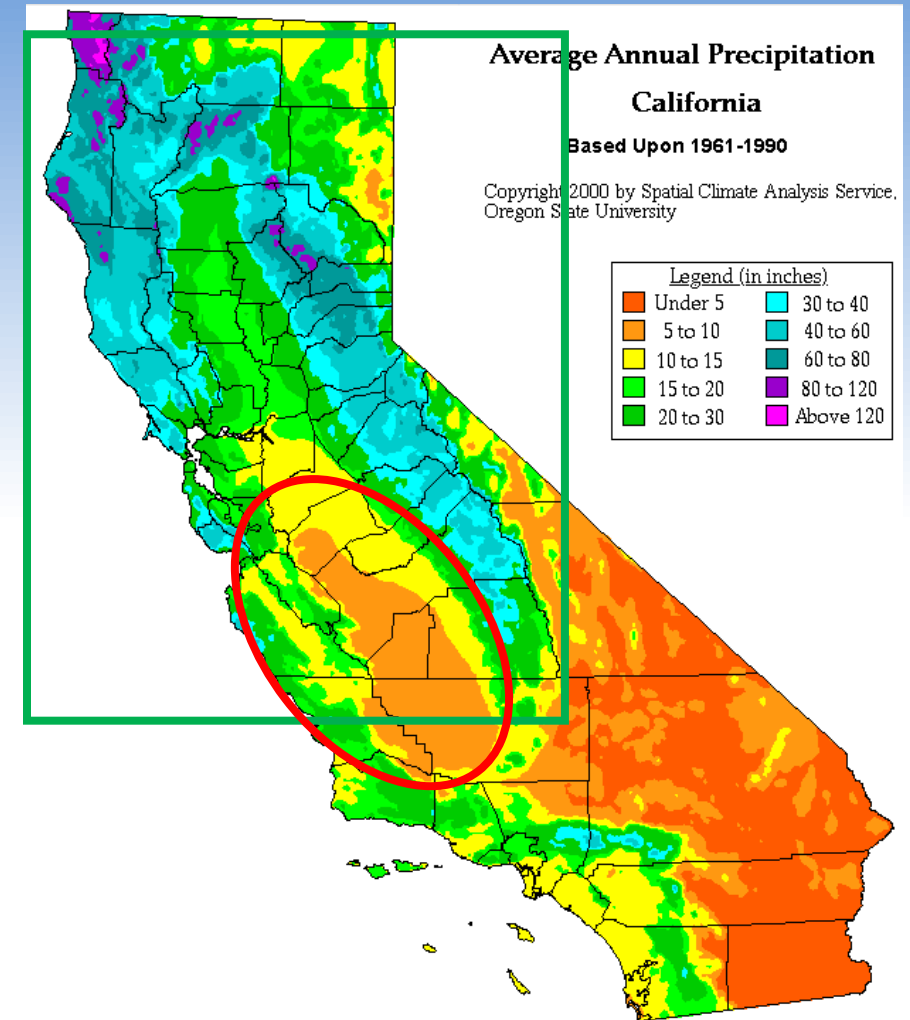


Exploring Pharmacokinetic SARs Early in Drug Discovery

Robert D. Clark
Simulations Plus, Inc.
Lancaster CA, USA

Motivation

- Late-stage attrition due to obviously bad physicochemical properties has been reduced by application of rules-of-thumb 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?*

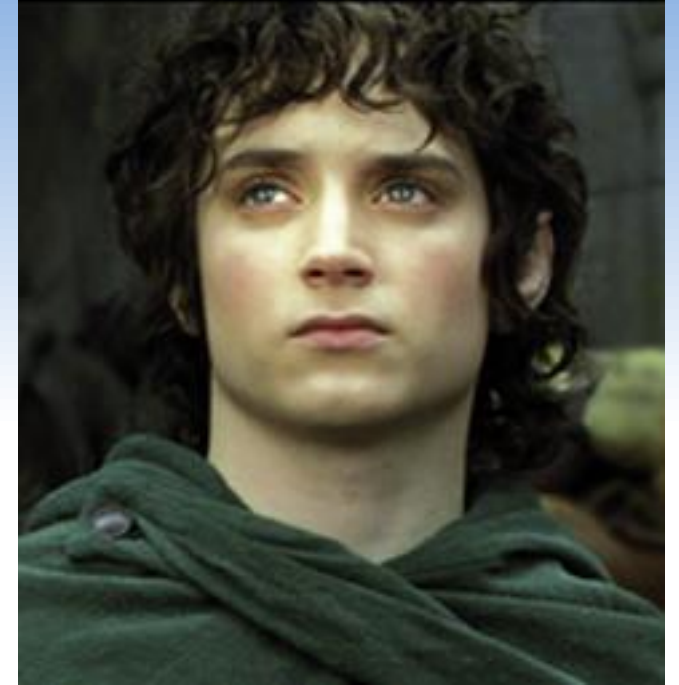


<http://www.geosci.sfsu.edu/Geosciences>

Many Key Properties Can be Estimated Individually *in silico*

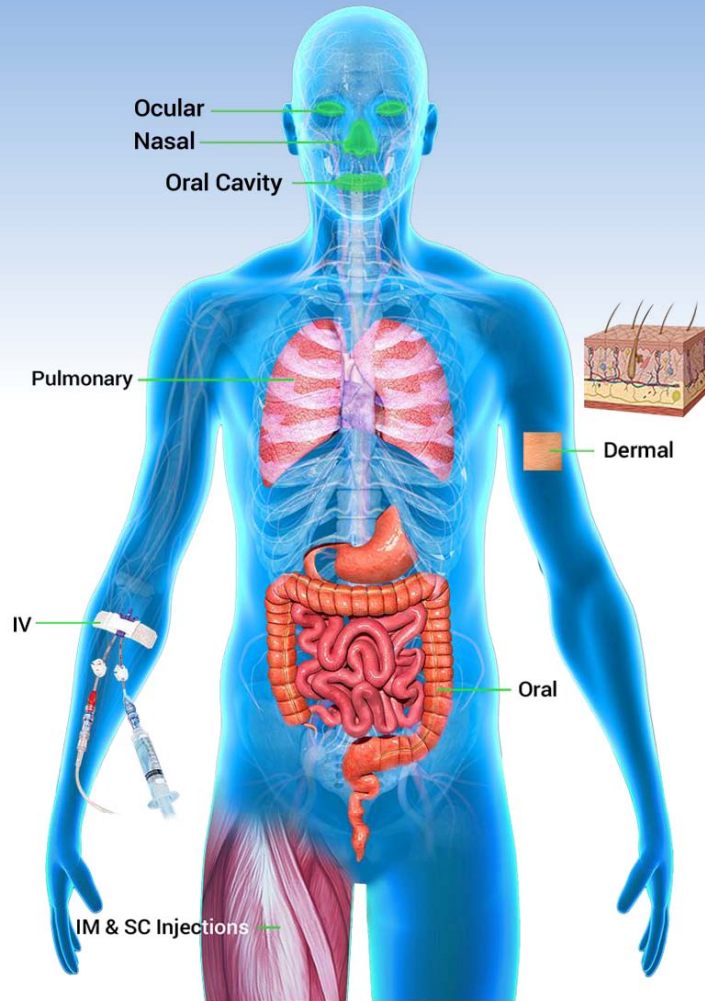
- Acid/base dissociation constants (pK_a 's) & ionization
- Solubility and tendency to supersaturate
- Lipophilicity
- Intestinal permeability
 - quantitative: passive permeability & local models
 - qualitative: susceptibility to active uptake or efflux on transporters
- Metabolism
 - quantitative: cytochrome 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.*



© 2010, The [Middle-Earth Encyclopedia](#)

Full-blown PBPK Simulation Takes *Everything* Into Account



GastroPlus(TM): Bioavails-rat-50.mdb (C:\Users\Bob\Desktop\9.x\SIMz -\Rat e.\)

File Edit Database Simulation Setup Controlled Release Tools Modules (Optional) Help

Compound CPD-651-1
Current=100, Total=100

Molecular Formula
Molecular Weight
logP (neutral)
pKa
Enzyme
Transporter

Experimental or user defined values used for properties:
All remaining properties are predictions from ADMET Predictor v8.1.0.0
Tendency Supersaturate=SupSat; Likelihood of BBB Penetration=High; Pgp-Inhibitor=Yes (97%); Pgp-Substrate=Yes (91%); OATP1B1-Inhibitor=No (53%);

Experimental or user defined values used for properties:
All remaining properties are predictions from ADMET Predictor v8.1.0.0
Tendency Supersaturate=SupSat; Likelihood of BBB Penetration=High; Pgp-Inhibitor=Yes (97%); Pgp-Substrate=Yes (91%); OATP1B1-Inhibitor=No (53%);

Compartmental Parameters

CPD-651-1

Reset All Values

Excrete all un-absorbed drug at the end of gut transit time
Zero-order gastric emptying

Compartment Data										Enzyme and Transporter Regional Distributions	
Compartment	Peff	ASF	pH	Transit Time (h)	Volume (mL)	Length (cm)	Radius (cm)	SEF	Bile Salt (mM)		
Stomach	0	0.0	3.90	0.25	3.360	1.07	1.00	1.000	0.0		
Duodenum	0	15.20	5.89	0.19	0.578	9.50	0.22	3.363	20.00		
Jejunum 1	0	14.22	6.13	0.83	2.494	45.00	0.21	2.982	17.29		
Jejunum 2	0	11.78	6.13	0.75	2.262	45.00	0.20	2.353	6.980		
Ileum 1	0	10.64	5.93	0.02	0.045	1.00	0.19	2.032	2.820		
Ileum 2	0	11.16	5.93	0.01	0.041	1.00	0.18	2.018	1.300		
Ileum 3	0	11.73	5.93	0.01	0.036	1.00	0.17	2.004	1.240		
Caecum	0	10.38	6.58	4.29	0.884	5.00	0.75	1.550	0.0		
Asc Colon	0	18.70	6.23	7.71	0.308	9.00	0.33	1.660	0.0		

C1-C4: 0.06944 0.43028 0.12147 0.46632

Physiology: Rat - Physiological - Fasted

ASF Model: Opt logD Model SA/V 6.1

Qh (L/min): 0.0118

Percent Fluid in SI: 40

Colon: 10

Experimental or user defined values used for properties:
All remaining properties are predictions from ADMET Predictor v8.1.0.0
Tendency Supersaturate=SupSat; Likelihood of BBB Penetration=High; Pgp-Inhibitor=Yes (97%); Pgp-Substrate=Yes (91%); OATP1B1-Inhibitor=No (53%);

