# **Predicting CYP Reactivity and Sites of Metabolism in ADMET Predictor**<sup>®</sup>

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Please note: this presentation, including questions from the audience, is being recorded and may be made available.







# **Drug Metabolism in Early Drug Discovery**

- Drug discovery is timeconsuming, expensive and laborintensive
- Up to 25% of compounds withdrawn due to metabolic, pharmacokinetic, or toxic problems
- Metabolic liability can lead to several diverse issues

Which parts of molecule are subject to metabolic reactions

- Modify portions of new molecule to modulate its metabolism to improve its safety and efficacy
- Crucial for rational drug design to mitigate ADME/toxicity issues

Which enzymes can metabolize a newly designed molecule

• Drug-Drug Interactions predicted and modified to improve drug's applicability

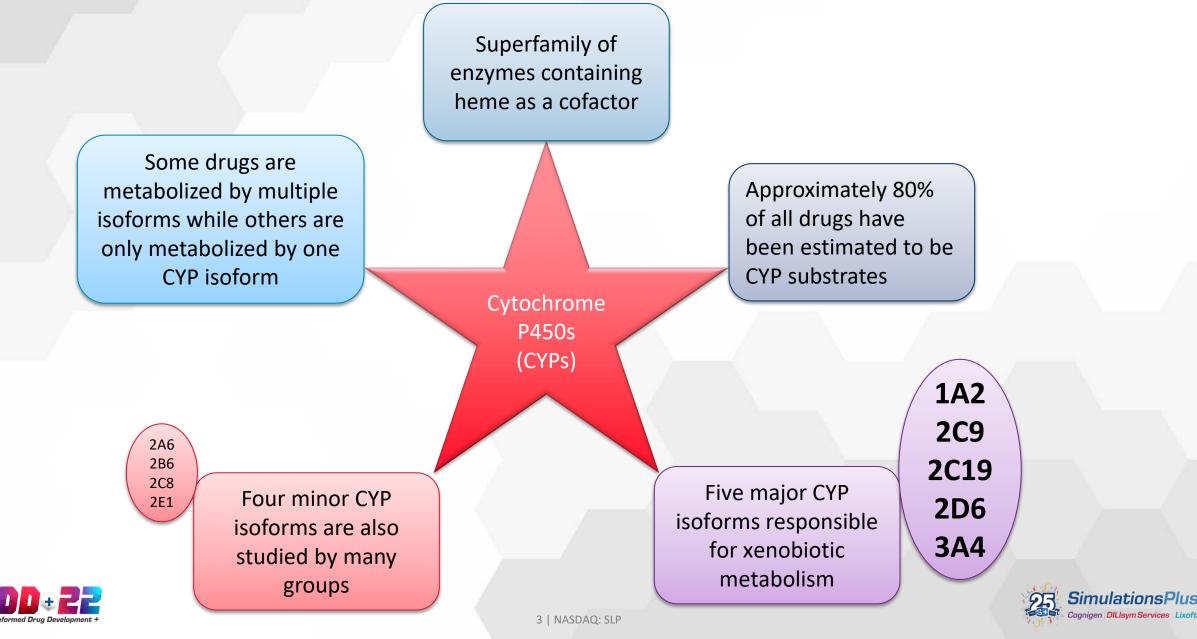
Metabolized into a Toxic/Non-toxic product

- Will have a low therapeutic index
- Need to be administered carefully





### Cytochrome P450s (CYPs)



# **Computational CYP Models**

- Experimental identification of SoMs or metabolites
  - expensive and time-consuming task
- *In silico* prediction of SoMs and metabolites
  - Reduce costs and time needed

#### Rule-based

- Derive likely metabolites and SoMs by applying a dictionary of biotransformation rules
- Compiled by human experts

#### Structure-based

- Focus on the substrate-protein interaction
- Requires high quality crystal structures and very time-consuming

#### Ligand-based

- Employ machine learning and atom-level descriptors
- Rely heavily on available experimental data on SoMs and metabolites



