

## SUPPLEMENTAL DATA

# Predictions of Hepatic Disposition Properties Using a Mechanistically Realistic, Physiologically Based Model

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**Example Calculations.** Application of the FCMA was automated. Nevertheless, below we walk through examples of predicting two ISL parameter values for prazosin and propranolol.

I. Predict a value of the parameter *B2CJumpProb* for prazosin and propranolol:

1. For prazosin:

**Step 0.** Selection of PCPs and creating the data set.

The data set contains  $n = 4$  compounds:  $S = \{\text{atenolol, antipyrine, labetalol, diltiazem}\}$  for which the simulation parameters are known:

ISL Parameters	ATENOLOL	ANTIPYRINE	LABETALOL	DILTIAZEM
<i>B2CJumpProb</i>	0.3	0.35	0.5	0.5

*B2CJumpProb* is one of the four parameters in the ISL subgroup B, moving between spaces (Table 1). All four are calculated in the same way. Figure 4 specifies that for the three PCP properties in subgroups II (partitioning) and IV (topology related) are expected to be equally influential in determining the value of all ISL subgroup B parameters.

PCP Subgroups II & IV	Atenolol	Antipyrine	Labetalol	Diltiazem	Prazosin
RB count	8	1	8	7	4
TPSA	84.6	26.9	95.6	59.1	107.0
$\log P_{app}$	0.14	0.33	2.69	3.53	1.88

**Step 1.** Let  $q = 4$ , and  $S_{new} = \{\text{atenolol, antipyrine, labetalol, diltiazem, prazosin}\}$ .

**Step 2.** If  $q = 1$  go to step 4. Else, classify  $S_{new}$  into  $q$  clusters using Fuzzy c-Means algorithm.

**Step 3.** If prazosin is not in the same group with *at least* another member, then decrease  $q$  to  $q - 1$ . Repeat steps 2 and 3. (For the values above, prazosin is not in the same group with *at least* one other member. When we repeat for  $q = 3$ , we observe that prazosin is in the same group with *at least* one other member, so we stop.)

Else, let *G-value* be the number of group-mates of prazosin, (the clustering results show that prazosin has two group-mates,  $G = 2$ .) Go to step 4.

**Step 4.** The five compounds are clustered into three groups with the following membership degrees.

cluster	Atenolol	Antipyrine	Labetalol	Diltiazem	Prazosin
C 3	<b>0.987</b>	0.001	0.050	0.060	0.316
C 2	0.006	<b>0.998</b>	0.020	0.045	0.179
C 1	0.007	0.001	<b>0.930</b>	<b>0.895</b>	<b>0.506</b>

PRAZOSIN'S simulation parameter,  $B2CJumpProb$ , is estimated as:

$$B2CJumpProb_{\text{prazosin}} = 0.506 \cdot P_{C1} + 0.179 \cdot P_{C2} + 0.316 \cdot P_{C3}$$

where  $P_{Ci}$  is the average ISL parameter values of members of cluster  $i$ :

$$P_{C1} = (B2CjumpProb_{\text{labetalol}} + B2CjumpProb_{\text{diltiazem}})/2 = 0.5$$

$$P_{C2} = B2CjumpProb_{\text{antipyrine}} = 0.35$$

$$P_{C3} = B2CjumpProb_{\text{atenolol}} = 0.3$$

Consequently,

$$B2CjumpProb_{\text{prazosin}} = 0.506 \times 0.5 + 0.179 \times 0.35 + 0.316 \times 0.3 = 0.4104 \text{ (rounded to 0.41)}$$

## 2. For PROPRANOLOL:

PCP Subgroups II & IV	Atenolol	Antipyrine	Labetalol	Diltiazem	Propranolol
RB count	8	1	8	7	6
TPSA	84.6	26.9	95.6	59.1	41.5
$\log P_{app}$	0.14	0.33	2.69	3.53	3.10

**Steps 1-3.** Let  $q = 4$ , and  $S_{new} = \{\text{atenolol, antipyrine, labetalol, diltiazem, propranolol}\}$ .

For  $q = 4$  and the values above, propranolol is in the same group with *at least* one other member, and  $G = 1$ .