pKa Table | logD: Struct-6.1

Diss Model: Johnson

PartSize-Sol: ON

BileSalt-Sol: ON | Diff: ON

ConstRad: OFF

Precip: Time

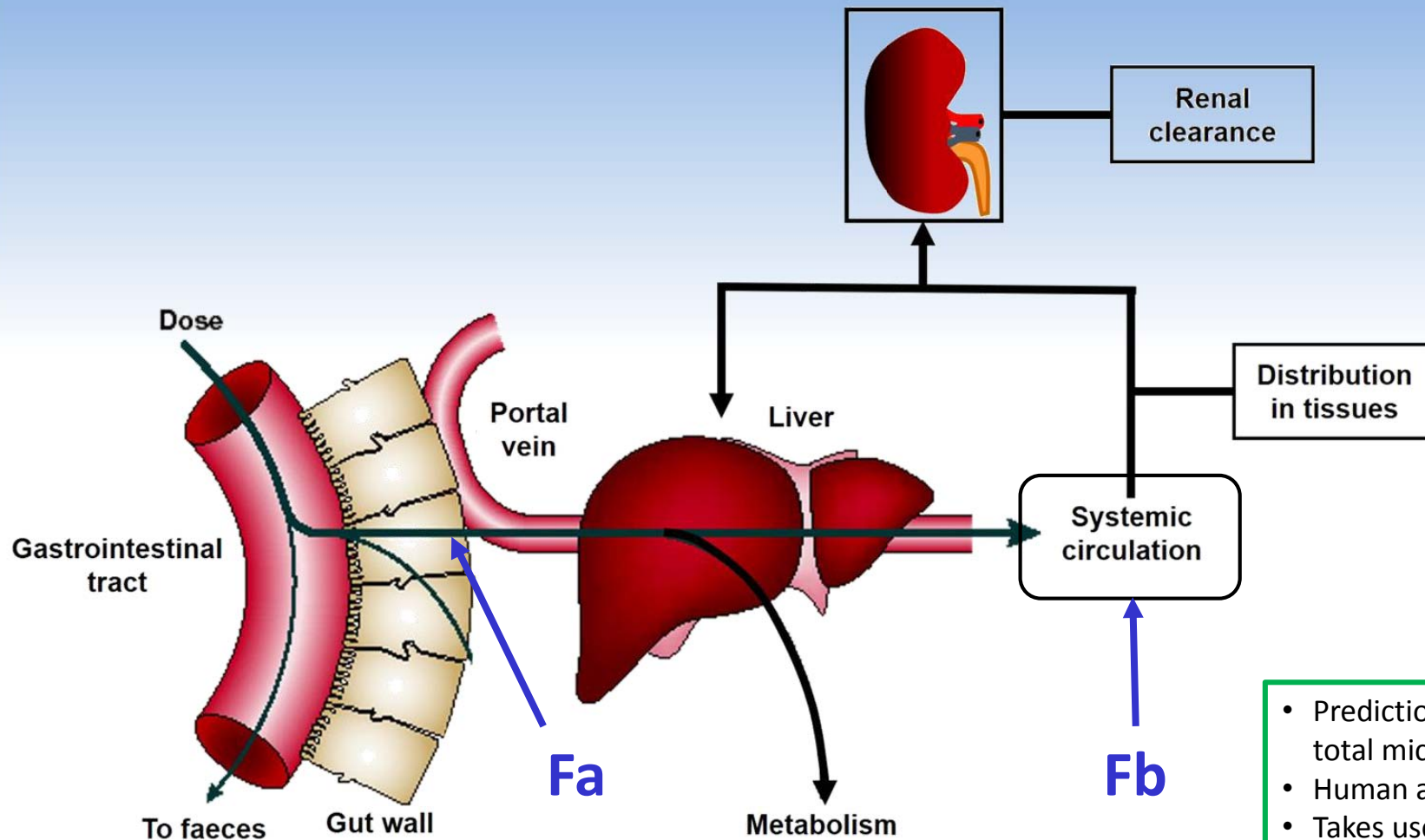
Ppara: Zhim

EHC: OFF

Graph

20 21 22 23 24

The HTPK Simulation Is Much Simpler...

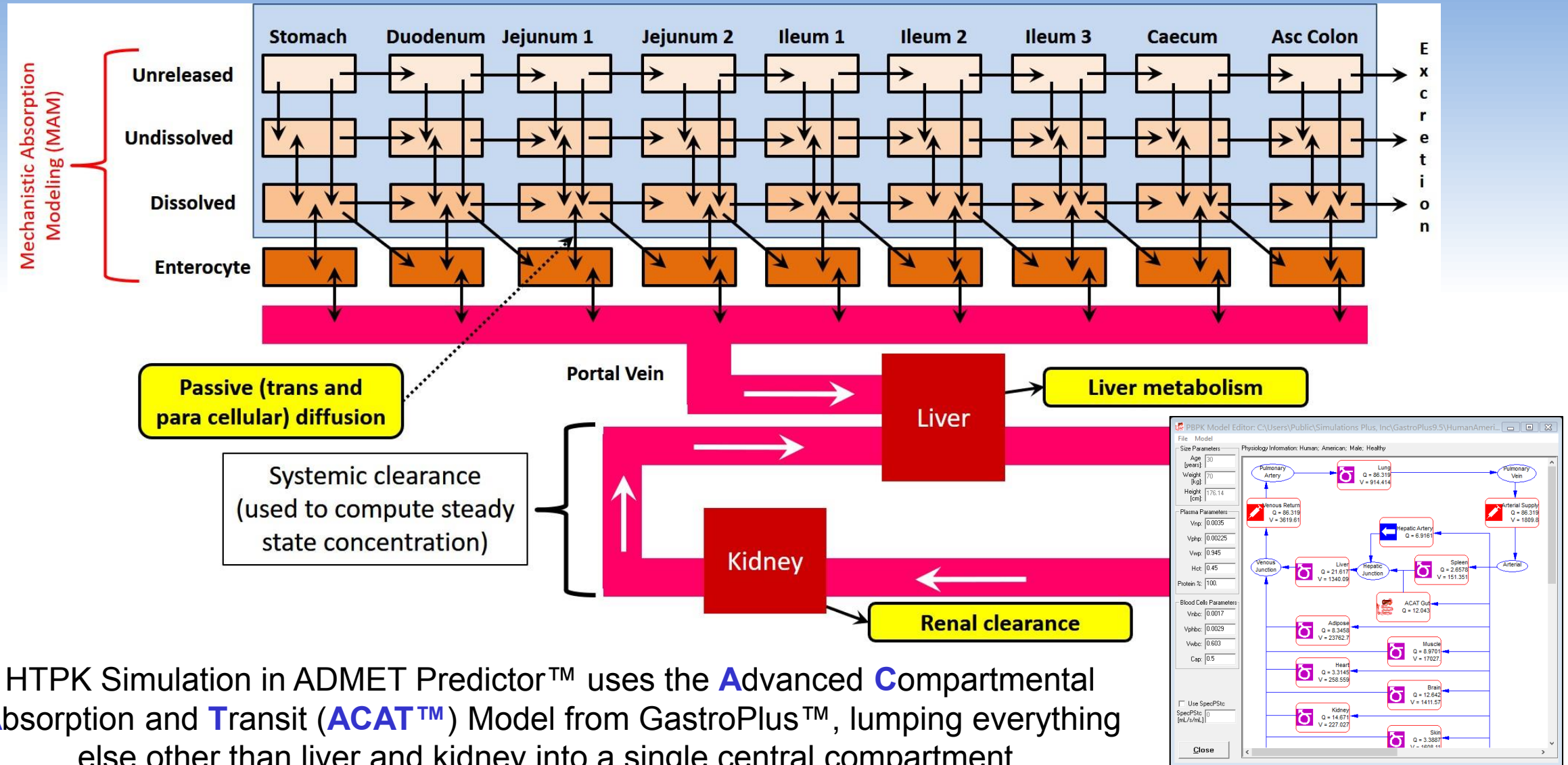


The screenshot shows the 'Simulate fraction absorbed and bioavailable' dialog box. It includes a 'Process status' section, a 'Species' dropdown set to 'Rat', and a 'Dose(s) [mg]' input field. There are checkboxes for '% Absorbed' (checked) and '% Bioavailable' (checked), each with a 'Prefix' field. The 'Clearance parameter' section has a 'Type' dropdown set to 'Liver microsomes', a 'Preferred value' field set to 'CYP_RLM_CLint', and a 'Fallback value' field set to 'CYP_RLM_CLint'. Both 'Preferred value' and 'Fallback value' have 'Unbound' checkboxes checked. The 'Advanced' button is highlighted with a green box.

- Predictions provided for physicochemical properties and total microsomal CYP clearance
- Human and rat both supported
- Takes user specified data or predictions where available

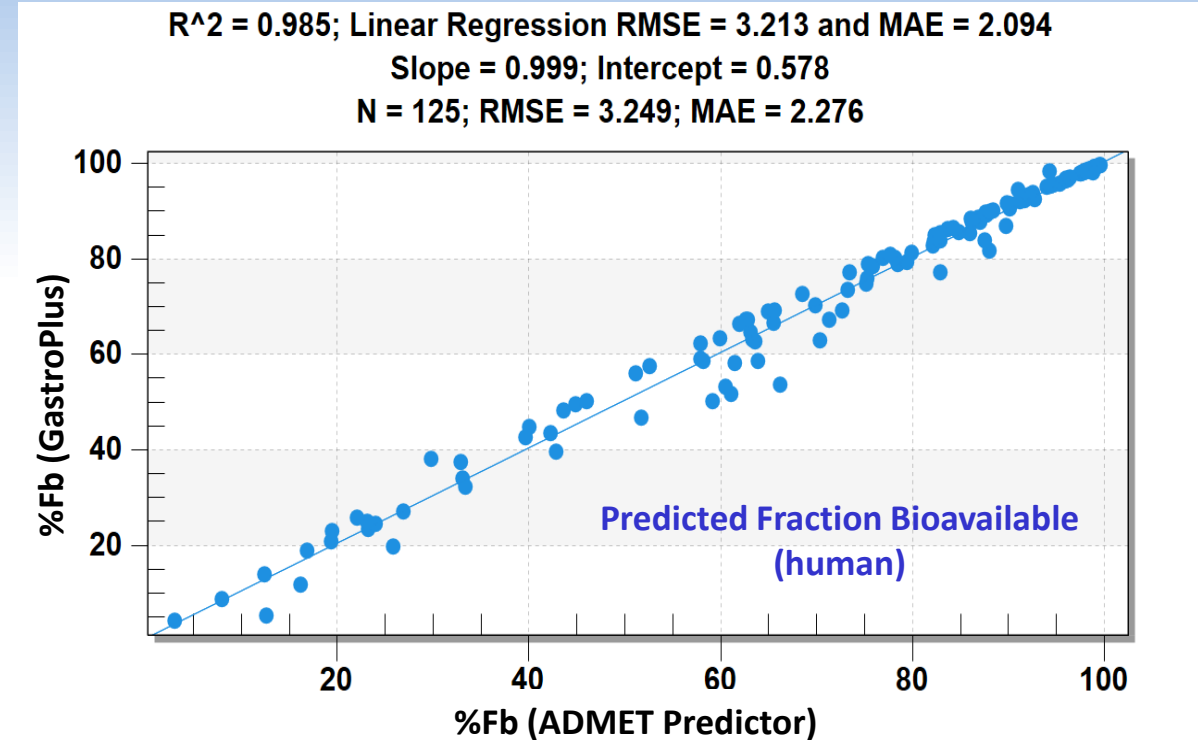
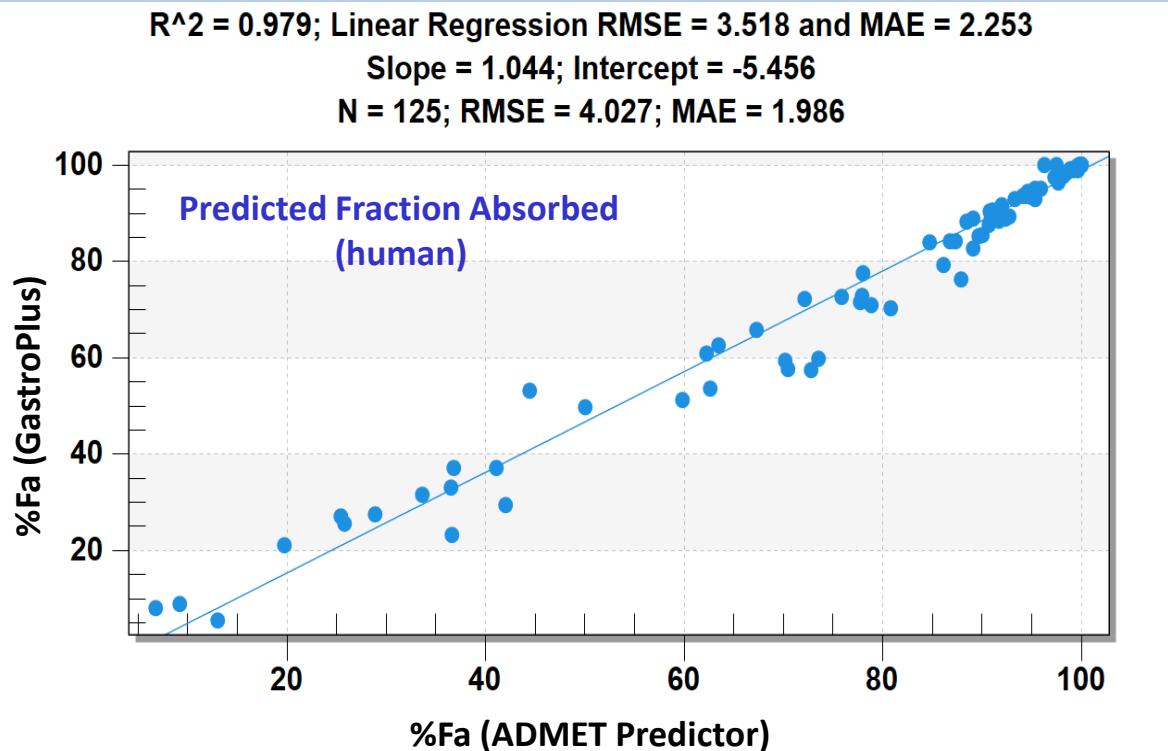
“A thing should be made as simple as possible, but no simpler.”
- Albert Einstein