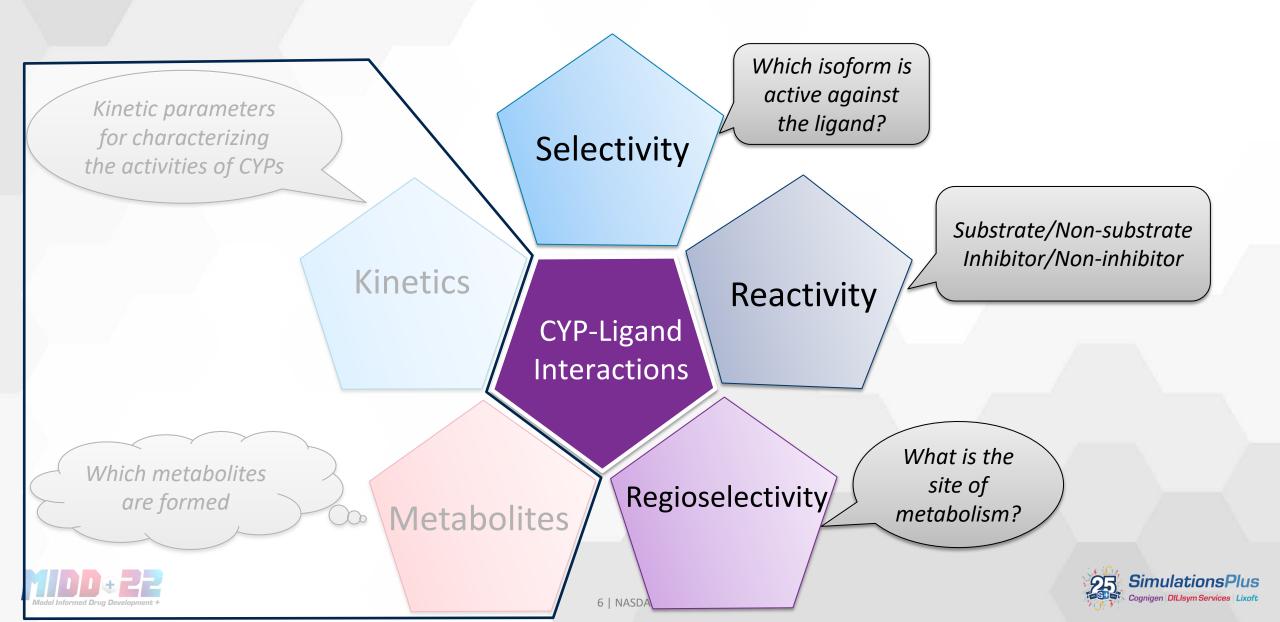


# **CYP-LIGAND INTERACTIONS IN ADMET PREDICTOR<sup>®</sup>**

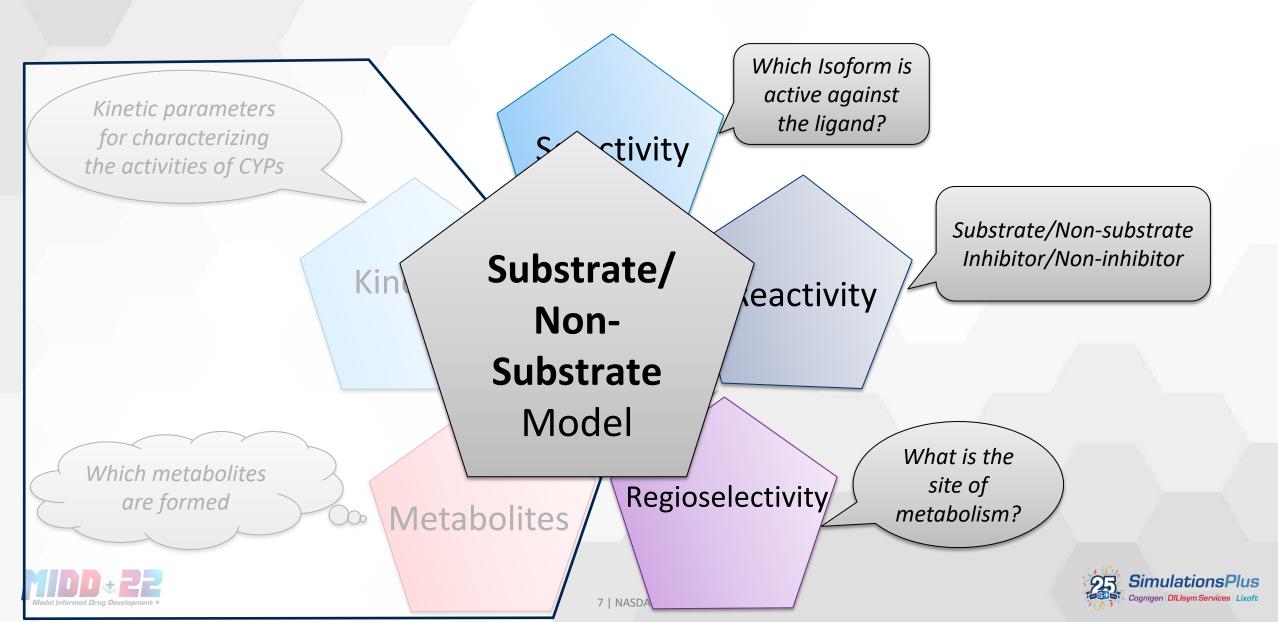




# What does ADMET Predictor<sup>®</sup> offer?



# **Selectivity : Reactivity Combined**



# **CYP Substrate Models: The Data**

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- CYP Substrate classification models for nine CYP isoforms
- Models built using data from various sources
  - Compiled datasets from literature (Sheridan et al, Zaretzki et al)
  - Commercial and academic databases
  - FDA review submissions
  - Literature
