 clinical trials, imparting considerable expense and potential delays to drug development
- Historical use of DILIsym, a QST model of drug-induced liver injury, required extensive *in vitro* assay data and PK exposure modeling, making it less amenable for use in early drug discovery
- New APD module empowers DILIsym use at the drug discovery stage, without the need for typical DILIsym toxicity assay data!



DILIsym QST Model Predicts Liver Toxicity via the Intersection of Exposure, Mechanisms, and Inter-Patient Variability





Predicted Liver Exposure Interacts with Data-Defined Mechanisms of Toxicity in the DILIsym Simulated *In Vivo* Environment

Mechanisms of toxicity in DILIsym



- Reactive oxygen species (ROS)
- Mitochondrial dysfunction
- Bile acid transporter inhibition
 - Bile salt export pump (BSEP)
 - Multidrug resistance associated protein 3 or 4 (MRP3/MRP4)
 - Sodium-taurocholate cotransporting polypeptide (NTCP)
- Phospholipid transporter inhibition
 - Multidrug resistance protein 3 (MDR3)





Liver Safety+ Prediction Package Tailored for Early Discovery Data





APD Module Applies Machine Learning to Bridge from Compound Structure to DILIsym



APD Module Outputs Include Values for Four Key Mechanisms of Hepatotoxicity

- APD module provides classifications (yes/no) and key parameter values for each of the four main mechanisms of toxicity represented in DILIsym
- Outputs are evaluated for potential toxicity
- If outputs suggest toxicity, user can move to identifying parameter values for DILIsym simulations
- Details on each of the APD module outputs and machine learning model construction are available in the ADMET Predictor 12 Manual

Toxicity Mechanism	APD classification [§] output	APD MEC ⁺ output	APD AC ₅₀ ‡ output	APD IC ₅₀ [∥] output
Mitochondrial dysfunction				-
Reactive oxygen species				-
BSEP inhibition		_	_	
MRP3/MRP4 inhibition		_	_	-
MDR3 inhibition		—	—	

§ yes/no prediction for in vitro signals

+ minimum effective concentration (MEC) that significantly crosses vehicle control threshold

‡ concentration at which 50% maximum effect is observed

|| concentration at which 50% inhibition is observed



Import/Load a Chemical Structure (e.g., SMILES) in ADMET Predictor

4 Canonical SMILES	⑦ 12 Solithromycin.smi 13	
ne Simplified Molecular-Input Line-Entry System (SMILES) is a widely-used line notation for chemical structures. PubCh ompounds: canonical SMILES (computed from chemical structures devoid of isotopic and stereochemical information), ructures containing isotopic and stereochemical information). This section shows the canonical SMILES of the compou ead more at: https://www.daylight.com/dayhtml/doc/theory/theory.smiles.html	$ \begin{array}{c} 1 \\ \text{computes two kinds of SMILES strings for} \\ \text{someric SMILES (computed from chemical} \end{array} \end{array} \\ \begin{array}{c} 1 \\ \text{computes two kinds of SMILES strings for} \\ \text{computed from chemical} \end{array} \\ \begin{array}{c} 1 \\ \text{computed from chemical} \end{array} \\ \begin{array}{c} 2 \\ \text{computed from chemical} \end{array} \\ \end{array} \\ \begin{array}{c} 2 \\ \text{computed from chemical} \end{array} \\ \end{array} \\ \begin{array}{c} 2 \\ \text{computed from chemical} \end{array} \\ \end{array} \\ \begin{array}{c} 2 \\ \text{computed from chemical} \end{array} \\ \end{array} \\ \end{array} \\ \end{array} \\ \end{array} $ \\ \begin{array}{c} 2 \\ \text{computed from chemical} \end{array} \\ \end{array} \\ \end{array} \\ \begin{array}{c} 2 \\ \text{computed from chemical} \end{array} \\ \end{array} \\ \end{array} \\ \end{array} \\ \end{array} \\ \end{array} \\ \end{array} \\ \end{array} \\ \begin{array}{c} 2 \\ \text{computed from chemical} \end{array} \\ \end{array}	C) F) C CCCC
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Generate 3D Structure/Coordinates and Calculate ADMET: DILIsym Properties

Solithromycin - ADMET Predictor							
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Review and Export APD Module Results