...But Complicated Enough to Get the Job Done



The Streamlined System Output is Similar to Full Simulation*

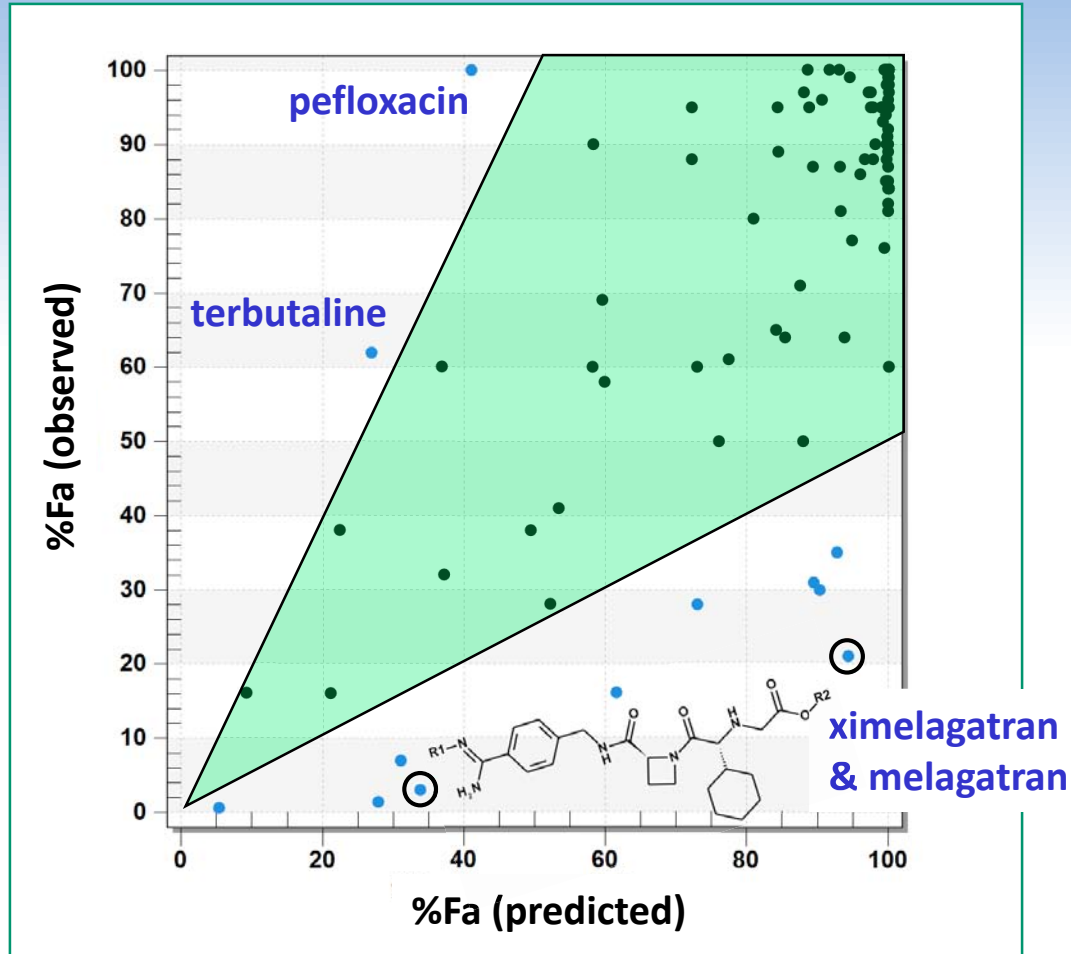
125 passively absorbed compounds from Zhao et al., *J. Pharm. Sci.* **2001**, 90:749



*GastroPlus run using a compartmental model with ACAT absorption vs HTPK in ADMET Predictor.

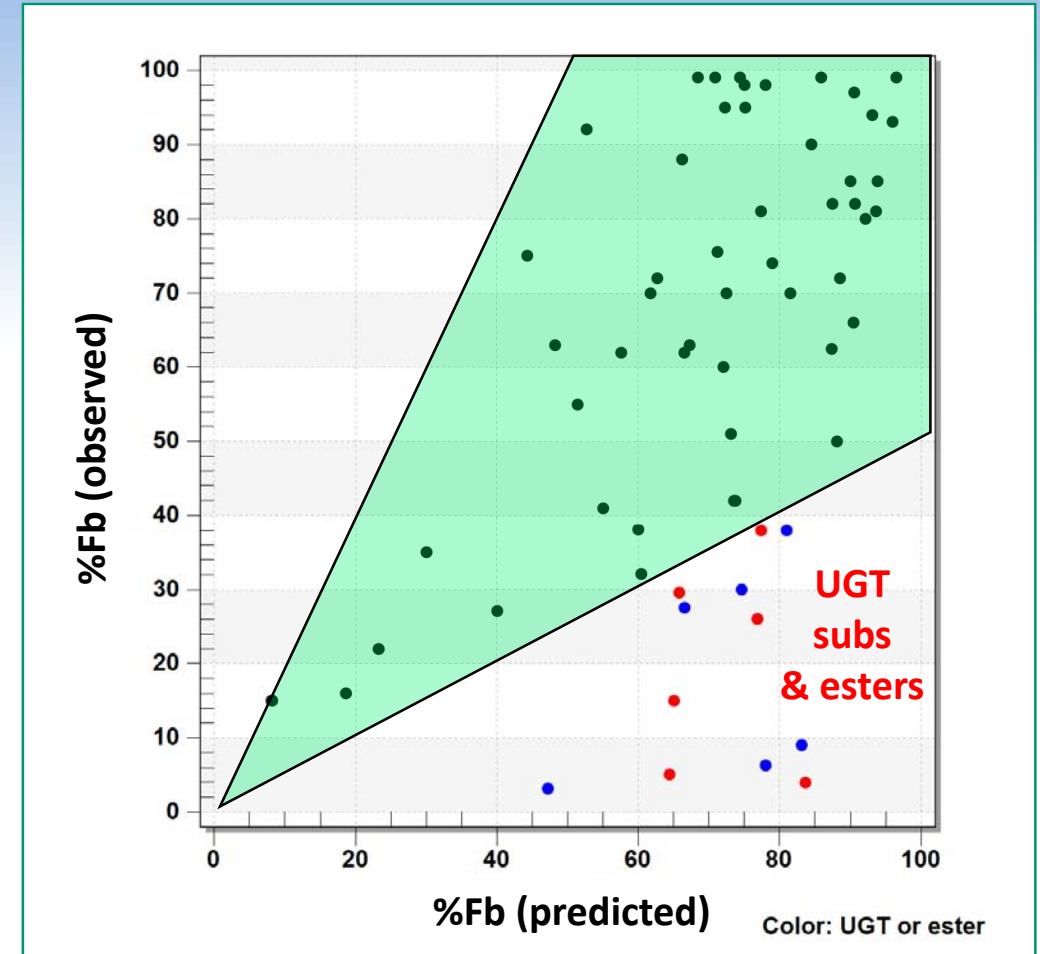
Does HTPK Give Good Enough Answers?

Human %Fa for 115 compounds
(Zhao et al. *J. Pharm. Sci.*, **2001**, 90: 749 and elsewhere)



90% predicted within 2-fold of the reported value.
83% predicted within 1.5-fold.

Human %Fb for 62 CYP-metabolized compounds
(Toshimoto et al. *Drug Metab. Dispos.* **2014**, 42:1811 etc.)