- Data included specific metabolites as well as cases where no metabolite is reported

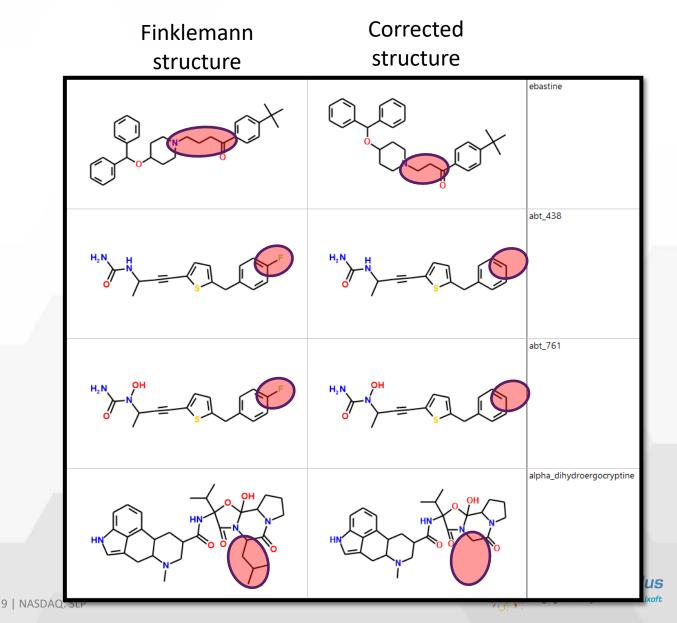
3A4	1A2
2E1	2A6
2D6 2C19	286
20	2C8

S Enzyme	S Object	S Object Metabolite
CYP2D6	(S)-gallopamil	(S)-gallopamil disappearance
CYP3A4	ebastine	ebastine disappearance
CYP2C19	tacrolimus	tacrolimus disappearance
CYP3A4	E2101	E2101 metabolite M4
CYP3A4	E2101	E2101 metabolite M1
CYP2C19	siponimod	siponimod disappearance
CYP2D6	(S)-fluoxetine	norfluoxetine
CYP3A4	dydrogester	dydrogesterone disappearance
CYP2C19	sildenafil	N-desmethyl sildenafil (UK-103 320)



