High-throughput prediction of fraction absorbed and bioavailability in silico

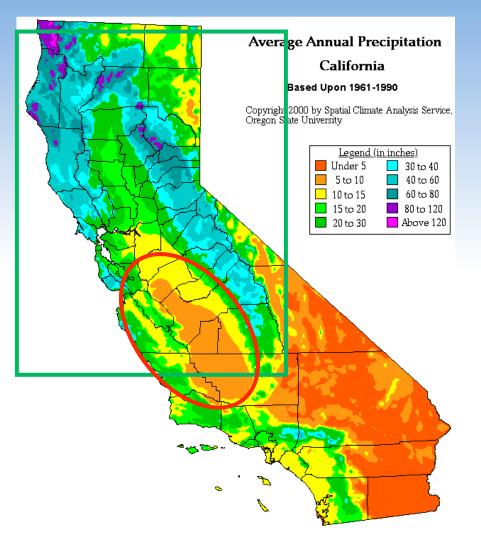
Robert D. Clark, Pankaj Daga and Robert Fraczkiewicz Simulations Plus, Inc. Lancaster CA, USA



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Motivation

- Late-stage attrition due to obviously bad physicochemical properties has been reduced by application of rules-ofthumb like Lipinski's Rule of Five.
- Failure due to lack of efficacy remains a major issue. This can be due to a poor choice of target, but can also reflect poor pharmacokinetics (PK).
- Combinations of physicochemical properties may be problematic even when each individual property seems acceptable.
- This can make it hard to know which property to modify next, e.g., decreasing hydrophobicity to try to address problems with solubility and metabolism is likely to decrease absorption as well.
- Is there a way to explore the pharmacokinetic landscape before enough compound has been synthesized to assess its PK behavior in vivo?



http://www.geosci.sfsu.edu/Geosciences

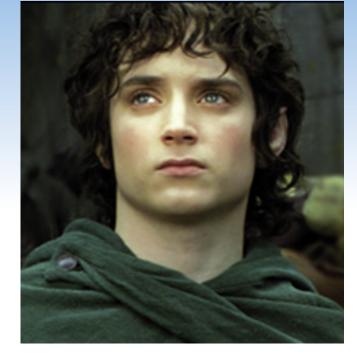


Many Key Properties Can be Estimated Individually in silico

- Acid/base dissociation constants (pK_a's) & ionization
- Solubility and tendency to supersaturate
- Lipophilicity
- Intestinal permeability
 - quantitative: passive permeability & local models
 - qualitative: susceptibility to active uptake or efflux on transporters
- Metabolism
 - quantitative: cytochrome P450 (CYP) clearance & local models
 - qualitative: glucuronidation & glutathione conjugation
- Binding to plasma proteins, to red blood cells & within tissues

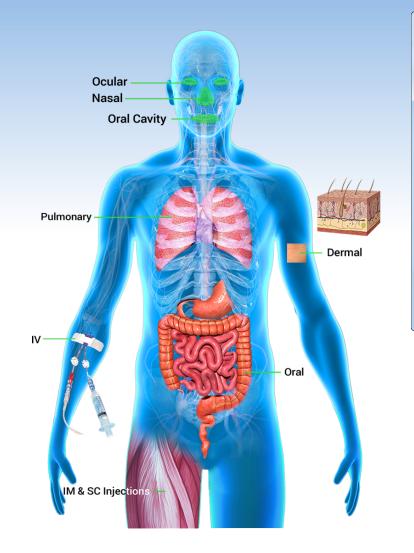
> It takes Physiologically-Based Pharmacokinetic (PBPK) simulation to bind them all.





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Full-blown PBPK Simulation Takes *Everything* **Into Account**