**Step 4.** The classification results are:

cluster	Atenolol	Antipyrine	Labetalol	Diltiazem	Propranolol
C 4	<b>1</b>	0	0	0.014	0.014
C 3	0	<b>1</b>	0	0.010	0.014
C 2	0	0	<b>1</b>	0.067	0.036
C 1	0	0	0	<b>0.908</b>	<b>0.936</b>

$$P_{C1} = B2CjumpProb_{\text{diltiazem}} = 0.5$$

$$P_{C2} = B2CjumpProb_{\text{labetalol}} = 0.5$$

$$P_{C3} = B2CjumpProb_{\text{antipyrine}} = 0.35$$

$$P_{C4} = B2CjumpProb_{\text{atenolol}} = 0.3$$

Consequently,

$$B2CJumpProb_{\text{propranolol}} = 0.936 \times 0.5 + 0.036 \times 0.535 + 0.014 \times 0.35 + 0.014 \times 0.3 = 0.4951 \text{ (rounded to 0.50)}$$

## II. Predict parameter $SoluteBindingProb$ for prazosin and propranolol

### 1. For prazosin:

**Step 0.** Selection of PCPs and creating the data set.

The data set contains  $n = 4$  compounds:  $S = \{\text{atenolol, antipyrine, labetalol, diltiazem}\}$  for which the simulation parameters are known:

ISL Parameters	ATENOLOL	ANTIPYRINE	LABETALOL	DILTIAZEM
<i>SoluteBindingProb</i>	0.35	0.5	0.6	0.35

*SoluteBindingProb* is one of the four parameters in the ISL subgroup C, binding to LOBULAR components (Table 1). All four are calculated in the same way. Figure 4 specifies that for the three PCP properties in subgroups V (hydrogen bound related) are expected to be equally influential in determining the value of all ISL subgroup C parameters.

PCP Subgroups III & V	Atenolol	Antipyrine	Labetalol	Diltiazem	Prazosin
fuB	0.74	0.6	0.52	0.28	0.54
HBD count	3	0	4	0	1
HBA count	4	2	4	5	8
Tautomer count	2	0	7	0	3

**Steps 1-3.** Let  $q = 4$ , and  $S_{new} = \{\text{atenolol, antipyrine, labetalol, diltiazem, prazosin}\}$ .

For  $q = 4$  and the values above, prazosin is in the same group with *at least* one other member, and  $G = 1$ .

**Step 4.** The classification results are:

Cluster	Atenolol	Antipyrine	Labetalol	Diltiazem	Prazosin
C4	<b>0.980</b>	0.001	0.001	0.010	0.320
C3	0.007	<b>0.994</b>	0.000	0.037	0.191
C2	0.007	0.000	<b>0.999</b>	0.003	0.118
C1	0.006	0.004	0.000	<b>0.950</b>	<b>0.371</b>

$$P_{C1} = \text{SoluteBindingProb}_{\text{Antipyrine}} = 0.35$$

$$P_{C2} = \text{SoluteBindingProb}_{\text{Labetalol}} = 0.6$$

$$P_{C3} = \text{SoluteBindingProb}_{\text{Diltiazem}} = 0.5$$

$$P_{C4} = \text{SoluteBindingProb}_{\text{Atenolol}} = 0.35$$

Consequently,

$$\text{SoluteBindingProb}_{\text{Prazosin}} = 0.371 \times 0.35 + 0.118 \times 0.6 + 0.191 \times 0.5 + 0.320 \times 0.35 = 0.4082 \text{ (rounded to 0.41)}$$

2. For propranolol:

PCP Subgroups III & V	Atenolol	Antipyrine	Labetalol	Diltiazem	Propranolol
fuB	0.74	0.6	0.52	0.28	2
HBD count	3	0	4	0	3
HBA count	4	2	4	5	0
Tautomer count	2	0	7	0	0.45

**Steps 1-3.** Let  $q = 4$ , and  $S_{new} = \{\text{atenolol, antipyrine, labetalol, diltiazem, propranolol}\}$ .

For  $q = 4$  and the values above, propranolol is in the same group with *at least* one other member, and  $G = 1$ .