# **CYP Substrate Models: Data Curation**

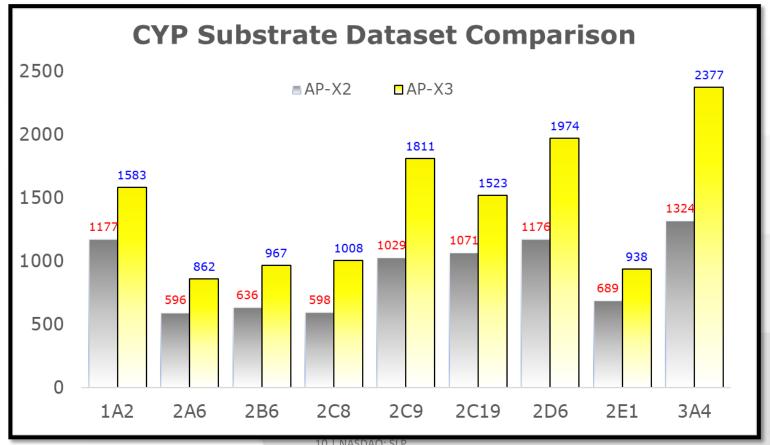
- Correcting chemical structures
- Updating correct site of metabolism
- Merge datasets from different sources
  - Remove identical duplicates
  - In case of extra SoM available, combine sites
- Exclude epoxidation sites





## **CYP Substrate Models: Final Datasets**

- 2931 cmpds in CYP Substrate/Non-Substrate dataset
  - At least ~35% larger dataset compared to previous SLP dataset
  - Average increment: ~55%



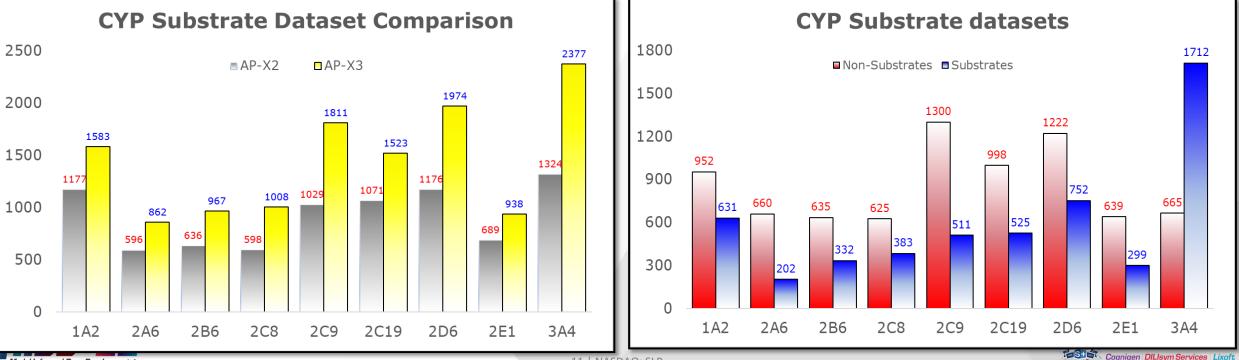


## **CYP Substrate Models: Final Datasets**

- Comparison with our previous dataset
  - At least ~35% larger dataset compared to previous SLP dataset
  - Average increment: ~55%