81% predicted within 2-fold of the reported value.
68% predicted within 1.5-fold

Rat Bioavailability: Data Set 1

- 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

Current Topics in Medicinal Chemistry 2003, 3, 1269-1288

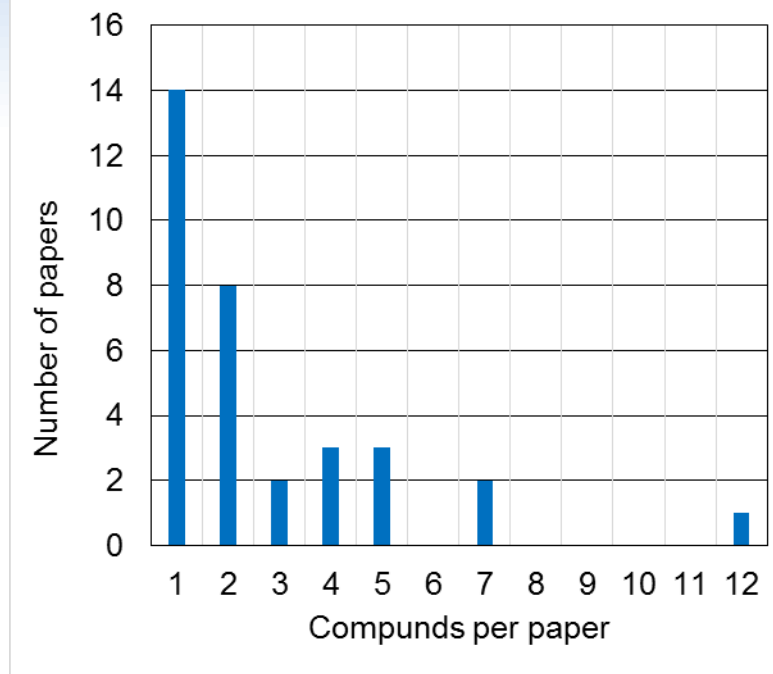
1269

Molecular Design and Bioavailability

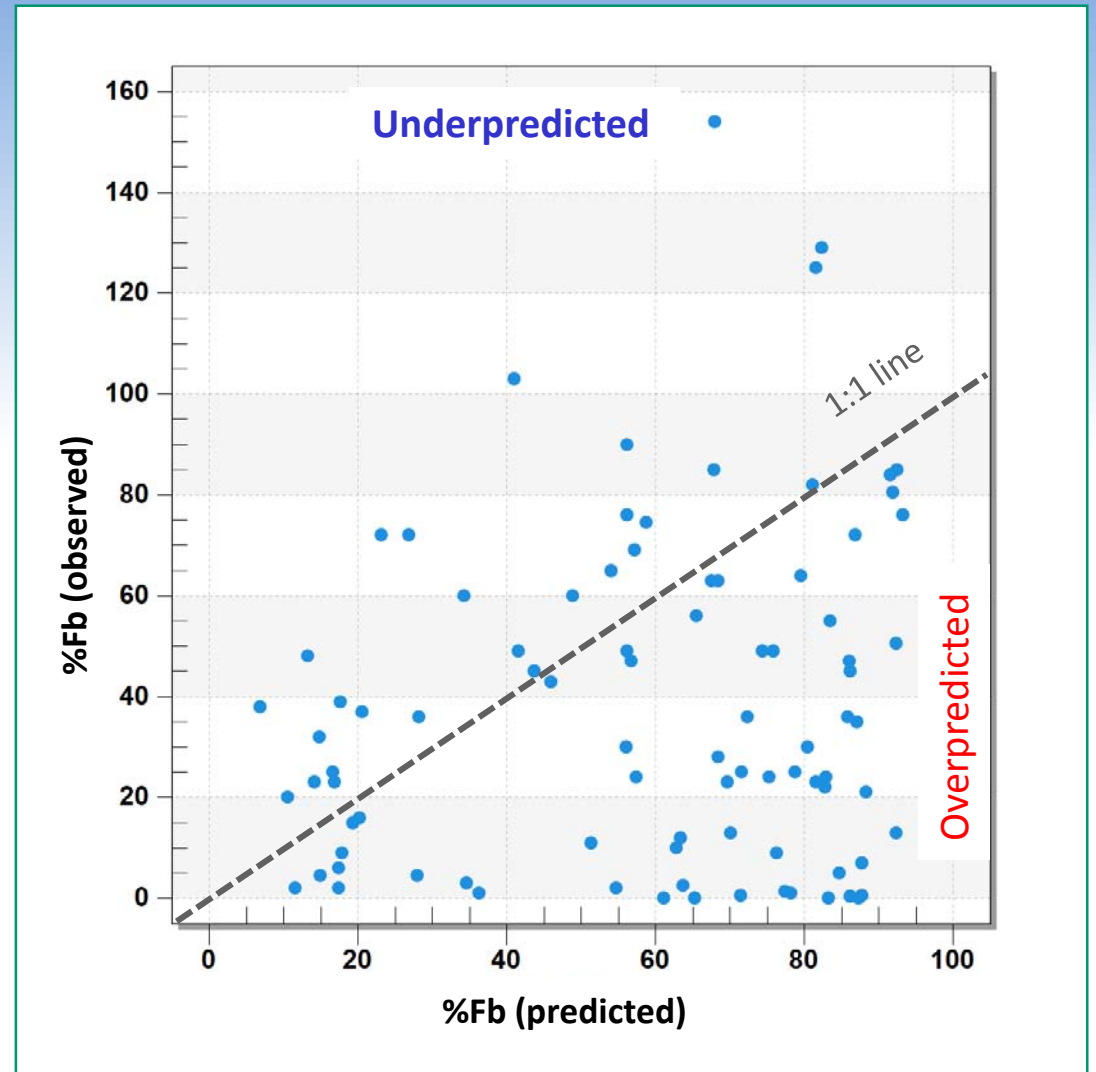
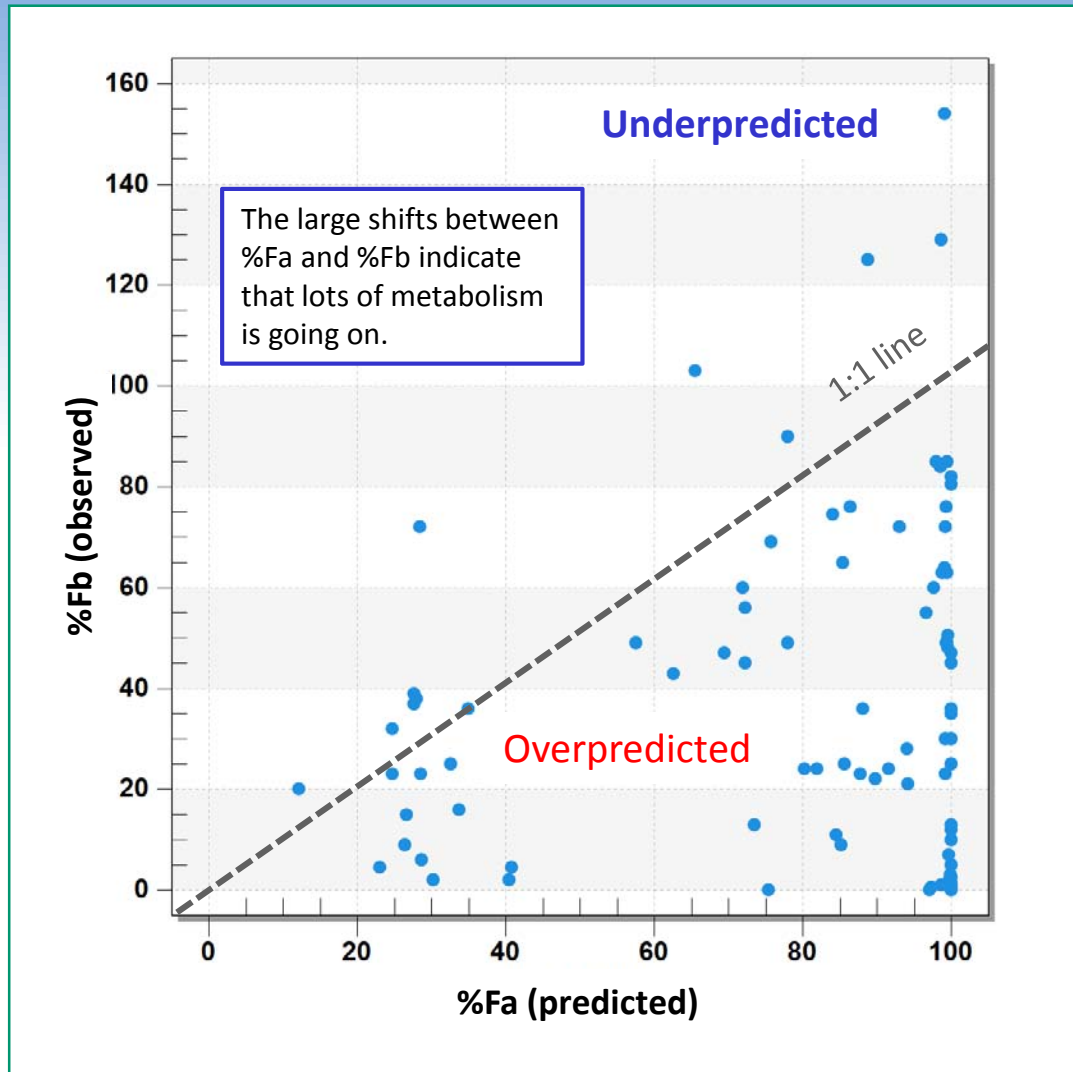
Robert D. Clark* and Philippa R.N. Wolohan

Tripos, Inc., 1699 S. Hanley Road, St. Louis MO 63144 USA

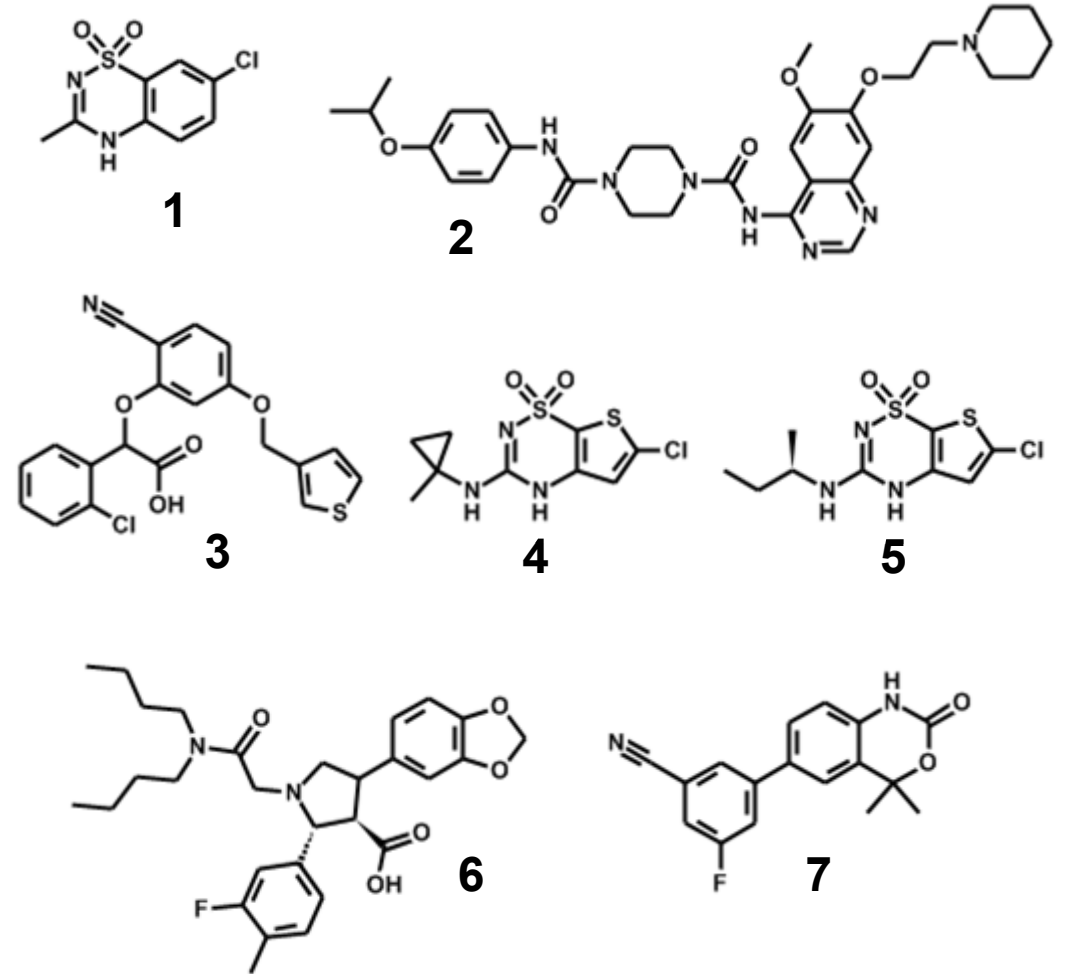
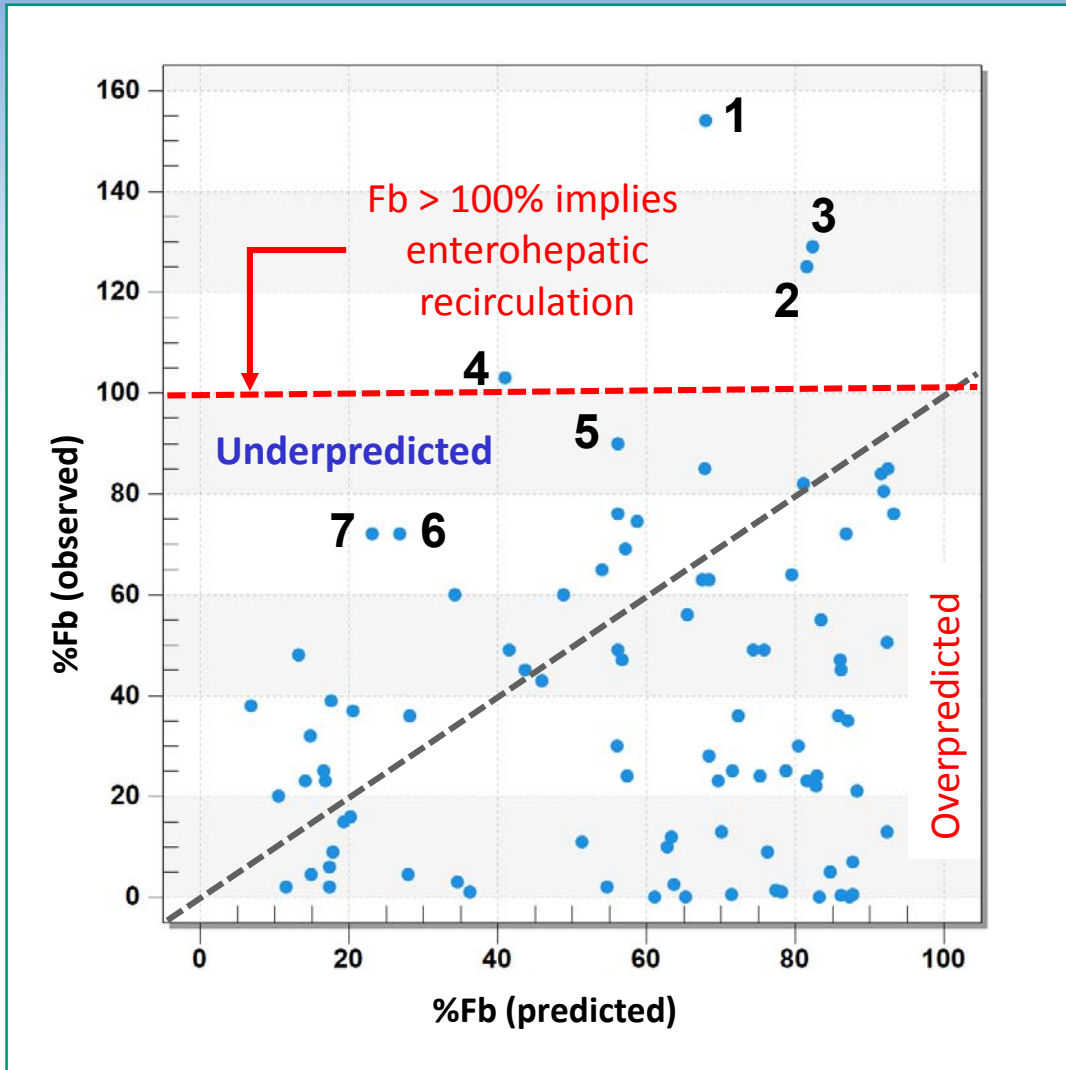
Abstract: A “snapshot” of current medicinal chemistry work on bioavailability is drawn from issues of *J. Med. Chem.* covering the time period between September 2001 and September 2002. An exhaustive compilation of reported absolute oral bioavailability (*F*) values for this period is included, covering 34 structural series and 107 distinct compounds, with data for multiple species in many cases. This is supplemented with a discussion of more qualitative oral bioavailability results, and with illustrative examples addressing clearance, prodrug design, and blood/brain barrier penetration problems. Papers discussing predictions pertaining to one or another aspect of bioavailability are also discussed, and some thoughts on future directions of work on *in silico* prediction in this area are presented.