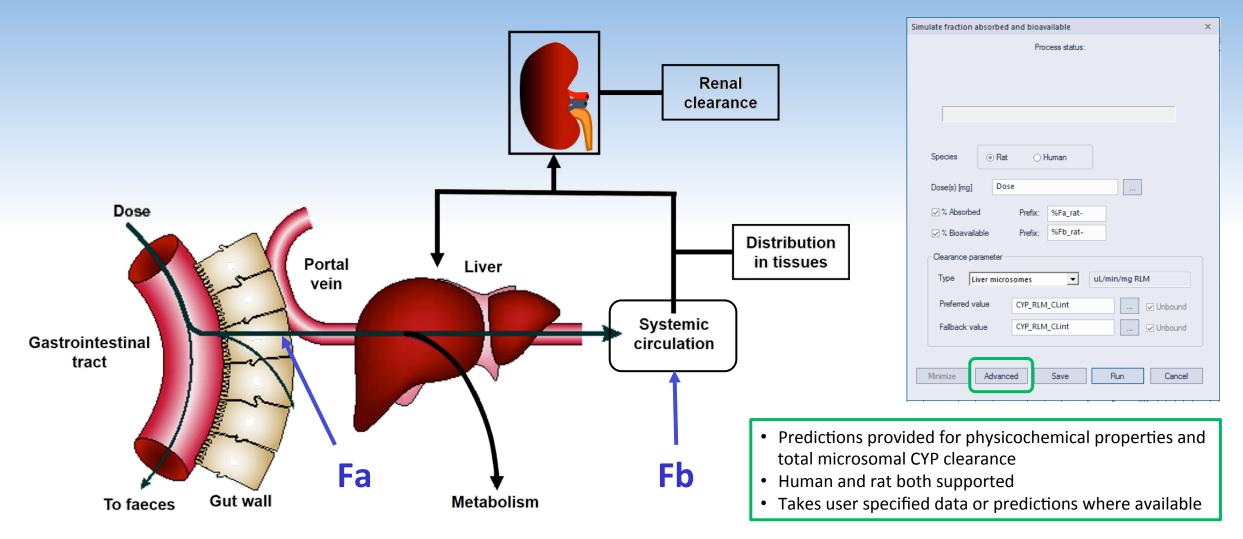


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		Compartment Data									Enzyme and Transporter	Regional Distributions		
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Molecular Weig		Duodenum	0	15.20	5.89	0.19	0.578	9.50	0.22	3.363	20.00			
logP (neutral):		Jejunum 1	0	14.22	6.13	0.83	2.494	45.00	0.21	2.982	17.29			
рКа		Jejunum 2	0	11.78	6.13	0.75	2.262	45.00	0.20	2.353	6.980			
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		lleum 2	0	11.16	5.93	0.01	0.041	1.00	0.18	2.018	1.300			
Transpo		lleum 3	0	11.73	5.93	0.01	0.036	1.00	0.17	2.004	1.240			
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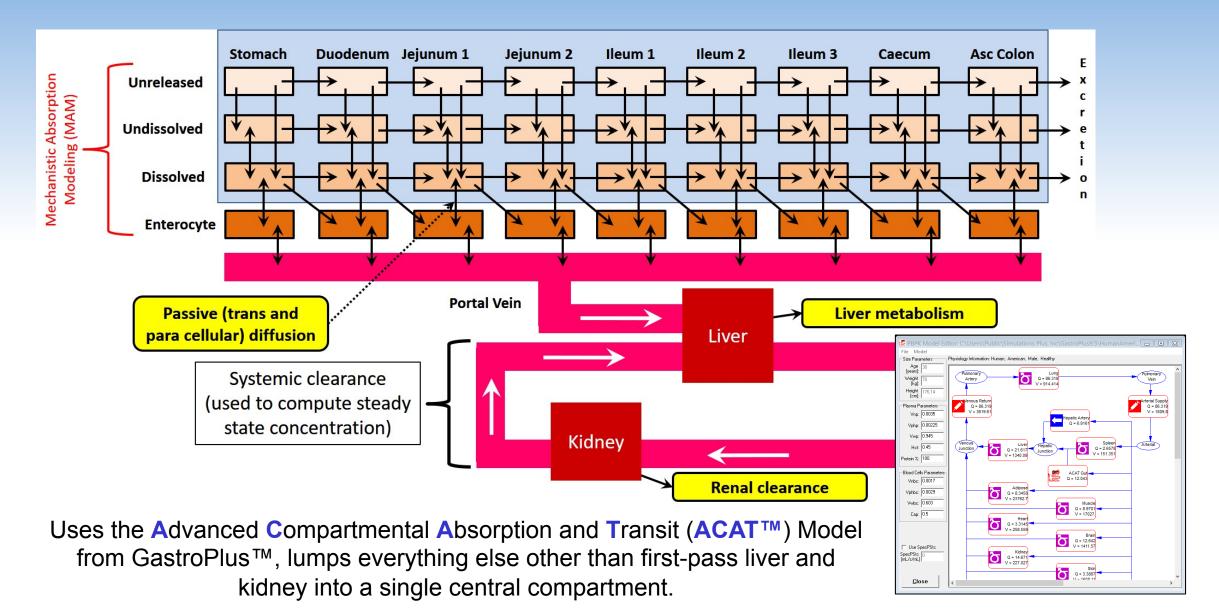
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The HTPK Simulation Module in ADMET PredictorTM is Simpler...