#### Composition of each dataset

Substrate Vs Non-Substrates

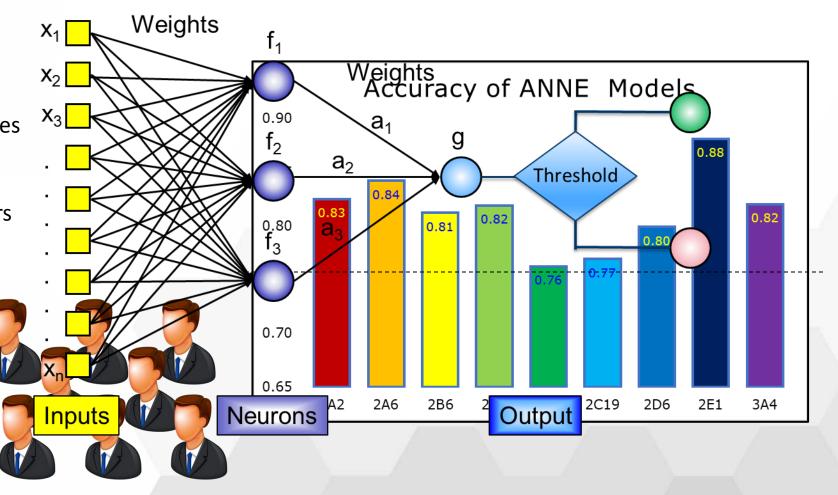


Model Informed Drug Development +

# **CYP Substrate Models: Approach**

#### **Artificial Neural Network Ensemble**

**Constitutional Descriptors Topological Indices Electrotopological State Indices** Charge-based Descriptors Hydrogen Bonding Descriptors **Ionization Descriptors Functional Groups** Moriguchi Descriptors /s