**Step 4.** The classification results are:

Cluster	Atenolol	Antipyrine	Labetalol	Diltiazem	Propranolol
C4	<b>0.989</b>	0.006	0.000	0.002	0.325
C1	0.004	<b>0.977</b>	0.000	0.004	<b>0.376</b>
C2	0.003	0.002	<b>1</b>	0.000	0.045
C3	0.003	0.014	0.000	<b>0.993</b>	0.254

$$P_{C1} = \text{SoluteBindingProb}_{\text{Antipyrine}} = 0.5$$

$$P_{C2} = \text{SoluteBindingProb}_{\text{Labetalol}} = 0.6$$

$$P_{C3} = \text{SoluteBindingProb}_{\text{Diltiazem}} = 0.35$$

$$P_{C4} = \text{SoluteBindingProb}_{\text{Atenolol}} = 0.35$$

Consequently,

$$\text{SoluteBindingProb}_{\text{Propranolol}} = 0.376 \times 0.5 + 0.045 \times 0.6 + 0.254 \times 0.35 + 0.325 \times 0.35 = 0.4176 \text{ (rounded to 0.42)}$$

TABLE S1

Description of ISL parameters and their values for the simulations in Figs. 5-7 and S3, except where noted otherwise

ISL Parameters <sup>a</sup>	Description	Values <sup>b</sup>	Ranges <sup>c</sup>
<b>Device Framework Parameters</b>			
<i>cycleLimit</i>	(B) <sup>d</sup> Provides the simulation with an artificial stopping criterion so that it will not run forever	200	200
<i>monteCarloRuns</i>	(I) <sup>d</sup> The number of runs in a Monte-Carlo set. Aggregate measures for the whole system are calculated at the end of all Monte-Carlo runs	48	[2, 50]
<i>similarityMeasure</i>	Specifies which similarity measure to use.	global_sd	global_sd
<i>nominalProfile</i>	(S) <sup>d</sup> Specifies which model to use as the nominal for use in calculating similarity between the experimental and nominal models		
<i>experimentalProfile</i>	(S) Specifies which model to use as the experimental for use in calculating similarity between the experimental and nominal models		
<b>In Silico Liver Parameters</b>			
<i>StepsPerCycle</i>	(I) Parameterizes the granularity of time for the Research model (the ISL). The data has a “sampling rate” that provides a default time scale for it. Finer granularity requires interpolation between the data points. The Reference (PK) model <sup>a</sup> is a time-reversible, continuous function, which allows sampling at any frequency	2	2
<i>GraphInputFile</i>	(S) Specifies the file to read if the SS graph is to be specified by an explicit graph (file format is GML)	null	
<i>GraphSpecFile</i>	(S) Provides the lobule graphical specification, and specifies the base file name ( <i>extension.ls</i> ) to be used if the graph is to be specified according to the Lobule Specification		
	Nodes in Zone I	45	[20, 50]
	Nodes in Zone II	21	[10, 35]