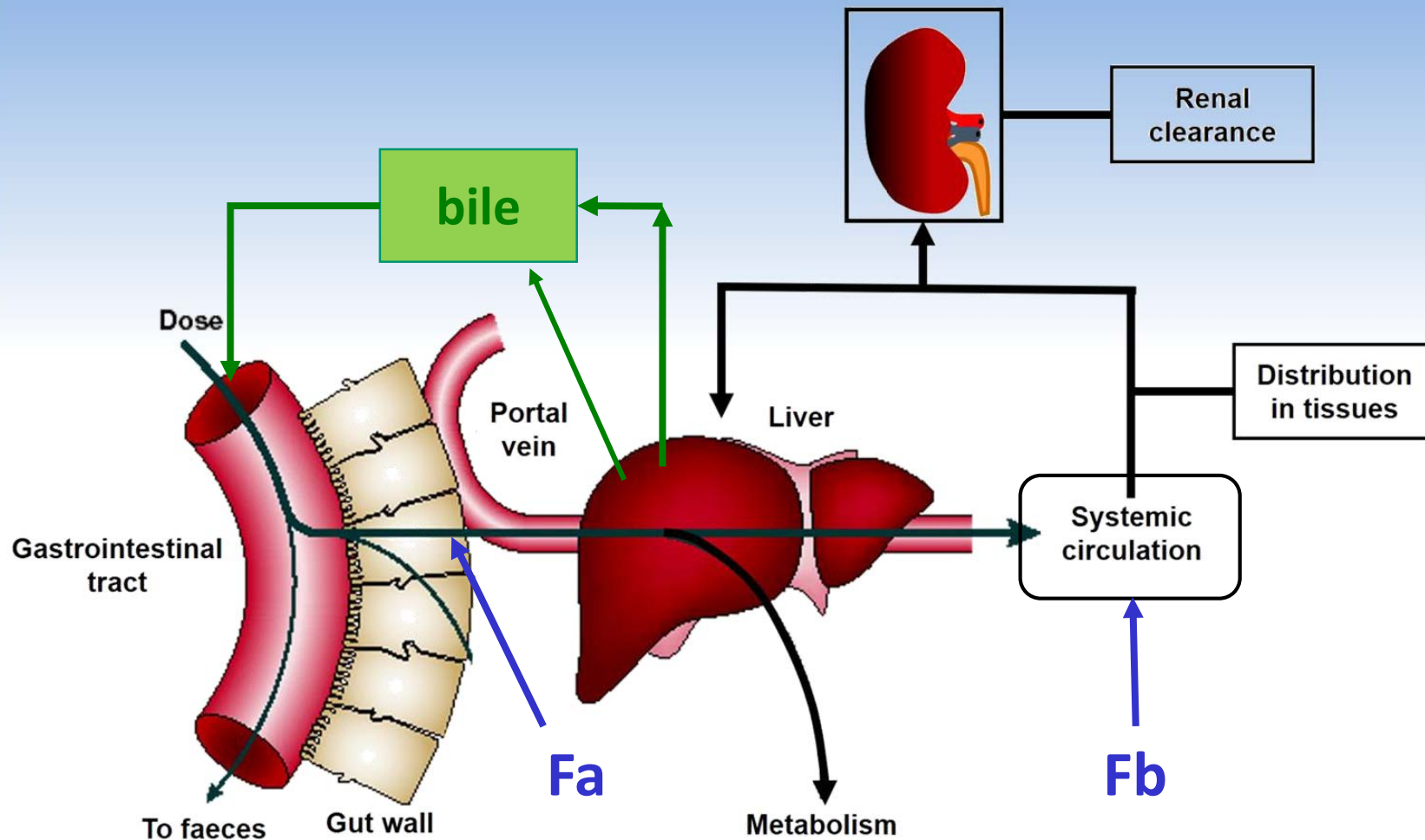
Experimental Fa and Fb vs. *in silico* Predictions



Why Are Some Bioavailabilities Underpredicted?



How Enterohepatic Circulation Works



Simulate fraction absorbed and bioavailable

Process status:

Species: ☒ Rat ☐ Human

Dose(s) [mg]:

☒ % Absorbed Prefix:

☒ % Bioavailable Prefix:

Clearance parameter:

Type:

Preferred value: ☒ Unbound

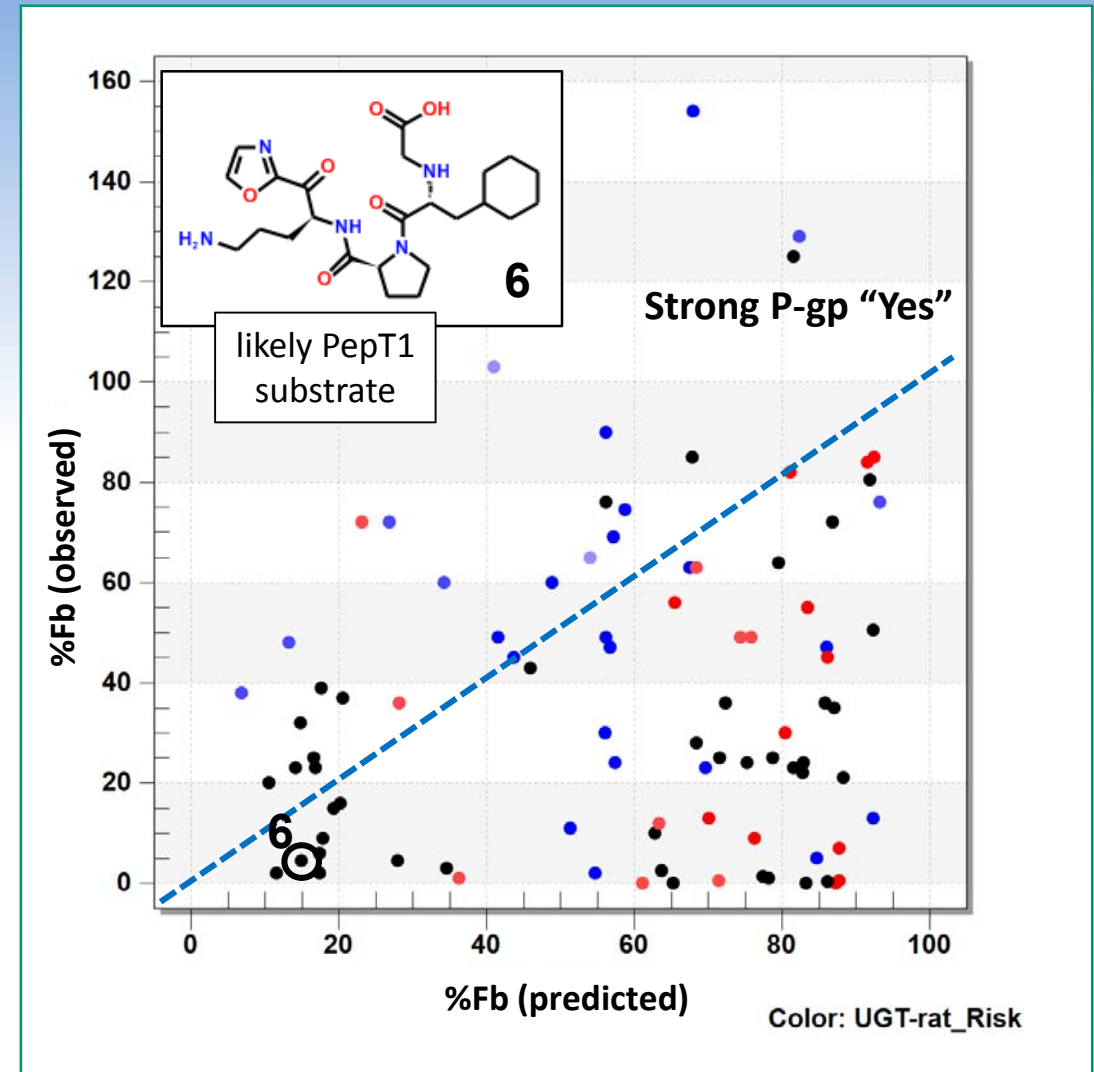
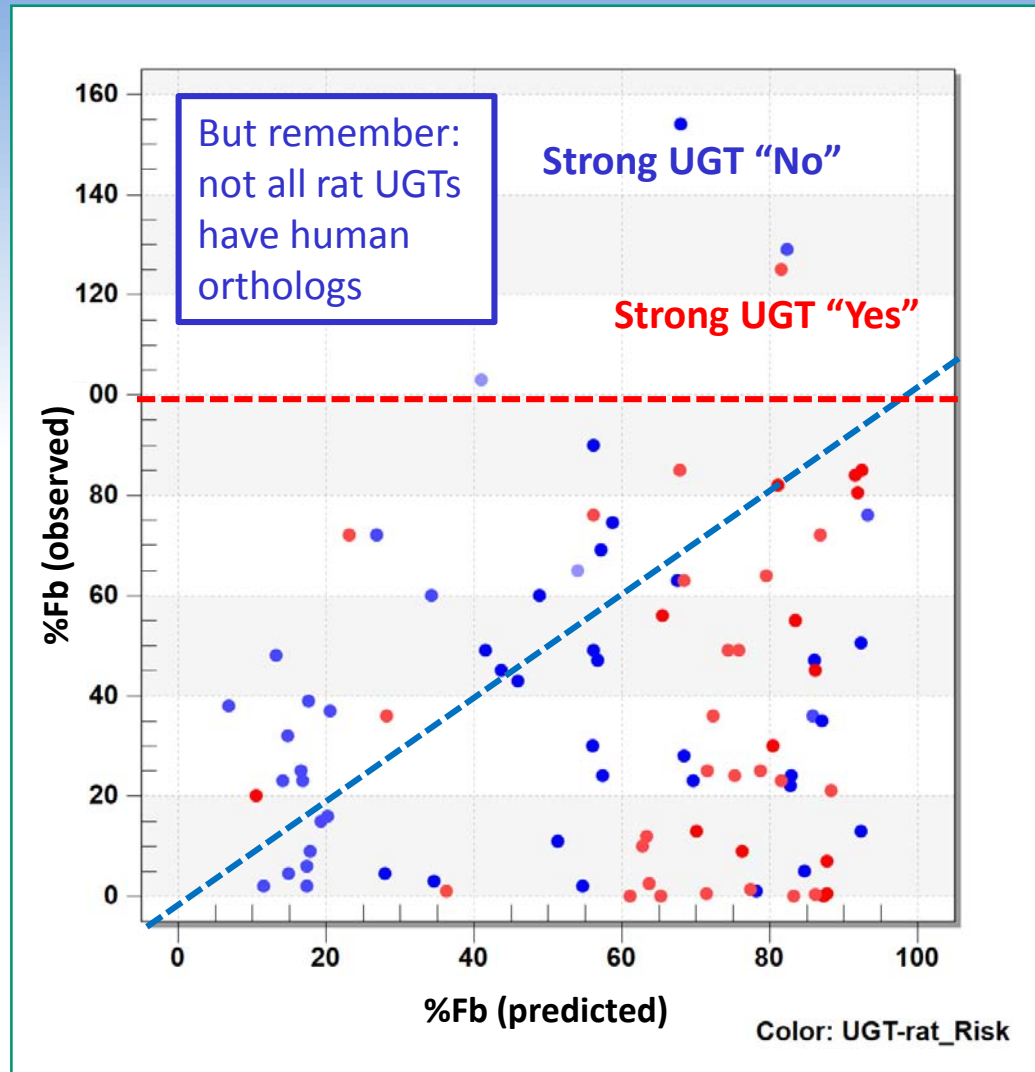
Fallback value: ☒ Unbound

Buttons: Minimize, Advanced, Save, Run, Cancel

“A thing should be made as simple as possible, but no simpler.”
- Albert Einstein

Strong P-gp "Yes"