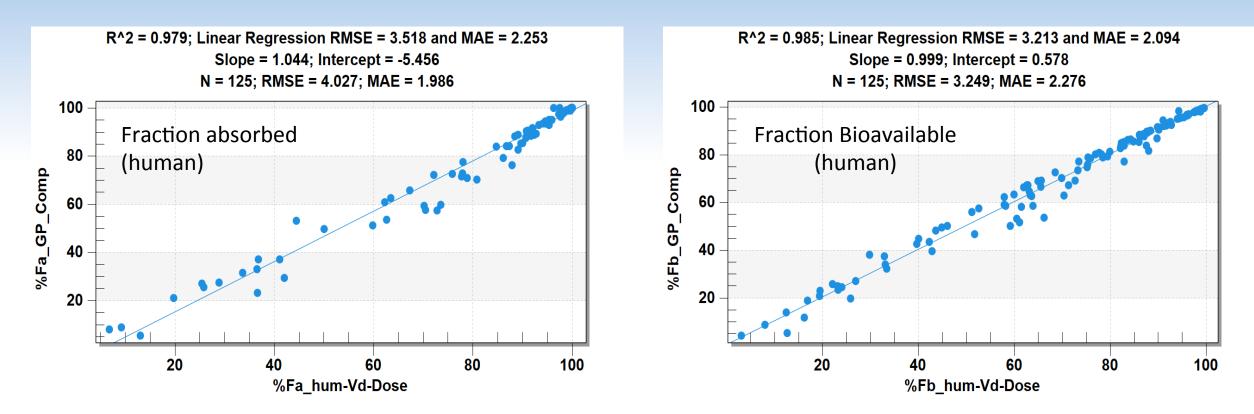
"A thing should be made as simple as possible, but no simpler." - Albert Einstein St Simulations Plus SCIENCE + SOFTWARE = SUCCESS

...But Is Complicated Enough to Get the Job Done



The Streamlined System Produces Similar Results*

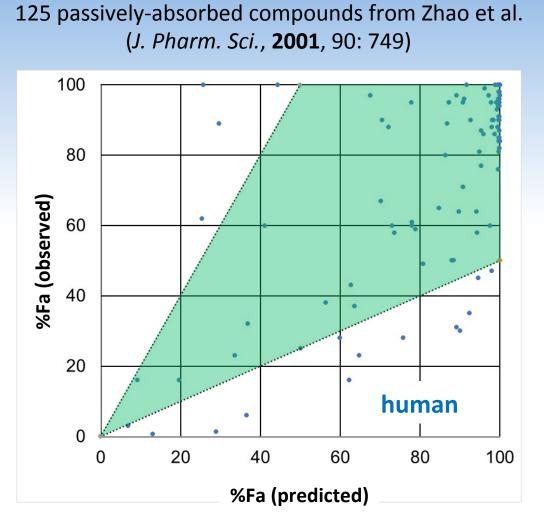
125 passively absorbed compounds from Zhao et al., J. Pharm. Sci. 2001, 90:749



*GastroPlus was run using a compartmental model with ACAT absorption.

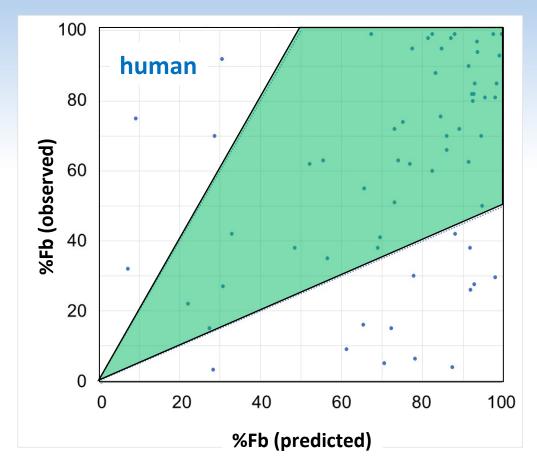
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Does HTPK Give Good Enough Answers?



86% predicted within 2-fold of the reported value.80% predicted within 1.5-fold.

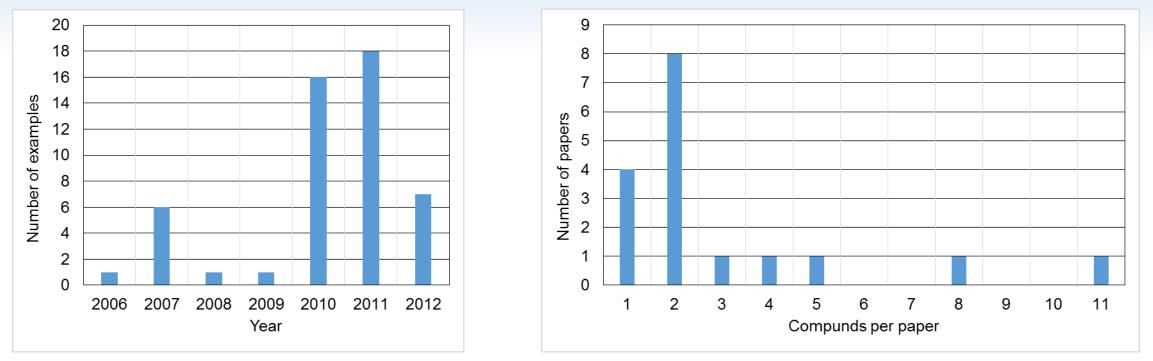
62 CYP-metabolized compounds from the Toshimoto et al. (*Drug Metab. Dispos.* **2014**, 42:1811)