	Nodes in Zone III	6	[3, 15]
	Intra-Zone I edges	18	[10, 30]
	Intra-Zone II edges	10	[5, 25]
	Intra-Zone III edges	0	0
	Zone I $\rightarrow$ Zone II edges	15	[2, 45]
	Zone I $\rightarrow$ Zone III edges	0	[0, 20]
	Zone II $\rightarrow$ Zone III edges	8	[5, 25]
<i>GraphSpecIterates</i>	(I) Tells the framework to modify LOBULE specification and run a Monte-Carlo set (consisting of $N$ runs) for each different LOBULE specification. Set to 1, it runs 1 set and provides 1 set of outputs. Set to 5, the first run uses the current contents of <i>lobule.ls</i> ; it then runs 4 more sets, slightly modifying the LOBULE specification each time, resulting in 5 sets.	1	[1, 10]
<i>DirSinRatio</i>	(F) <sup>d</sup> Specifies the percentage of SS that are type $S_A$ (“direct”)	0.75	[0, 1]
<i>TortSinRatio</i>	(F) Specifies the percentage of SS that are $S_B$ (“tortuous”)	0.25	[0, 1]
<i>DirSinCircMin</i>	(I) Sets the upper and lower bounds for a pseudo-RNG (random number generator) using the uniform distribution to choose a circumference for each SS	50	[1, 50]
<i>DirSinCircMax</i>		50	[1, 100]
<i>TortSinCircMin</i>		4	[1, 8]
<i>TortSinCircMax</i>		4	[1, 20]
<i>DirSinLenAlpha</i>	(F) Sets the parameters for a pseudo-RNG using a modified form of the Gamma distribution; The modification consists of a left-right shift of the distribution, allowing the user to clip off the front	3.0	[0.5, 5.0]
<i>DirSinLenBeta</i>		0.215	[0.1, 0.325]
<i>DirSinLenShift</i>		0.0	0.0
<i>TortSinLenAlpha</i>		10.0	[4, 18]
<i>TortSinLenBeta</i>		0.125	[0.03, 0.15]
<i>TortSinLenShift</i>		-40	[40.0, 40.0]
<i>SinusoidTurbo</i>	(F) The complement of the amount of turbulence allowed in any given SS. Lower turbo means more tendency of any one COMPOUND to wander sideways or backwards. Higher Turbo means a stronger flow from the input to the output of the SS	0.3	[0.05, 0.95]
<i>ECDensity</i>	(F) Specifies the relative ENDOTHELIAL CELL density, given that the number of grid points in Grid B is set by the geometry parameters of the SS; it specifies the percentage of Grid B points that index an ENDOTHELIAL CELL	0.8	[0.05, 1.0]
<i>HepDensity</i>	(F) Specifies the relative HEPATOCYTES density within Grid C, given that the number of grid points in Grid C is set by the geometry parameters of the SS, it specifies the percentage of Grid C points that index a HEPATOCYTE	0.6	[0.05, 1]
<i>CoreFlowRate</i>	(I) The number of slots solute in the SS core moves forward during each step	2	[1, 8]
<i>A2BJumpProb</i>	(F) Specifies the probability that, when given the option, SOLUTE will jump from Grid A to Grid B	0.1	[0.05, 1.0]
<i>B2AJumpProb</i>	(F) Specifies the probability that, when given the option, SOLUTE will jump from Grid B to Grid A	0.6	[0.05, 1.0]
<i>B2CJumpProb</i>	(F) Specifies the probability that, when given the option, SOLUTE will jump from Grid B to Grid C	0.35	[0.05, 1.0]
<i>C2BJumpProb</i>	(F) Specifies the probability that, when given the option, SOLUTE will jump from Grid C to Grid B	0.65	[0.05, 1.0]

<i>BindersPerCellMin</i>	(I) Max and min for a uniform pseudo-random draw for the number of binding agents inside each CELL. Simple binders for ECs and ENZYMES for HEPATOCYTES	5	[1, 30]
<i>BindersPerCellMax</i>		10	[10, 100]
<i>MetabolizeProb</i>	(F) Probability that an ENZYME will metabolize a SOLUTE before it releases it	0.4	[0, 1]
<i>SoluteBindingProb</i>	(F) Probability that, when a BINDER and SOLUTE make contact, the SOLUTE will be bound	0.5	[0, 1]
<i>SoluteBindingCycles</i>	(I) Number of cycles a binder holds a SOLUTE.	25	[5, 100]
<i>isMembraneCrossing</i>	(S) Specifies whether the SOLUTE can cross membrane or not	SUCROSE:NO DRUG: YES	YES NO
<i>ISL2WetLabScaling</i>	(F) Provides the precise validation mapping from ISL output to the wet-lab (IPRL) output fraction	7.0	[1.0, 8.0]
<i>BolusStartTime</i>	(I) Dictates when to let SOLUTES flow into the LOBULE	5	[2, 9]
<i>DosageParamA<sup>e</sup></i>	(F) Parameter (A) of the dosage function: $D(t) = A(B^C t^{(C-1)} e^{-Bt}) / (C - 1)!$ , where t = current cycle; Parameter (A) simply raises the amplitude of the function	2000	[1000, 7000]
<i>DosageParamB</i>	(F) Parameter (B) of the dosage function	1	[1, 2]
<i>DosageParamC</i>	(F) Parameter (C) of the dosage function	2	[2, 3]
	Actual dose for Figs. 5-7 and S3.	3,682	