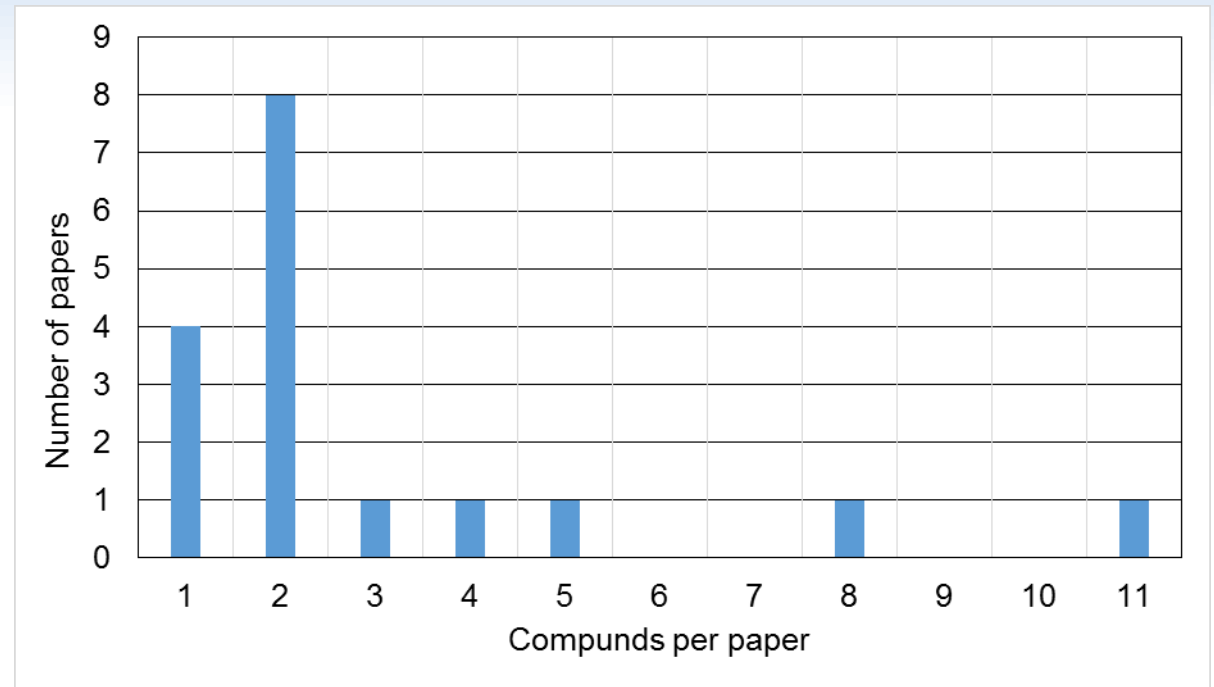
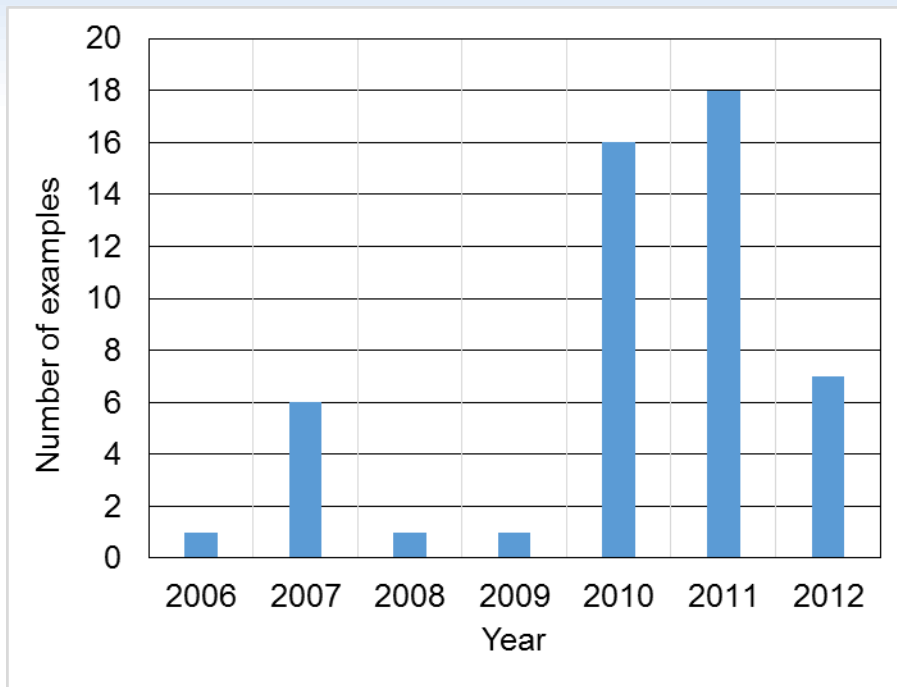
Effect of Being a UGT* or hP-gp Substrate on Predicted Rat PK



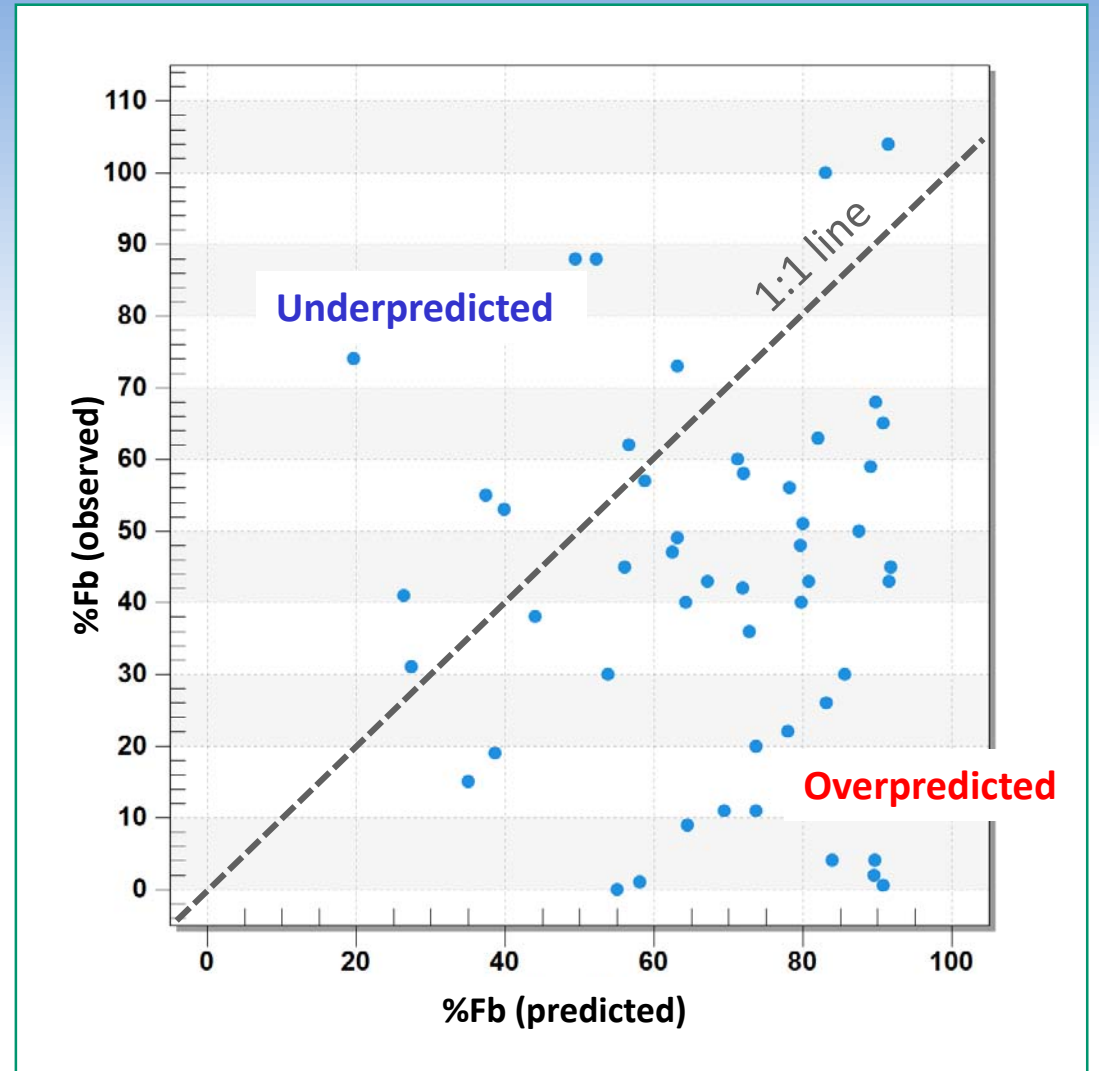
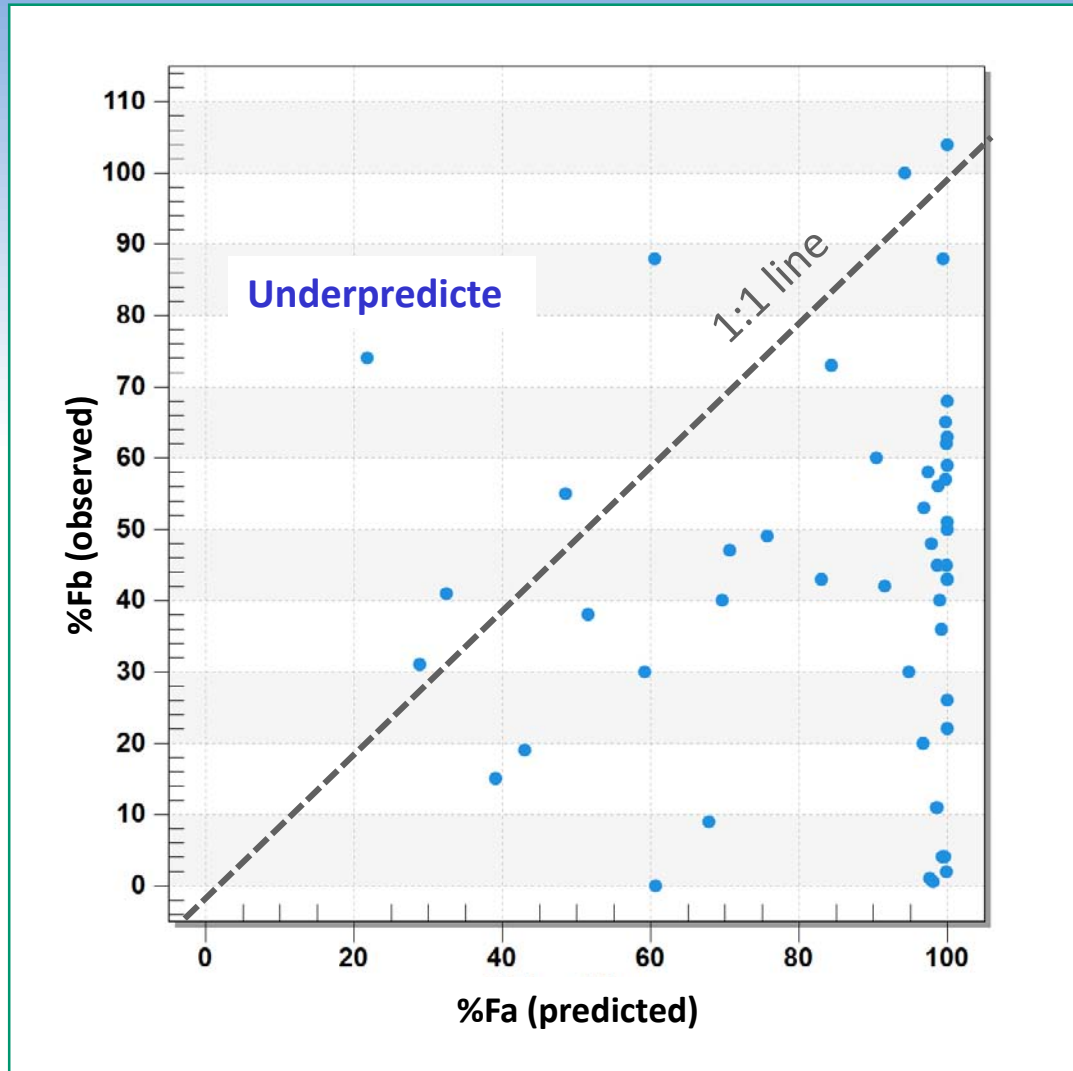
*Based on expression levels of rat orthologs of human UGTs human UGTs and hUGT substrate classifications