73% predicted within 2-fold of the reported value. 65% predicted within 1.5-fold

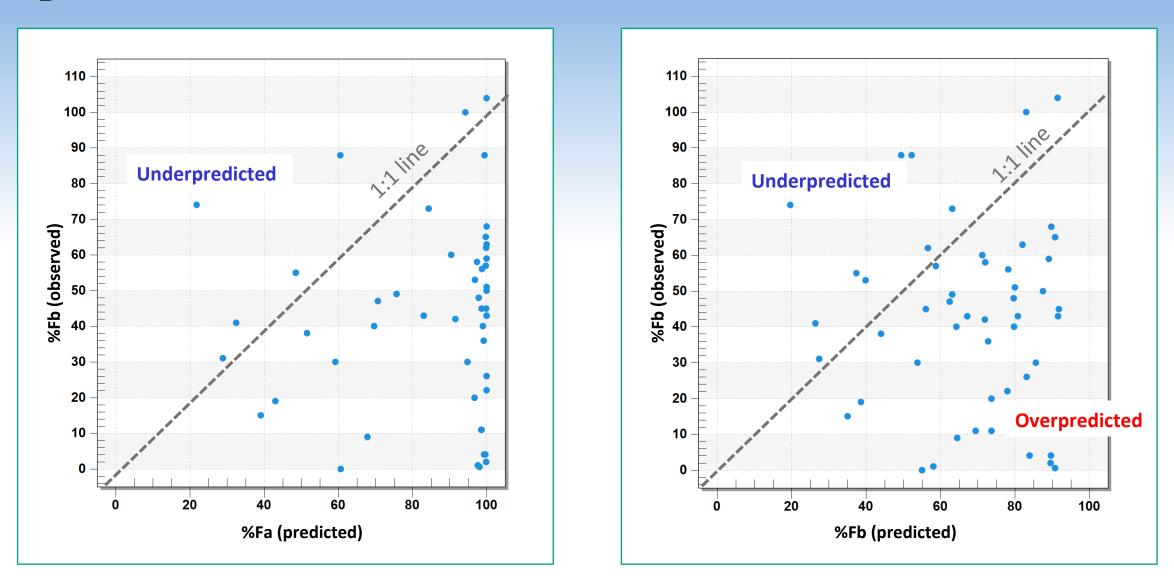
Rat Bioavailability Data Set 1

- Data taken from drug design and development papers published in the *Journal of Medicinal Chemistry* between 2006 and 2012 that report oral bioavailability in rats for one or more analogs
 - rat bioavailabilities for 51 compounds with many different targets
 - 1-11 examples from 17 papers
 - many also include other experimental PK data and results from in vitro assays