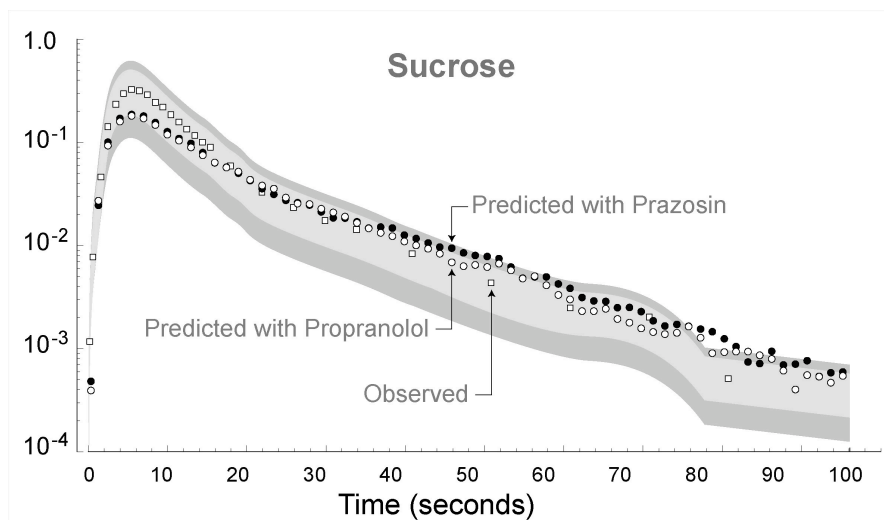
<sup>a</sup>There are four classes of parameters: Device Framework, Research Model, Reference Model, and Data Model. Here, the Research Model is the ISL. Only a subset of the Device Framework parameters is listed here. The Data Model includes all the data against which the SM is being applied, and uses a parameter that specifies whether to interpolate between observations of the in silico data. The Reference Model is a traditional PK model previously fit to the in silico experimental data; it is run concurrently with the Research Model.

<sup>b</sup> Parameter values when SUCROSE and ANTIPYRINE were dosed in combination.

<sup>c</sup> Ranges from which values were drawn during searches of model and parameter space.

<sup>d</sup> B: Boolean; F: floating; I: integer; S: string

<sup>e</sup> The total dose is the area under the dosage function curve.



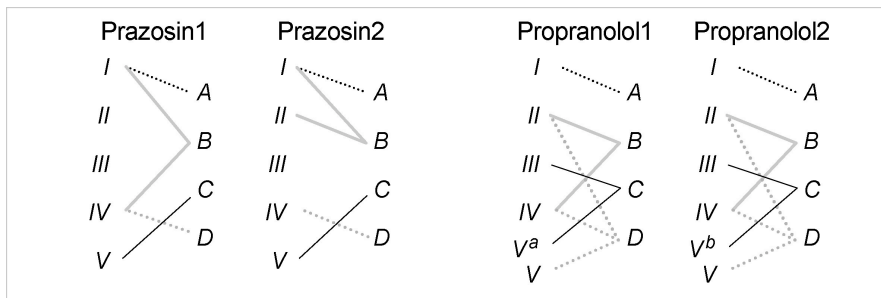
**Figure S1.** Expected outflow profiles for sucrose co-administrated with prazosin and propranolol using predicted parameter values. These semilog plots show results of simulation experiments for which the PCP-sensitive, PVs were predicted using the Method 2 values listed in Table 5. Black circles: SUCROSE co-administrated with PRAZOSIN; open circles: SUCROSE co-administrated with PROPRANOLOL. Otherwise, the graph components are same as for Fig. 5.