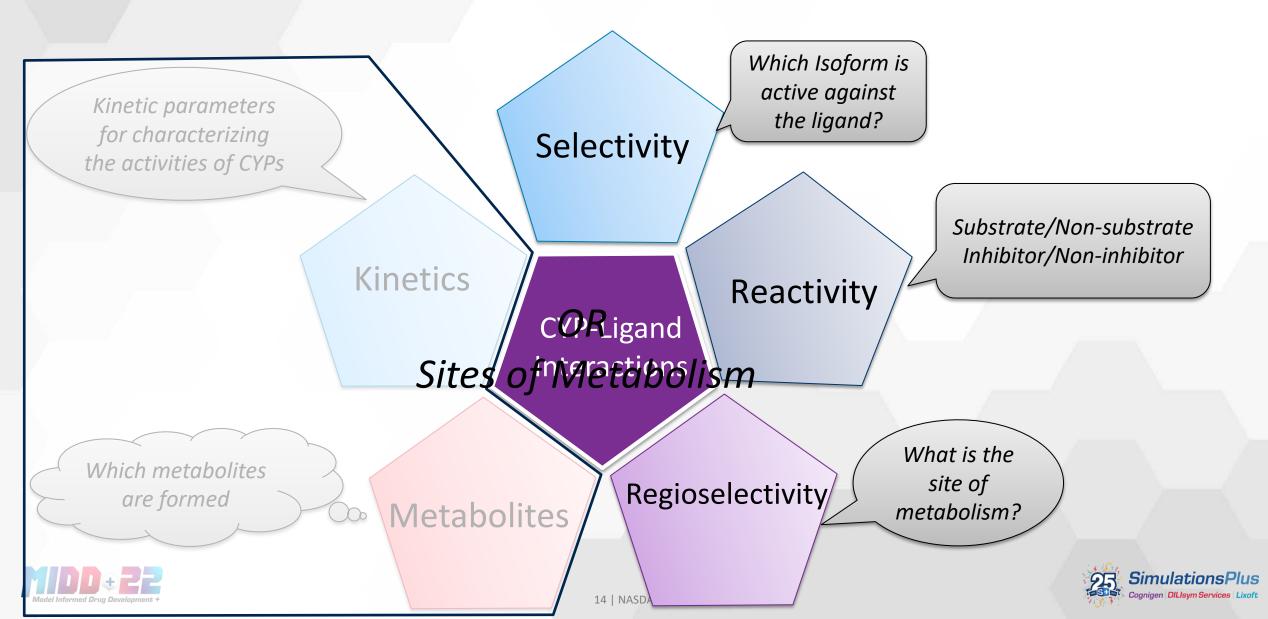


# **Final Output**

	1.1		CUDAND C. L.								
Structure	Identifier	*CYP Substr*	CYP1A2 Substr			CYP2C8 Substr	CYP2C9 Substr	CYP2C19 Substr		CYP2E1 Substr	CYP3A4 Substr
	Lorlatinib		Yes (66%)	No (98%)	No (89%)	Yes (56%)	No (65%)	No (65%)	No (95%)	No (98%)	Yes (98%)
	Chloramphenicol		No (54%)	No (84%)	No (98%)	No (54%)	No (98%)	Yes (67%)	No (95%)	No (98%)	No (44%)
	O-deethyl-Loratid		No (75%)	No (98%)	No (63%)	Yes (91%)	Yes (66%)	Yes (57%)	No (66%)	No (89%)	Yes (89%)
	loratadine		Yes (79%)	No (94%)	Yes (89%)	Yes (91%)	Yes (66%)	Yes (67%)	Yes (56%)	No (67%)	Yes (98%)
	desloratadine		Yes (70%)	No (98%)	Yes (71%)	Yes (83%)	No (65%)	Yes (67%)	Yes (87%)	No (87%)	Yes (98%)
	Clorazepic acid		No (81%)	No (98%)	No (83%)	Yes (60%)	No (76%)	No (80%)	No (95%)	No (98%)	No (33%)

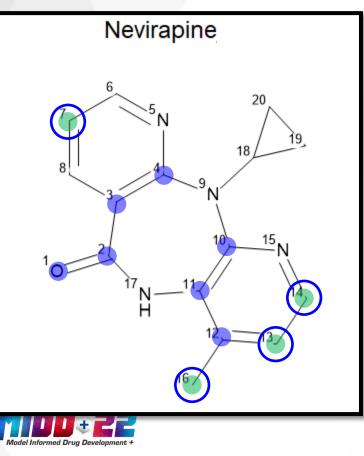


## **Regioselectivity Models in ADMET Predictor®**

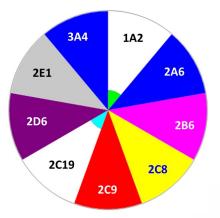


# **SoM Models : The Data**

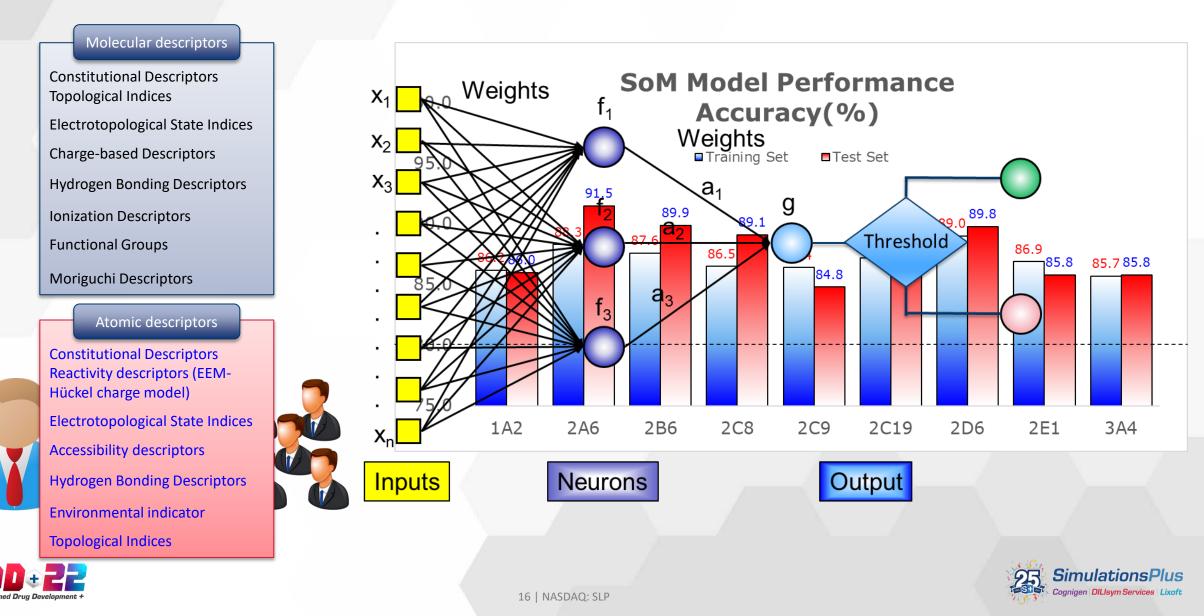
- Site of metabolism (SOM) models for nine CYP isoforms
- 1440 unique compounds
- >5500 unique metabolic sites