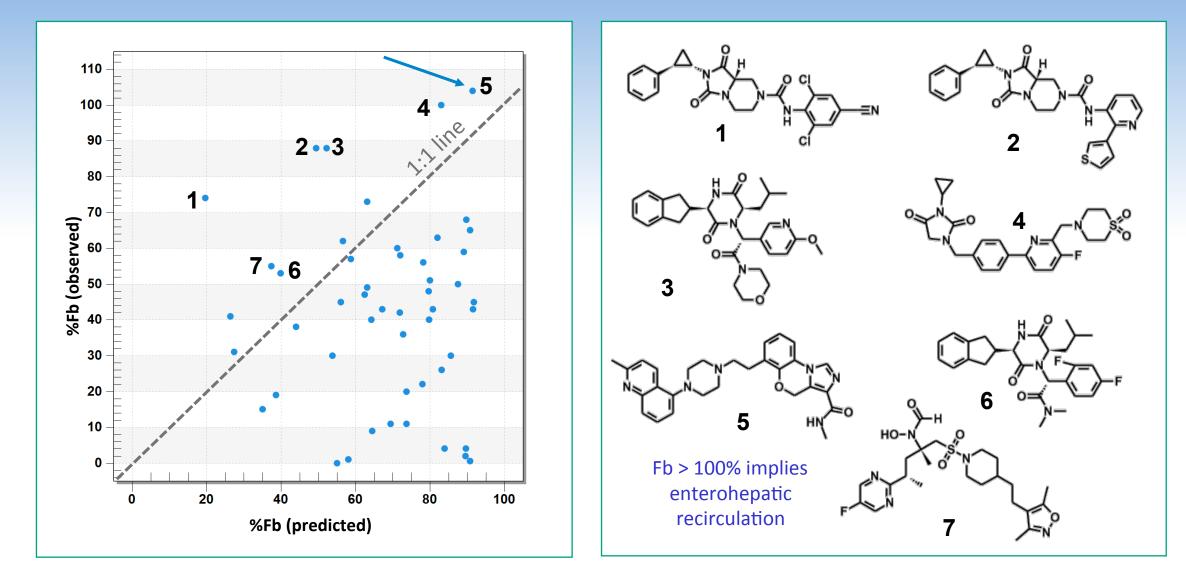




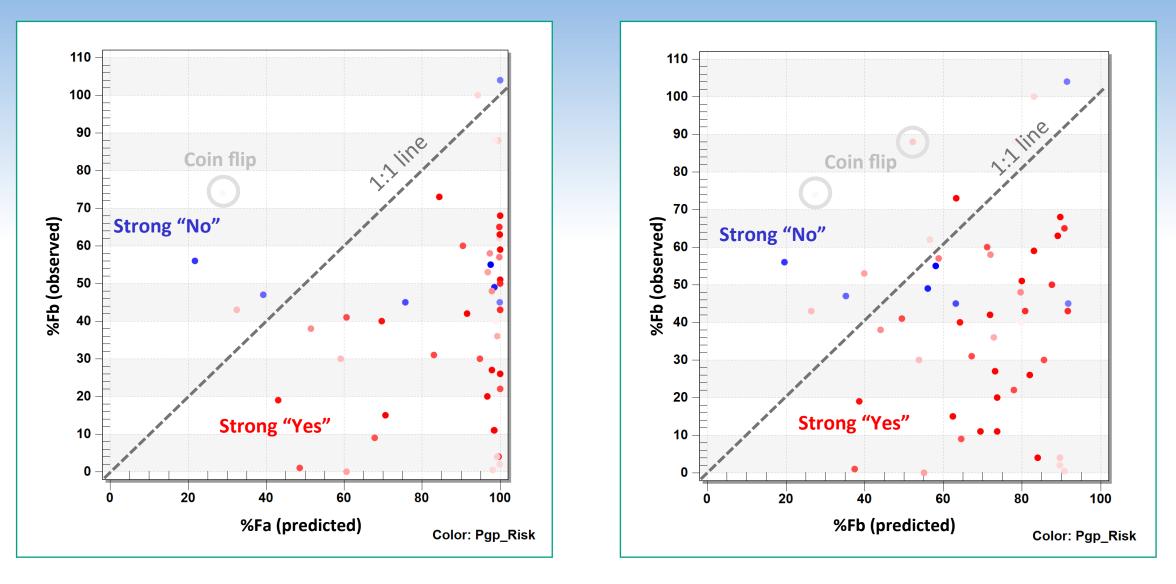
Experimental Fa and Fb vs. in silico Predictions



Why Are Some Bioavailabilities Underpredicted?