TABLE S2

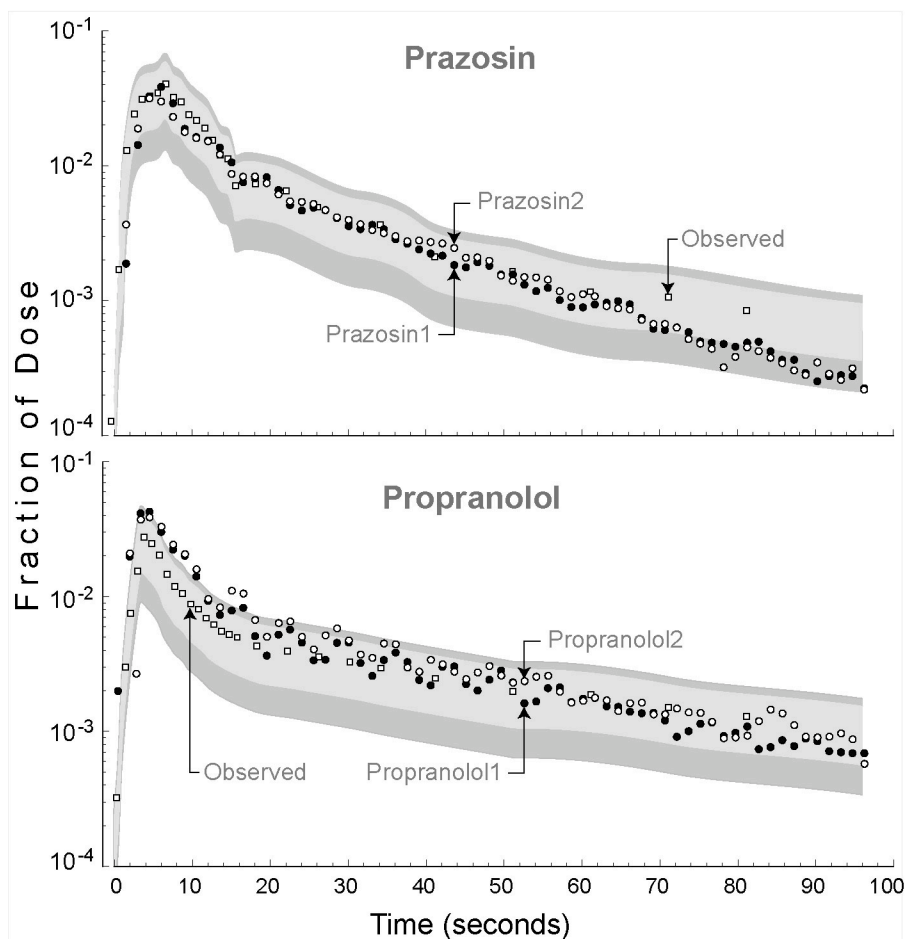
*Two sets of predicted ISL parameter values for prazosin and propranolol calculated using the two, different, FCMA-generated, drug-specific prediction strategies illustrated in Fig. S2.*

The listed parameter values (as for Table 5) were combined with the PCP-insensitive parameter values. The resulting ISLs were used, as for Figs. 5-7, to generate the expected outflow profiles in Fig. S3

ISL parameter values	FCMA Drug-specific Prediction Strategies			
	Prazosin1	Prazosin2	Propranolol1	Propranolol2
<i>ISL2WetLabScaling</i>	3.0	3.0	7.0	7.0
<i>A2BJumpProb</i>	0.37	0.54	0.86	0.86
<i>B2AJumpProb</i>	0.21	0.32	0.21	0.21
<i>B2CJumpProb</i>	0.50	0.44	0.50	0.50
<i>C2BJumpProb</i>	0.49	0.39	0.22	0.22
<i>MetabolizeProb</i>	0.23	0.23	0.12	0.12
<i>BindersPerCellMin</i>	6	6	5	6
<i>BindersPerCellMax</i>	11	11	10	13
<i>SoluteBindingProb</i>	0.40	0.40	0.56	0.46
<i>SoluteBindingCycles</i>	24	24	24	24
SM values ( $\pm 1$ SD)	0.76	0.73	0.85	0.70
SM values ( $\pm 1.5$ SD)	0.97	0.97	0.88	0.85



**Figure S2.** Mappings between PCP subgroups and ISL parameter subgroups (as in Fig.4). The four ISL subgroups are specified in Table 1; the five PCP subgroups are specified in Table 2. Each line represents a hypothesized strong influence. Except as follows, the identified PCP subgroups are the ones used by the FCMA to provide the membership degree values used in Eq. 1 to predict ISL parameter values for prazosin and propranolol in Table S2, exactly the same as done for Table 5.  $V^a$ : only HBA from group V was used;  $V^b$ : only Tautomer count from group V was used. The FCMA treats the identified PCPs as having equal influences. The influence of the other PCPs is neglected.



**Figure S3.** Expected outflow profiles for prazosin and propranolol using predicted parameter values. These semilog plots show results using the PCP-sensitive PVs in Table S2. Except as noted, the graph components are same as specified in Fig. 5.