Rat Bioavailability: Data Set 2

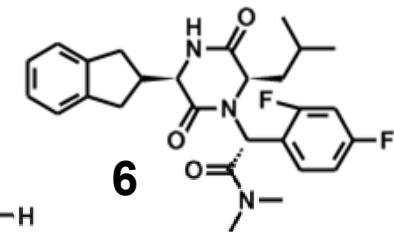
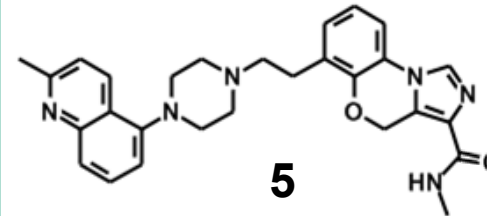
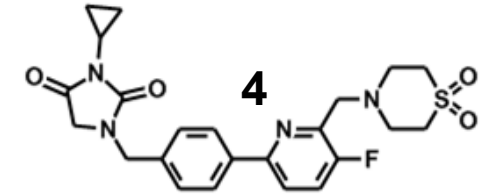
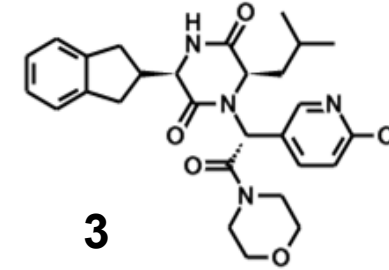
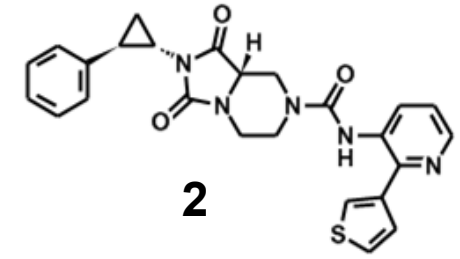
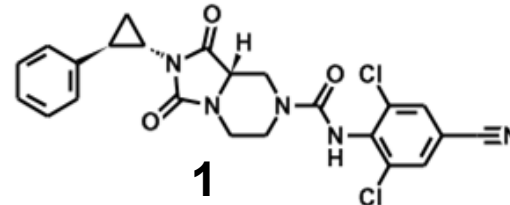
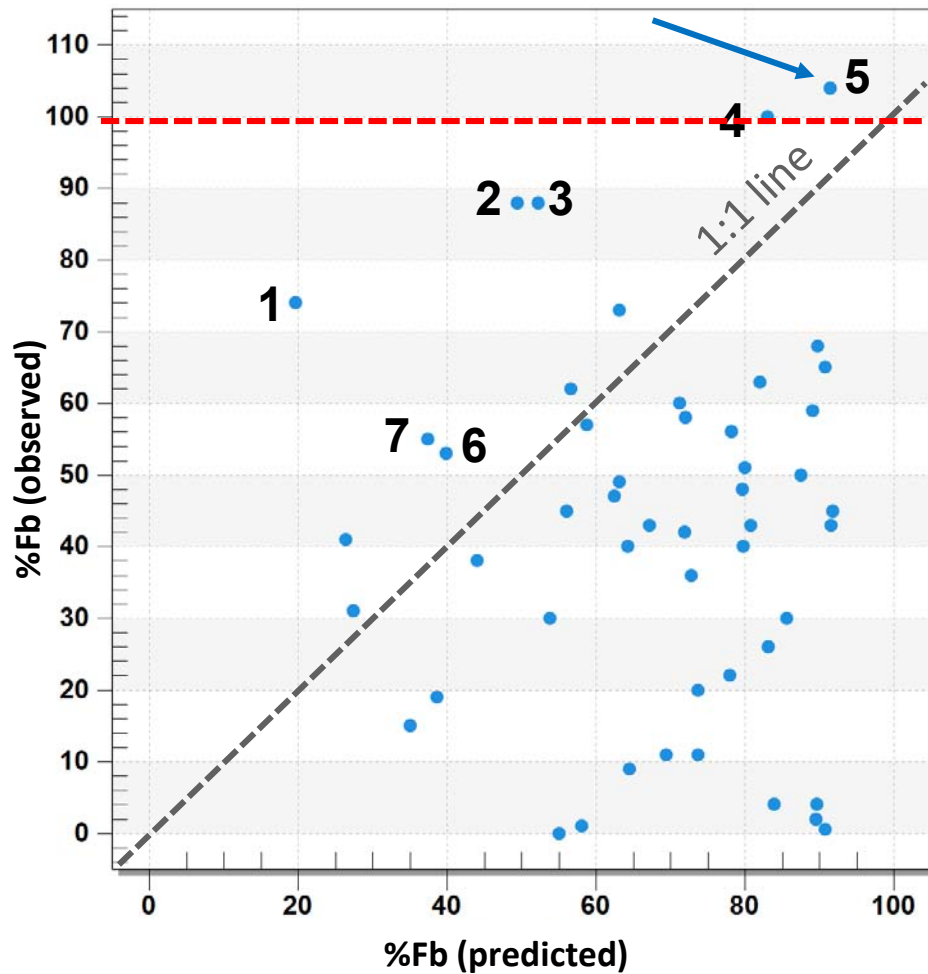
- Data taken from drug design and development papers published in the *Journal of Medicinal Chemistry* between 2006 and 2012 that report oral bioavailability in rats for one or more analogs
 - rat bioavailabilities for 51 compounds with many different targets
 - 1-11 examples from 17 papers
 - many also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



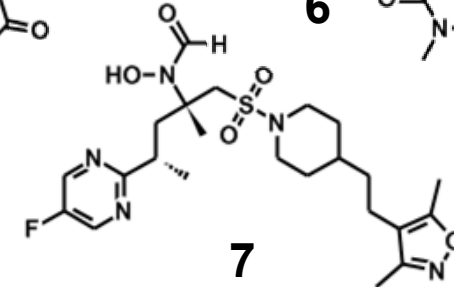
Experimental Fa and Fb vs. *in silico* Predictions in Rats



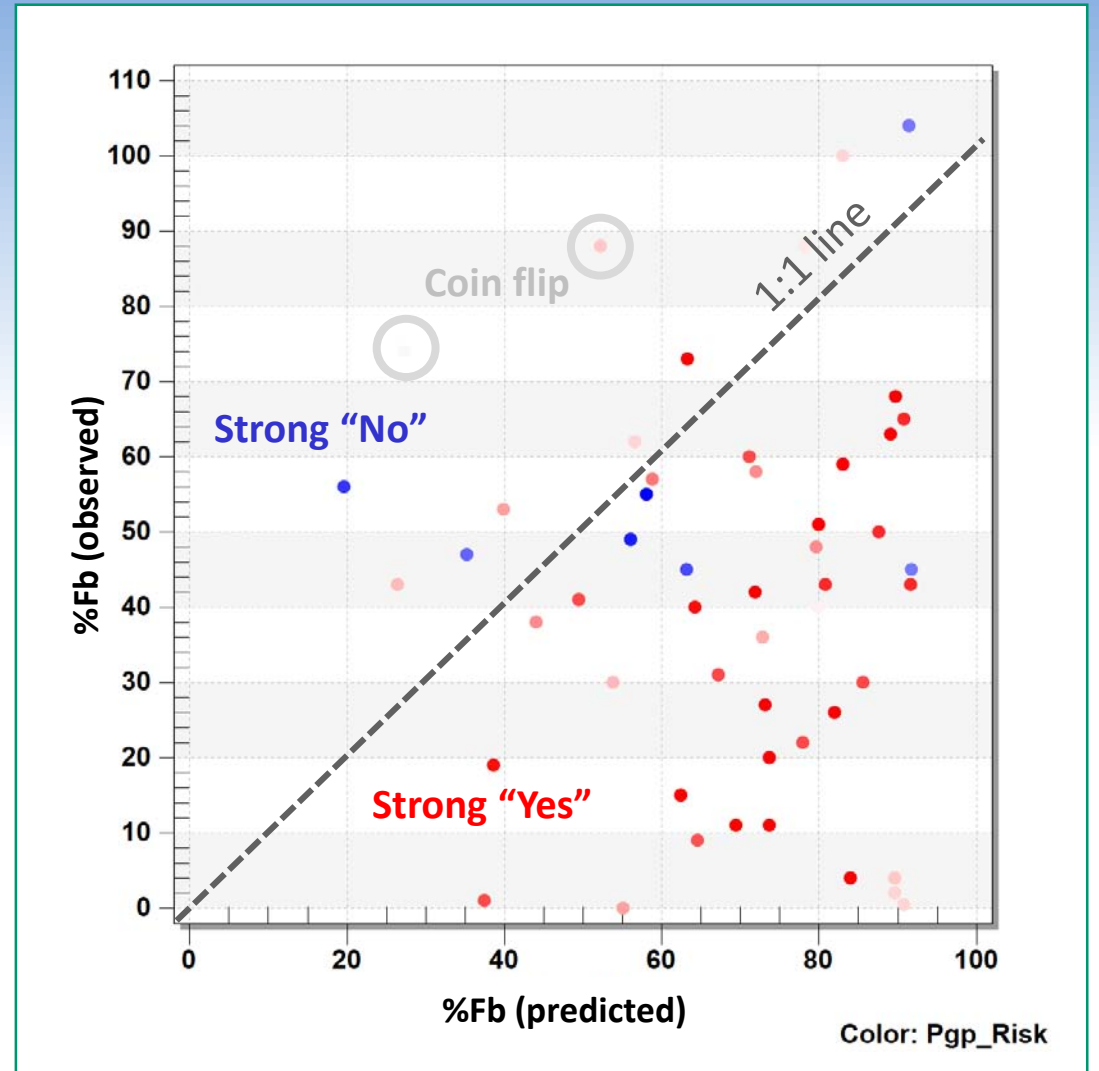
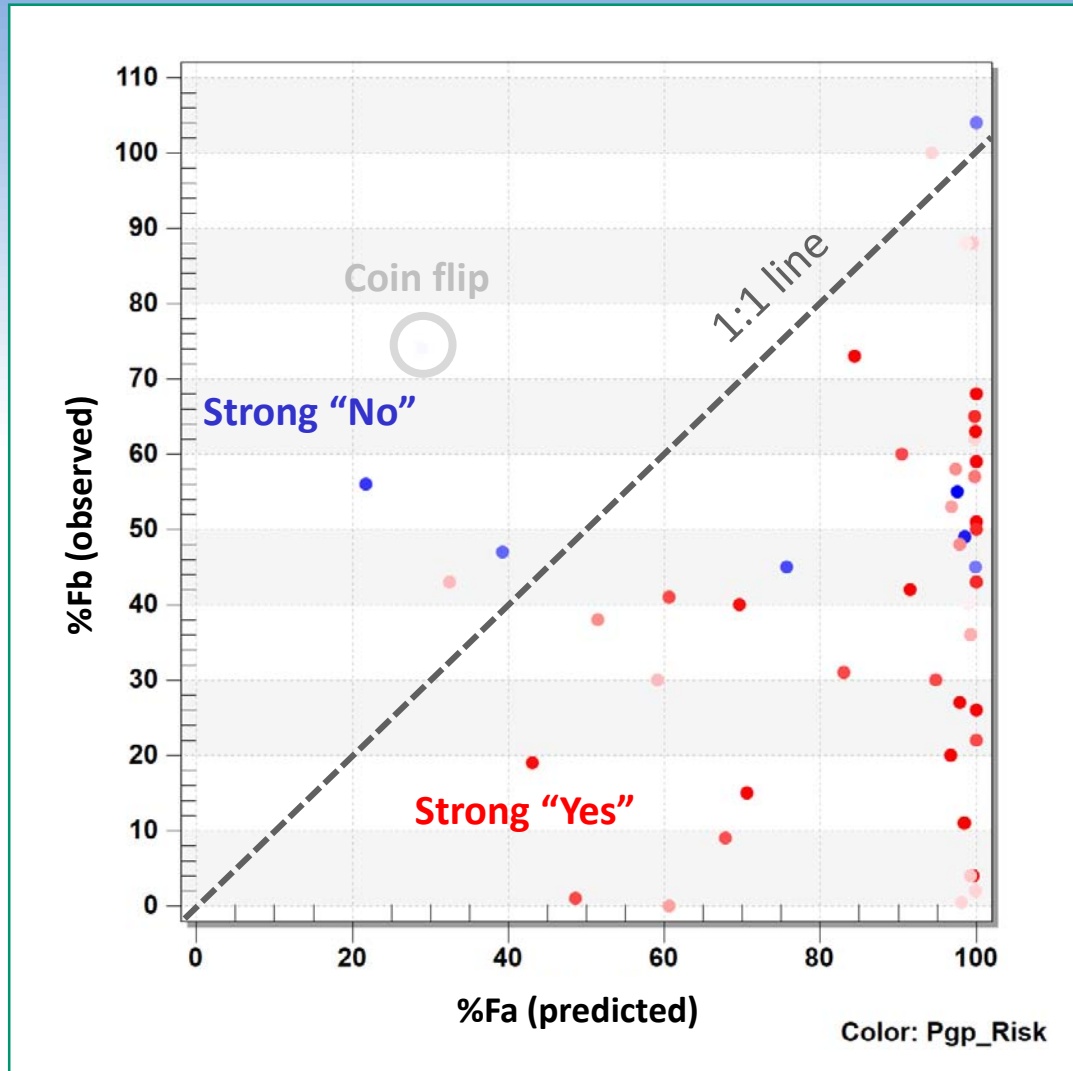
Why Are Some Bioavailabilities Underpredicted?