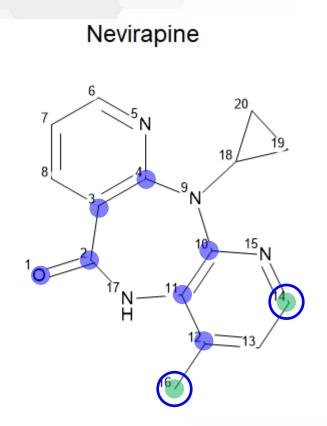
СҮР	# Molecules	Total Atoms	Excluded Atoms	Included Atoms	Negatives	Positives
	Nevirapine	20	7	13	9	4
1A2	435	8804	3769	5035	4277	758
2A6	123	1958	756	1202	999	203
2B6	201	3841	1444	2397	2064	333
2C8	208	5304	2162	3142	2776	366

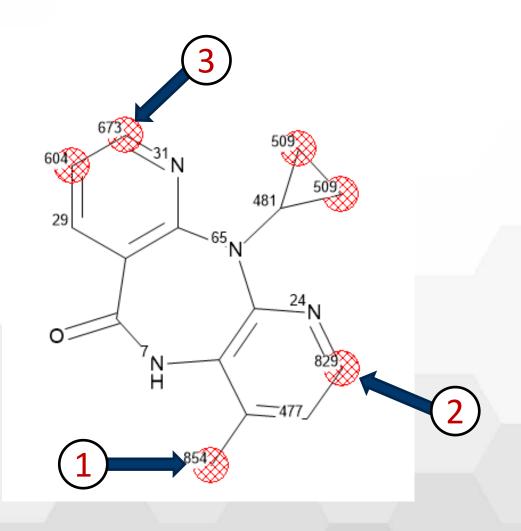


# **SoM Models: Approach**



### **Measuring Performance: Top Scoring Sites**









### **Top-two Metric to Evaluate Performance**

Table 2. XenoSite Is the Most Accurate Method for Predicting the Metabolism of the Majority of Curated CYP Substrates<sup>*a,b*</sup>

isozyme	1A2	2A6	2B6	2C8	2C9	2C19	2D6	2E1	3A4	HLM	
number of substrates	271	105	151	142	226	218	270	145	475	680	average
XenoSite <sup>c</sup>	87.1	85.7	<b>83.4</b> <sup>(1)</sup>	<b>88.</b> 7 <sup>(2)</sup>	86.7	<b>89.0</b> <sup>(3)</sup>	88.5	83.5	87.6	<b>89.4</b> <sup>(4)</sup>	87.0
RS-Predictor <sup>d</sup>	83.4	<b>85.</b> 7 <sup>(4)</sup>	82.1	83.8	84.5	86.2	85.9	82.8	82.3	86.2 <sup>(4)</sup>	84.3
SMARTCyp	80.0	86.0	77.0	83.0	84.0	86.0	83.0	82.0	78.0		82.1
StarDrop					78.0		75.3		74.1		75.8
Schrödinger					72.1		68.1		76.4		72.2
Fingerprint <sup>e</sup>	66.1	63.8	64.2	65.5	68.1	69.3	74.4	64.1	71.2	75.3	68.2
random model	26.0	31.9	24.8	22.6	22.2	20.2	21.1	36.5	21.0	26.3	25.3
	435	123	201	208	345	342	455	166	960		
<b>ADMET Predictor</b> ®	89.2	88.9	87.1	96.6	81.0	81.6	85.5	92.0	84.0		87.3