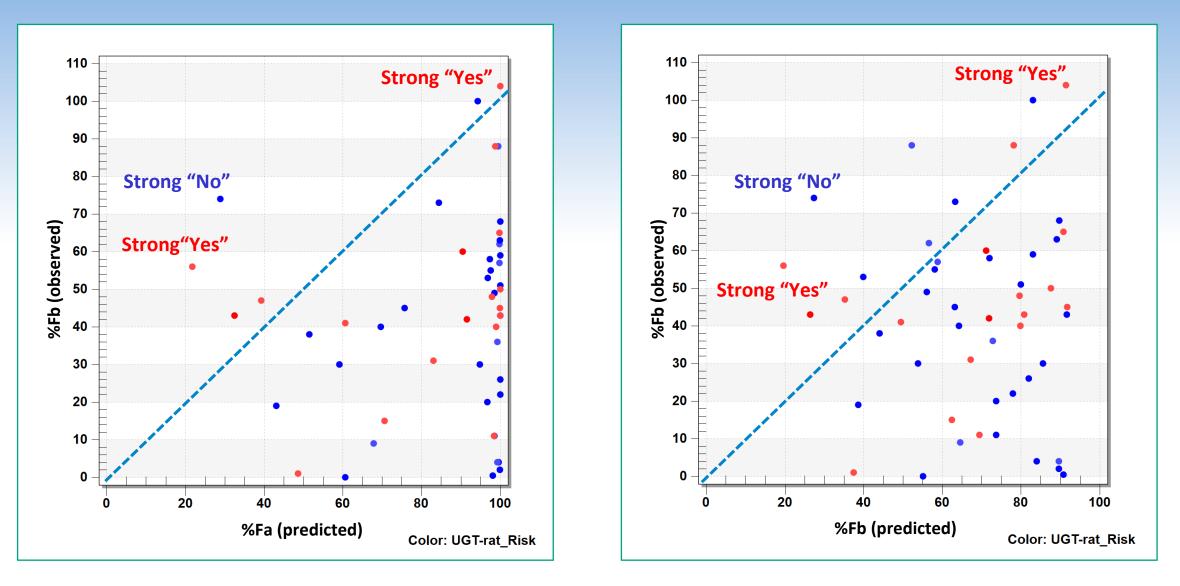


Effect of Being a P-gp Substrate* on Predicted Rat PK



*Based on confidence in human P-gp substrate classification

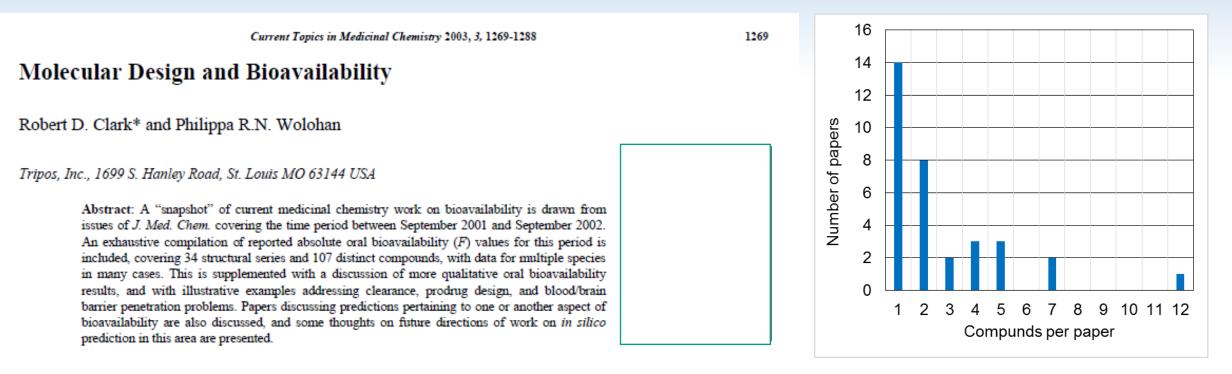
Effect of Being a UGT Substrate* on Predicted Rat PK



*Based on expression levels of rat orthologs of human UGTs and hUGT substrate classifications

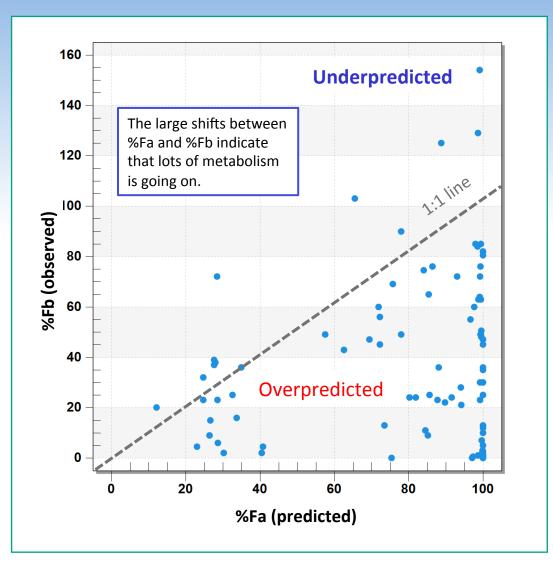
Rat Bioavailability Data Set 2

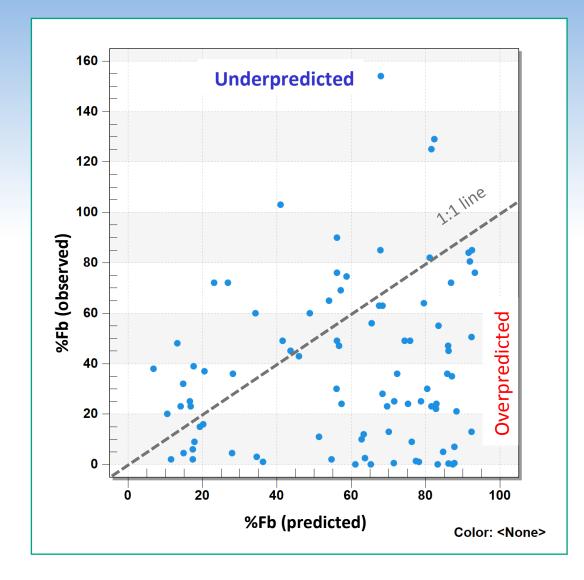
- Data taken from a survey of all bioavailability enhancement studies published in the Journal of Medicinal Chemistry between September 2001 and September 2002
 - rat bioavailabilities for 89 compounds with 20 different targets
 - 1-12 examples from 40 papers