Fb > 100% implies
enterohepatic
recirculation

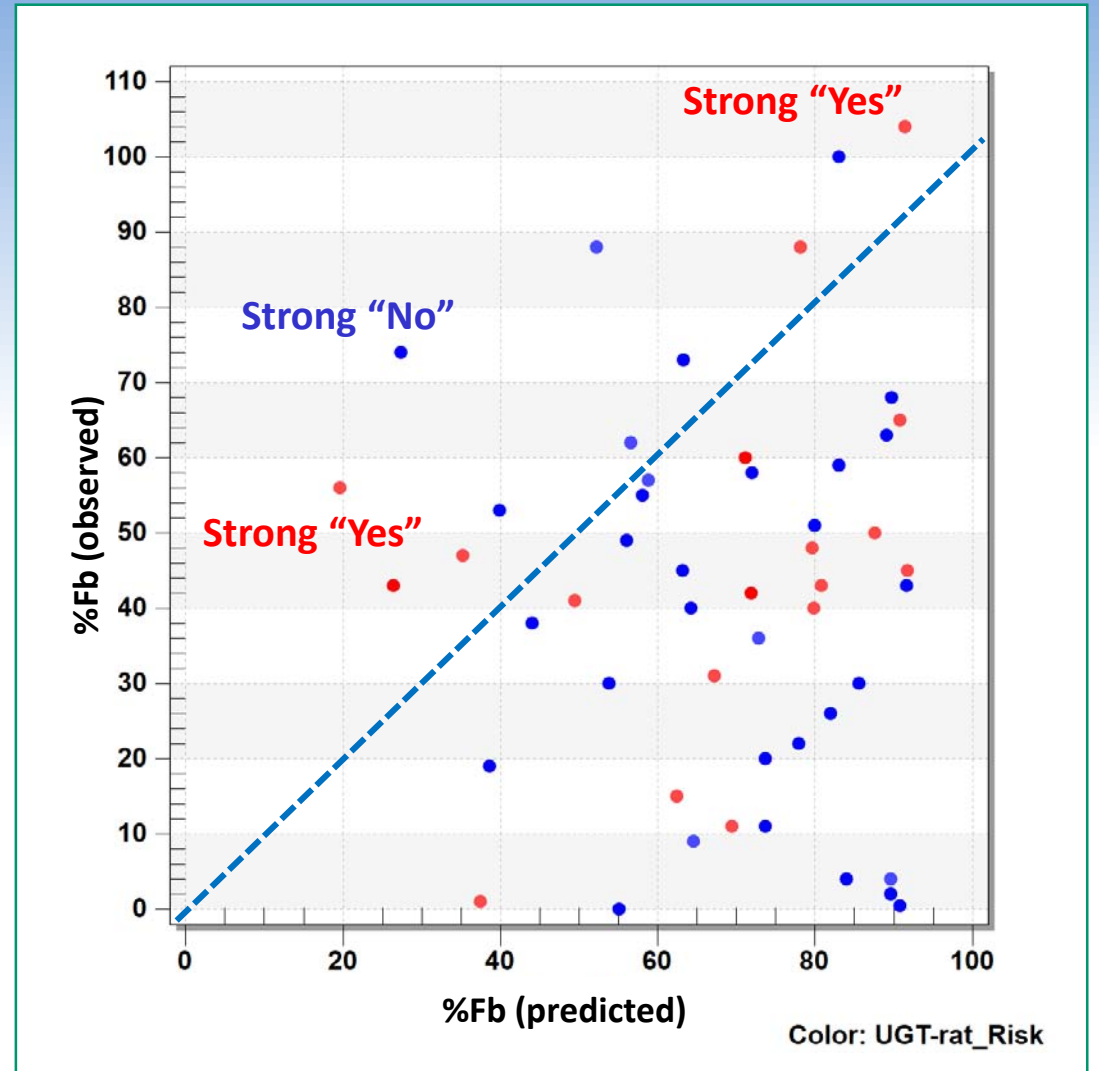
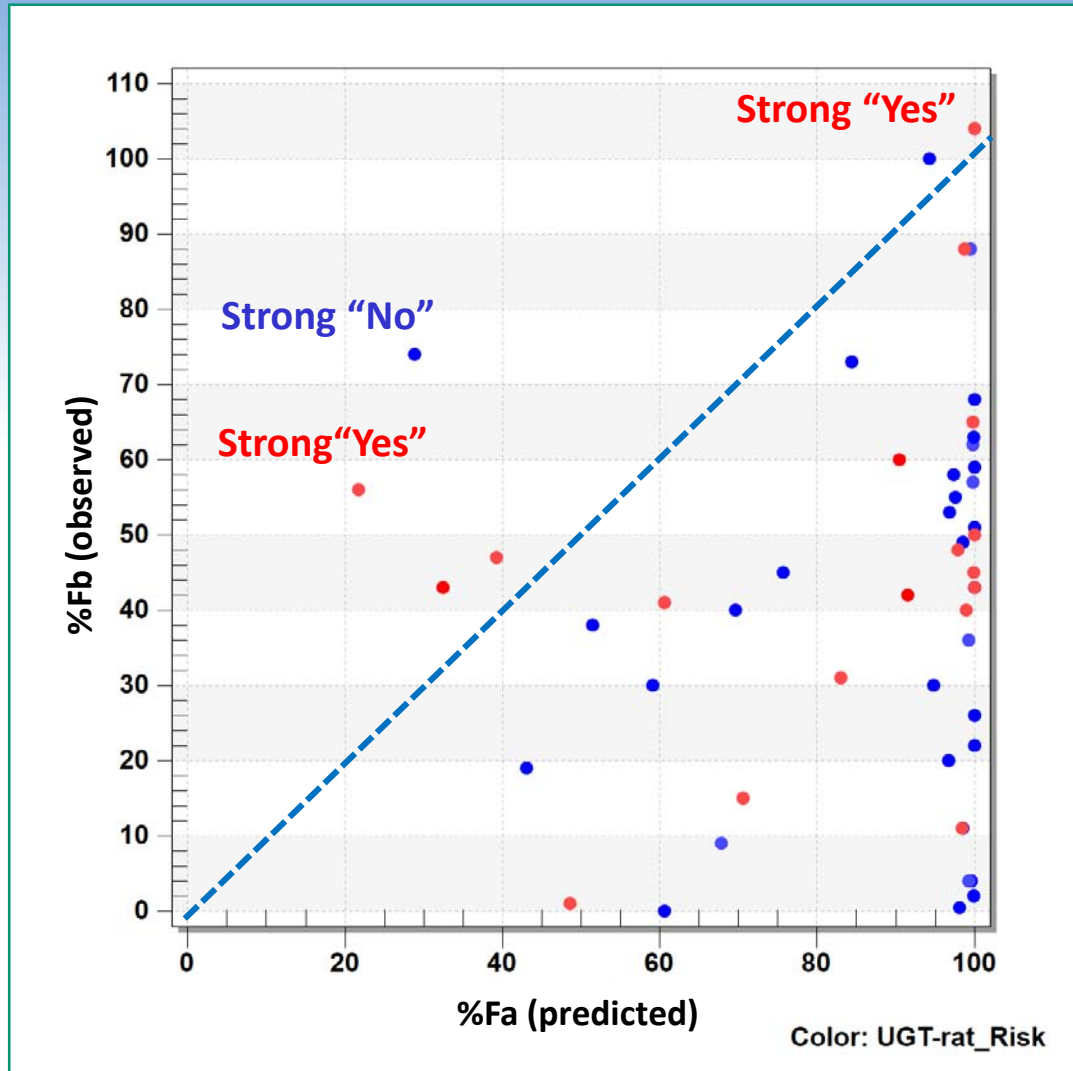


Effect of Being a P-gp Substrate* on Predicted Rat PK



*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

Recent Related Publications

tm00 | ACSJCA | JCA10.01465/W Unicode | research:3f (R3.6.12 HF03:4459 | 2.0 alpha 39) 2017/11/27 07:41:00 | PROD-JCA1 | rq_11676879 | 1/25/2018 11:27:33 | 9 | JCA-DEFAULT

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

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

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

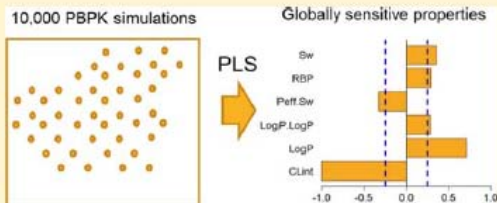
6 [§]Simulations Plus, Inc., 42505 10th Street West, Lancaster, California 93534, United States

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

9  Supporting Information

10 **ABSTRACT:** When medicinal chemists need to improve oral bioavailability (%F) during lead optimization, they systematically modify compound properties mainly based on their own experience and general rules of thumb. However, at least a dozen properties can influence %F, and the difficulty of multiparameter optimization for such complex nonlinear processes grows combinatorially with the number of variables. Furthermore, strategies can be in conflict. For example, adding a polar or charged group will generally increase solubility but decrease permeability. Identifying the 2 or 3 properties that most influence %F for a given compound series would make %F optimization much more efficient. We previously reported an adaptation of physiologically based pharmacokinetic (PBPK) simulations to predict %F for a lead series from purely computational inputs within a 2-fold average error. Here, we run thousands of such simulations to generate a comprehensive "bioavailability landscape" for the series. A key innovation was recognition that the large and variable number of pK_a 's in drug molecules could be replaced by just the two straddling the isoelectric point. Another was use of the ZINC database to cull out chemically inaccessible regions of property space. A quadratic partial least squares regression (PLS) accurately fits a continuous surface to these thousands of bioavailability predictions. The PLS coefficients indicate the globally sensitive compound properties. The PLS surface also displays the %F landscape in these sensitive properties locally around compounds of particular interest. Finally, being quick to calculate, the PLS equation can be combined with models for activity and other properties for multiobjective lead optimization.

30 **KEYWORDS:** PBPK, lead optimization, lead series



RD Clark, Predicting mammalian metabolism and toxicity of pesticides in silico. *Pest Management Science* 2018, in press.

tm00 | ACSJCA | JCA10.01465/W Unicode | research:3f (R3.6.12 HF03:4459 | 2.0 alpha 39) 2017/11/27 07:41:00 | PROD-JCA1 | rq_6742617 | 1/25/2018 11:23:18 | 10 | JCA-DEFAULT

molecular
pharmaceutics

Article

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

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

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

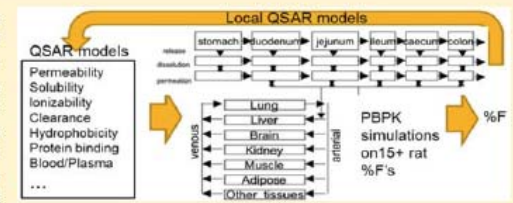
6 [§]Simulations Plus, Inc., 42505 10th Street West, Lancaster, California 93534, United States

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

9  Supporting Information

10 **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_{loc}). All inputs are subsequently computed from structure alone, so the models can be applied in advance of synthesis. 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.

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



 **SimulationsPlus**
SCIENCE + SOFTWARE = SUCCESS

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:

- Robert Fraczekiewicz
- David Miller
- Pankaj Daga
- Eric Martin & Ben Madej (Novartis)
- Michael Lawless
- Mike Bolger
- Viera Lukacova
- John DiBella
- Karen Webster

...and to you for your kind attention.

bob@simulations-plus.com