Use and Interpretation of APD Module Classifications and Parameter Values

- Model provides yes/no classification predictions for active toxicity mechanisms based on compound structure
 - Mitochondrial dysfunction (solithromycin: yes)
 - ROS production (solithromycin: yes)
 - BSEP inhibition (solithromycin: yes)
 - MRP3 inhibition (solithromycin: yes)
 - MDR3 inhibition (solithromycin: no)
- Within current framework, Simulations Plus recommends prioritizing yes/no classification before utilizing the predicted, quantitative toxicity effects (MEC, AC₅₀, IC₅₀), if available
- The predicted MEC and AC₅₀ values predicted for mitochondrial dysfunction and ROS production can be used for subsequent toxicity parameter estimation in DILIsym
- The predicted BSEP and MDR3 IC₅₀ values can be used directly as DILIsym input parameters



Model-Predicted MEC and AC₅₀ for Mitochondrial Toxicity Can Be Utilized to Derive DILIsym Parameters for Mitochondrial Effects



MEC: minimum effective concentration that significantly crosses vehicle control threshold

.

 AC₅₀: concentration at which 50% maximum effect is observed

* Assume concentration of 0.001 μ M causes no change in OCR compared to control

⁺ Assume MEC causes OCR to drop to 0.8x control for ETC inhibitor

[‡] Assume maximal reduction in OCR is complete inhibition (0x control)



Model-Predicted MEC and AC₅₀ for ROS Mechanism Can Be Utilized to Derive DILIsym Parameters for Effect on ROS Production



- MEC: minimum effective concentration that significantly crosses vehicle control threshold
- AC₅₀: concentration at which 50% maximum effect is observed

* Assume concentration of 0.001 μ M causes no change in ROS compared to control

⁺ Assume MEC causes ROS production to increase 1.2x control

[‡] Assume maximal ROS production response is 5x control



Model-Predicted IC₅₀ Values for BSEP and MDR3 Can Be Utilized Directly as DILIsym Parameters for Bile Acid and Phospholipid Transport Inhibition Effects







Multiple Options for Liver Exposure in DILI Toxicity Ranking Process

- APD module is designed to provide insight into DILI toxicity rankings at any stage in the drug development pipeline
- Based on where a compound is in the drug development pipeline, different information about exposure in humans is available
 - Compounds further along in the pipeline likely have more information available to define exposure
 - Compounds very early on in development may have minimal data to inform exposure



- Potential options for liver exposure to drive hepatotoxicity mechanisms in DILIsym:
 - 1 Constant liver exposure based on molar concentrations [focus in today's Lunch and Learn]
 - > DILIsym simulations to be performed at a range of constant liver concentrations
 - For rank-ordering hepatotoxicity risk of multiple in-class compounds using the "constant liver exposure" approach, liver concentrations need to be normalized using a relevant metric which provides consideration to compound-specific efficacy ranges
 - 2 Assume or estimate liver profiles from preclinical PK data
 - 3 Estimate liver exposure from ADMET Predictor HTPK using predicted C_{max} and liver partition coefficient from user-specified doses
 - Predict liver exposure from GastroPlus PBPK model





Setting up the Drug Parameter Set in DILIsym and Defining Constant Liver Concentrations Using the Specified Data Feature

	Create New DILlsym SimSingle Configurat	ion ? X			
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APD Module Predictions Are Used to Set Up Active Toxicity Mechanisms in DILIsym

Machine Learning Algorithms DILIsym Mitochondrial dysfunction . Oxidative stress ٠ Bile acid efflux transporter inhibition • DILI Mechanism Selector for Solithromycin (Solithromycin_1nM) Phospholipid transporter inhibition ٠ Select Molecul ¥ CompY **ADMET Predictor 12** Customized Variable Filter By Name **DILIsym module** Iolecule / Mechanism Value CompY Mech inhBAtransport Compound Y NTCP inhibition constant 1.000000e+10 umol/L Compound Y NTCP alpha constant for inhibition 1.000000e+10 dimensionless Compound Y NTCP switch 1.000000e+00 dimensionless Compound Y BSEP inhibition constant 8.86 umol/L Compound Y BSEP alpha constant for inhibition dimensionless Compound Y BSEP switch dimensionless Compound Y basolateral inhibition constant 1.000000e+10 umol/I Novel Compound Y basolateral alpha constant for inhibition 1.000000e+10 dimensionless Compound Y basolateral switch 1.000000e+00 dimensionless CompY_Mech_inhETC3 Compound Coefficient for ETC inhibition 3 0.040746 umol/L (Solithromycin) Max inhibitory effect for ETC inhibition 3 0.39355 dimensionless CompY Mech inhETC1 Coefficient for ETC inhibition 1 2379.481 umol/L CompY Mech incRNSROSproduction4 Liver RNS-ROS production rate Vmax 4 5.8195 1/hour Liver RNS-ROS production rate Km 4 9.1224 umol/L Liver RNS-ROS production rate Hill 4 4.5496 dimensionless CompY Mech incRNSROSproduction1 Liver RNS-ROS production rate constant 1 0.053744 mL/nmol/hour Identifier Geometry 3D Quality AP_FWeight BSEP Inh BSEP IC50 MDR3 IC50 MDR3 Inh Mito AC50 Mito MEC Mito Tox MRP3 Inh ROS AC50 ROS MEC ROS Tox Structure Save with Custom Save As New