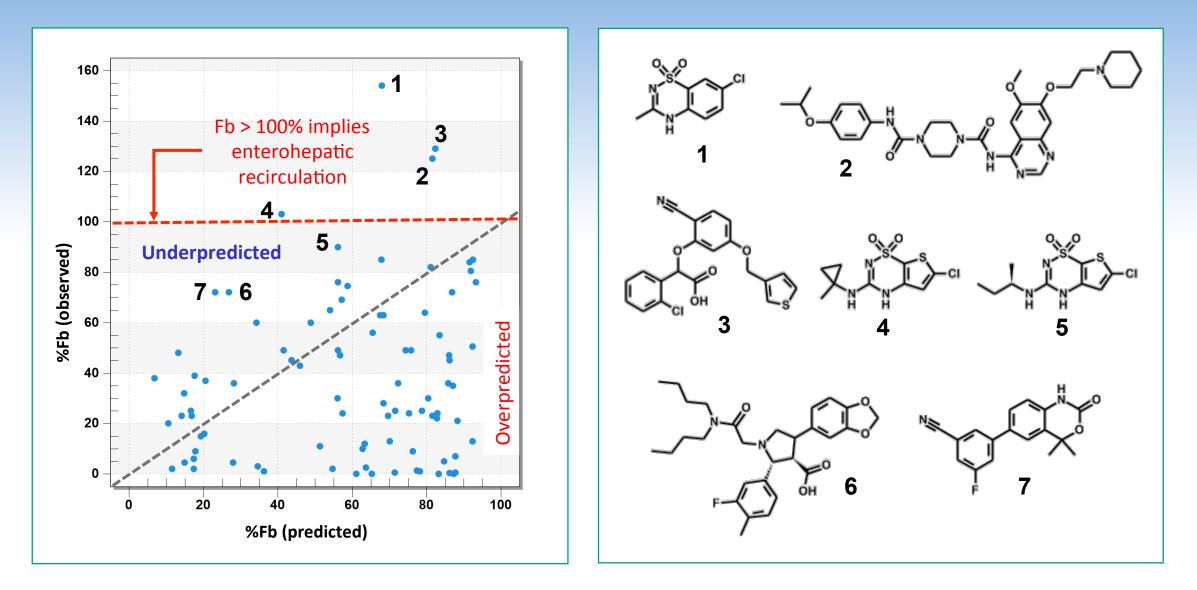


Experimental Fa and Fb vs. in silico Predictions

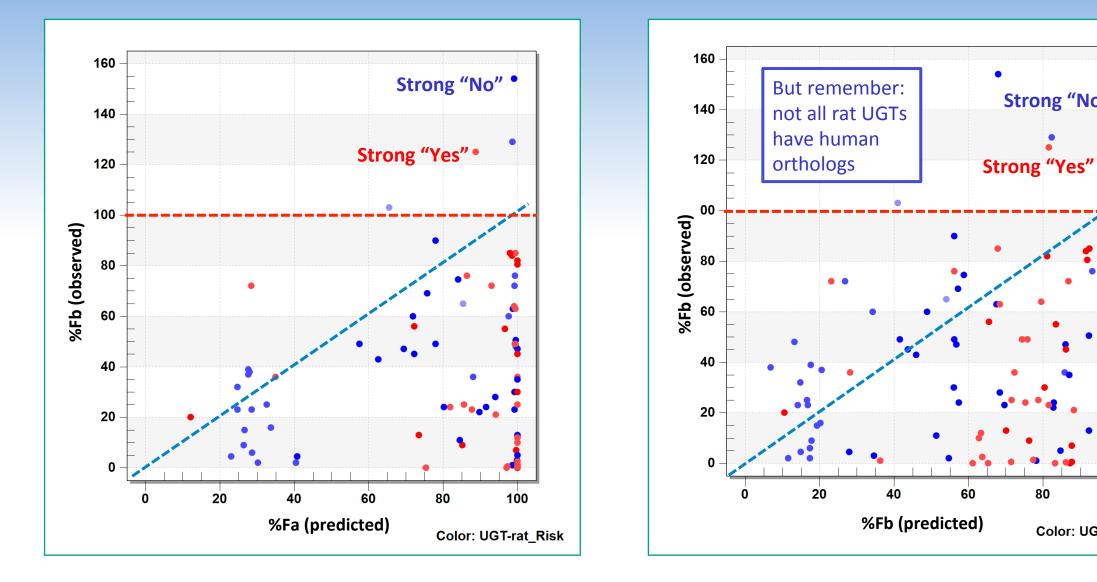




Why Are Some Bioavailabilities Underpredicted?