Zaretzki, et al, J. Chem. Inf. Model. 2013, 53, 3373-3383

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### **CYP Substrate Dependent Models in ADMET Predictor®**

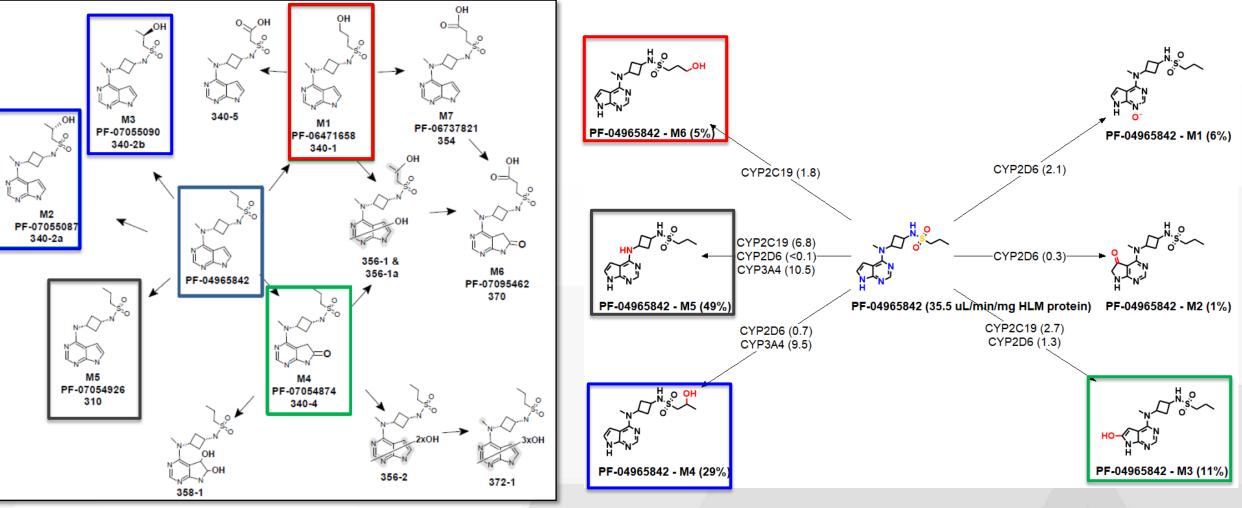
Structure	Identifier	CYP3A4_Substr	CYP3A4_Sites	CYP3A4_Km	CYP3A4_Vmax	CYP3A4_CLint
	Chlorpromazine	Yes (92%)	C20(972); C21(972); S8(962)	93.039	3.936	4.696
	Amiloride	No (84%)	NonSubstrate	NonSubstrate	NonSubstrate	NonSubstrate
	Loratadine	Yes (98%)	C4(923); C17(581); C18(543)	49.191	35.132	79.277
	Lornoxicam	No (40%)	NonSubstrate	NonSubstrate	NonSubstrate	NonSubstrate
	Chlordiazepoxide	Yes (98%)	C9(996)	43.507	6.859	17.500

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## **Case Study: Abrocitinib**

Recently approved by FDA To treat refractory, moderate-to-severe atopic dermatitis (Jan 14, 2022)





https://www.accessdata.fda.gov/drugsatfda\_docs/nda/2022/213871Orig1s000MultidisciplineR.pdf

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# Conclusions

- We recently expanded chemical applicability of CYP models
  - Nine CYP Substrate models
  - Nine CYP regioselectivity models (Sites of Metabolism)
- The new models showed higher accuracy, over 80%
- Models are applicable to a very broad range of molecules
- When combined with pre-defined set of biotransformation rules, can be successfully used to predict plausible metabolites of small molecules





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- Marvin Waldman
- Bob Clark







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