S+ SimulationsPlus

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APD Module Predictions Are Used to Set Up Active Toxicity Mechanisms in DILIsym

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ي ت	Solithromycin_1000nM	•••	Human ROS apop mito BA v8A 1 Multi16 A Create Import Delete	CDCA amidation Vmax (nmol/hour) CDCA basolateral transport Vmax (nmol/h
-+ +	Solithromycin_5000nM		Select Output Panel	CDCA canalicular transport Vmax (nmol/h
%	Solithromycin_pt01nM		5 [↑] ¢	CDCA uptake Vmax (nmol/hour)
	Solithromycin_pt05nM		Common Outputs	CDCA-amide basolateral transport Vmax (CDCA-amide canalicular transport Vmax (r
	Solithromycin_pt1nM		CPU Thread Count Selector	CDCA-amide gut uptake Vmax (nmol/hour CDCA-amide uptake Vmax (nmol/hour)
	Solithromycin_pt5nM	***	Min Low Normal High MAX	Canalicular transporter regulation expone Caspase-mediated apoptosis scaling cons
\otimes	Create		Load Initial Conditions for SimPops	GSH basal level (mmol/L)

Liver concentration (fold change from IC_{50})*	Liver concentration (nM)	Maximum ALT (U/L)
4.3e-7	0.01	30
2.2e-6	0.05	30
4.3e-6	0.1	30
2.2e-5	0.5	30
4.3e-5	1	30
2.2e-4	5	30
4.3e-4	10	30
2.2e-3	50	35
4.3e-3	100	112
2.2e-2	500	2999
4.3e-2	1000	9510
2.2e-1	5000	6114

* For the compounds tested in this class of compounds (macrolide antibiotics), IC₅₀ values for OATP1B1 were measured consistently for all compounds; IC₅₀ used as normalization metric in this example



Interpretation of Toxicity Ranking Results

For drugs early on in development pipeline (using constant liver exposure method)



For drugs further along in development pipeline (using known liver concentrations or predicted using ADMET Predictor HTPK module or PBPK model

3xULN

5

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APD Module Outputs Reproduce Clinical and Previous DILIsym Simulation Toxicity Rankings: <u>Macrolide Antibiotics</u>

ML Tox Model Predictions

Clinical Data & Previous DILIsym Simulation Results

the Literature (3, 10, 31)

Compound

Solithromyain

Clarithromycin

Erythromyain

Telithromycin

Azithromycin

arm Res (2019) 36: 48

RESEARCH PAPER

ttps://doi.org/10.1007/s11095-019-2582-

Protocol

Oral (CE01-300)

IV-to-Oral (CE01-301)

500 mg BID 7 days

800 mg OD 10 days

250 mg OD days 2-5

Analyzing the Mechanisms Behind Macrolide Antibiotic-Induced

500 mg OD day I

500 mg

Upper limit of normal (ULN) in DILlsym is 40 U/L ^a (9); 2.8% among patients with normal baseline ALT ^b (8): 6.6% among patients with normal baseline ALT

OID 10 days

Table III Results in the v4A_I SimPops for Each of the Five Macrolides in DILlsym v5A Compared to Reported Clinical data. Observed Data are from

Peak ALT >3X ULN

Observed

5.4%^a

9.1%^b

1 - 2%

1-2%

~0.5%

1.2%

(38/417)

(22/411)

Simulated

3.9%

6.0%

2.8%

(8/285)

2.8%

(8/285)

0%

0%

CrossMa

(11/285)

(17/285)



Liver concentrations were normalized to OATP1B1 IC₅₀ values for macrolide antibiotics



APD Module Outputs Reproduce Clinical and Previous DILIsym Simulation Toxicity Rankings: CGRP Receptor Antagonists

ML Tox Model Predictions

Clinical Data & Previous DILIsym Simulation Results



Liver concentration were normalized to CGRP receptor Ki values for CGRP receptor antagonists



1.9% (5/263)

3.2% (8/265)

1.9% (5/263)

3.2% (8/265)

Workflow Summary: APD Module Enables Efficient Assessment of Hepatotoxic Rankings for In-Class Compounds at Any Stage of Drug Development!



How to Engage with SLP, Here and Elsewhere?

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5-6:30PM!