Effect of Being a UGT Substrate* on Predicted Rat PK



*Based on expression levels of rat orthologs of human UGTs human UGTs and hUGT substrate classifications

Strong "No"

80

100

Color: UGT-rat_Risk

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Article

¹ Physiologically Based Pharmacokinetic Modeling in Lead ² Optimization. 2. Rational Bioavailability Design by Global Sensitivity ³ Analysis To Identify Properties Affecting Bioavailability

^₄ Pankaj R. Daga,^{†,‡}[©] Michael B. Bolger,[§] Ian S. Haworth,[∥] Robert D. Clark,[§] and Eric J. Martin^{*,†}[©]

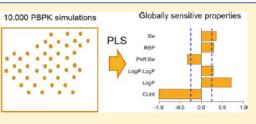
s [†]Novartis Institute of Biomedical Research, Emeryville, California 94608, United States

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⁷ Department of Pharmacology and Pharmaceutical Sciences, University of Southern California, Los Angeles, California 90089, United
 ⁸ States

9 Supporting Information

ABSTRACT: When medicinal chemists need to improve oral bioavailability (%F) during lead optimization, they systemati-11 cally modify compound properties mainly based on their own 12 experience and general rules of thumb. However, at least a 13 dozen properties can influence %F, and the difficulty of 14 multiparameter optimization for such complex nonlinear 15 processes grows combinatorially with the number of variables. 16 Furthermore, strategies can be in conflict. For example, adding 17 a polar or charged group will generally increase solubility but 18 decrease permeability. Identifying the 2 or 3 properties that 19



most influence %F for a given compound series would make %F optimization much more efficient. We previously reported an 20 adaptation of physiologically based pharmacokinetic (PBPK) simulations to predict %F for a lead series from purely 21 computational inputs within a 2-fold average error. Here, we run thousands of such simulations to generate a comprehensive 22 "bioavailability landscape" for the series. A key innovation was recognition that the large and variable number of pK,'s in drug 23 molecules could be replaced by just the two straddling the isoelectric point. Another was use of the ZINC database to cull out 24 chemically inaccessible regions of property space. A quadratic partial least squares regression (PLS) accurately fits a continuous 25 surface to these thousands of bioavailability predictions. The PLS coefficients indicate the globally sensitive compound 26 properties. The PLS surface also displays the %F landscape in these sensitive properties locally around compounds of particular 27 interest. Finally, being quick to calculate, the PLS equation can be combined with models for activity and other properties for 28 29 multiobjective lead optimization.

30 KEYWORDS: PBPK, lead optimization, lead series, local model, global sensitivity analysis, PLS, multiobjective optimization

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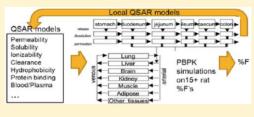
Physiologically Based Pharmacokinetic Modeling in Lead Optimization. 1. Evaluation and Adaptation of GastroPlus To Predict Bioavailability of Medchem Series

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- ⁷ Department of Pharmacology and Pharmaceutical Sciences, University of Southern California, Los Angeles, California 90089, United
 ⁸ States
- 9 Supporting Information

ABSTRACT: When medicinal chemists need to improve

- bioavailability (%F) within a chemical series during lead
- 12 optimization, they synthesize new series members with
- 3 systematically modified properties mainly by following
- experience and general rules of thumb. More quantitative
- 15 models that predict %F of proposed compounds from
- 16 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
- 22 average error needed to guide lead optimization. In this work, local GastroPlus PBPK models are instead customized for
- 23 individual medchem series. The key innovation was building a local QSPR for a numerically fitted effective intrinsic clearance
- 24 (CL_{loc}). All inputs are subsequently computed from structure alone, so the models can be applied in advance of synthesis.
- 25 Training CL_{loc} 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
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Conclusions

- The high-throughput implementation of PBPK simulation in ADMET Predictor yields results in good agreement with analogous analyses in GastroPlus.
- HTPK simulations run using purely *in silico* property estimates are in reasonable but imperfect agreement with experimental results in humans and rats.
- Most experimental bioavailabilities for lead-type compounds in rat fall near or below the predicted fraction absorbed, which suggests that %Fa estimates are accurate.
- Increasing the accuracy of bioavailability estimation using *in silico* property estimates will require quantitative accounting for transporters and non-CYP metabolism.
- Measuring absorption and bioavailability is hard, as is modeling them. Validating the models designed to estimate them may be even harder.



My thanks to:

- My co-authors
- David Miller
- Michael Lawless
- Eric Martin & Ben Madej (Novartis)
- Mike Bolger
- Viera Lukacova
- John DiBella
- Karen Webster

...and to you for your kind attention.

bob@simulations-plus.com

