

# Evaluation of Some Models in ADMET Predictor (v 9.5) used in Early Discovery Drug Metabolism and Pharmacokinetics Project Work

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# Early Drug Discovery

- Predictions of basic compound properties (e.g. LogD, solubility, metabolic stability and permeability) from virtual structures can help prioritizing synthesis and testing and thus save cost and time for projects.
- To filter out compounds, the predictions need to be reliable.
- A few compounds with poor properties should be made and tested along with the best compounds to verify the predictions.
- **Note!** There is always a compromise to be made between phys chem, ADME, potency and toxicity.

# Dataset characteristics (4794 compounds with measured HLM CL<sub>int</sub> values)

Compounds from Medivir AB

## Mostly protease and polymerase inhibitors

(i.e. many peptidomimetics and nucleoside/nucleotide analogs)

Purity >80%, vast majority >95%

Unstable or insoluble compounds were not included

2236 Zwitterions

1888 Bases (most weak, with pKa below 7. Only 385 had pKa >8)

623 Acids (most weak, with pKa above 7. Only 52 had pKa <5)

44 Neutrals

3 Mixed pKa

## Measured quality data for

- HLM CL<sub>int</sub> 4794 compounds
- LogD 1198 compounds (from the 4794)
- Solubility 2778 compounds
- Caco-2 P<sub>app</sub> 2586 compounds

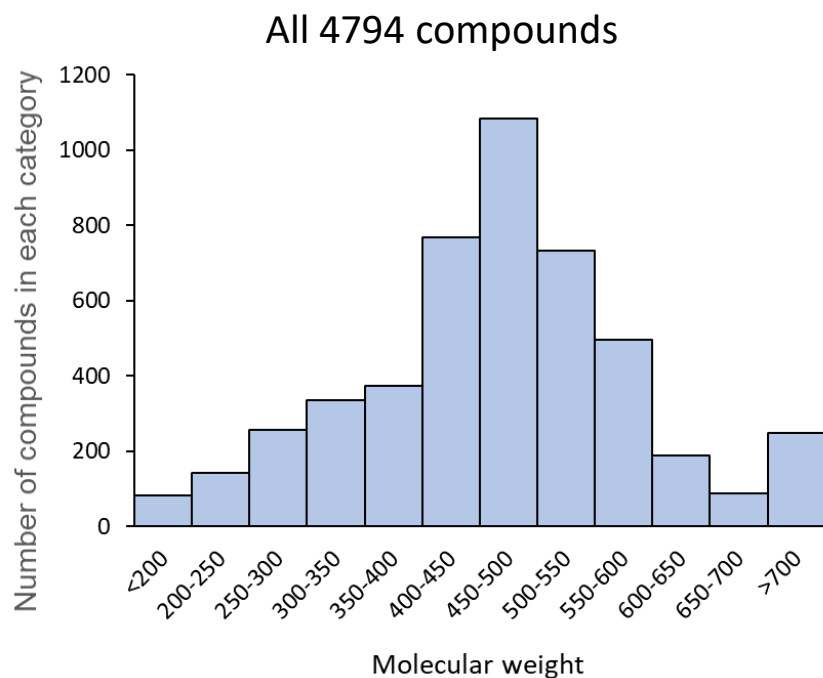
## Predicted by global AP (v 9.5) model

CYP\_HLM\_CLint  
S+LogD  
S+Sw\_pH (here pH 7.4)  
S+P<sub>eff</sub>

# Molecular weight and LogD distribution

(4794 compounds with measured HLM  $CL_{int}$  values)

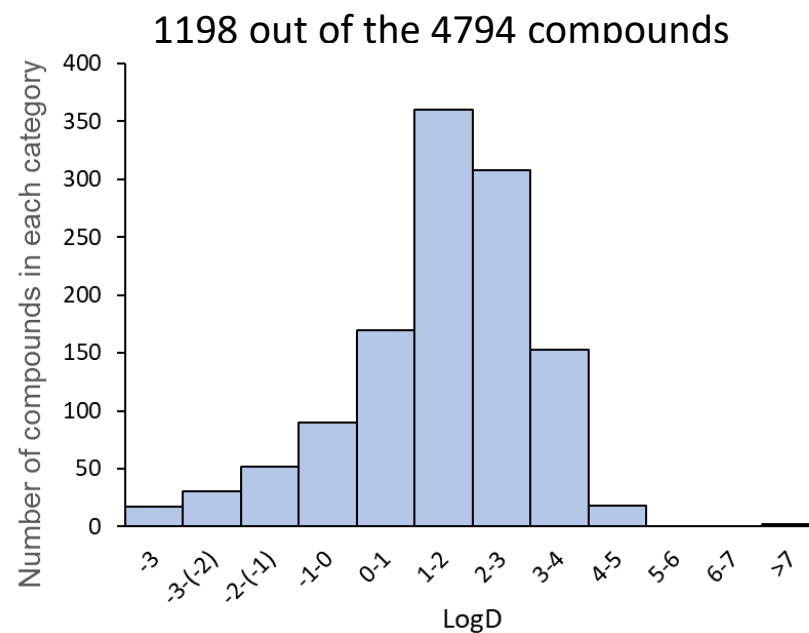
## Molecular weight distribution



Median: 467.7

Average: 468.6

## Distribution of measured LogD



Median: 1.7

Average: 1.4

# Measured data - Assays

- **LogD** Compound (15  $\mu\text{M}$  final, from 10 mM in DMSO) was vortexed with octanol/10mM phosphate buffer, pH 7.4, then allowed to settle and centrifuged.
- **Solubility** Kinetic solubility, 100-fold dilution of a 10 mM compound (in DMSO stock) in 10 mM phosphate buffer, pH 7.4. Precipitate removed by vacuum filtration. Since 100  $\mu\text{M}$  was the starting concentration in the assay, higher solubilities were reported as  $>100 \mu\text{M}$ .
- **Permeability** Caco-2 cells from ATCC were used at passage 36, seeded in 96-well plates and cultured for 21 days. Permeability from A to B was measured during 120 min after adding 10  $\mu\text{M}$  compound with 1% BSA in the basolateral buffer. Possible efflux was investigated for compounds with low permeability by blocking P-Gp (MDR1) and BCRP with 5  $\mu\text{M}$  Elacridar GF120918). For some compounds a full ABBA assay was performed. Estimated +GF/-GF ratios of  $>1.5$  in the A to B assay and efflux ratios  $>2$  in the ABBA experiments were used as indications of efflux and compared with the P-gp substrate Yes/No predictions in AP.
- **HLM total  $\text{CL}_{\text{int}}$**  1  $\mu\text{M}$  compound and 0.5 mg protein/mL in 100 mM phosphate buffer, pH 7.4. Ice-cold stop solution with losartan as internal standard was added. Protein precipitate was removed by centrifugation. A time curve 0-45 min was obtained and the disappearance of compound was fitted to a first-order elimination equation. As the  $f_u$  was not known, the AP predicted, unbound  $\text{CL}_{\text{int}}$  values were converted to total  $\text{CL}_{\text{int}}$  using the predicted  $S+f_{u,\text{mic}}$  for comparison.

# Measured data

- Reference compounds were included in all experiments as quality controls and to check for reproducibility. Data from rejected experiments were not used.
- Data was characterized into 4 bins (A-D) for practical purposes so that A and D should be clearly separated and B and C were intermediate and gave less clear answers and were not used to predict actual values. The total number of compounds in categories B plus C was around 30%.

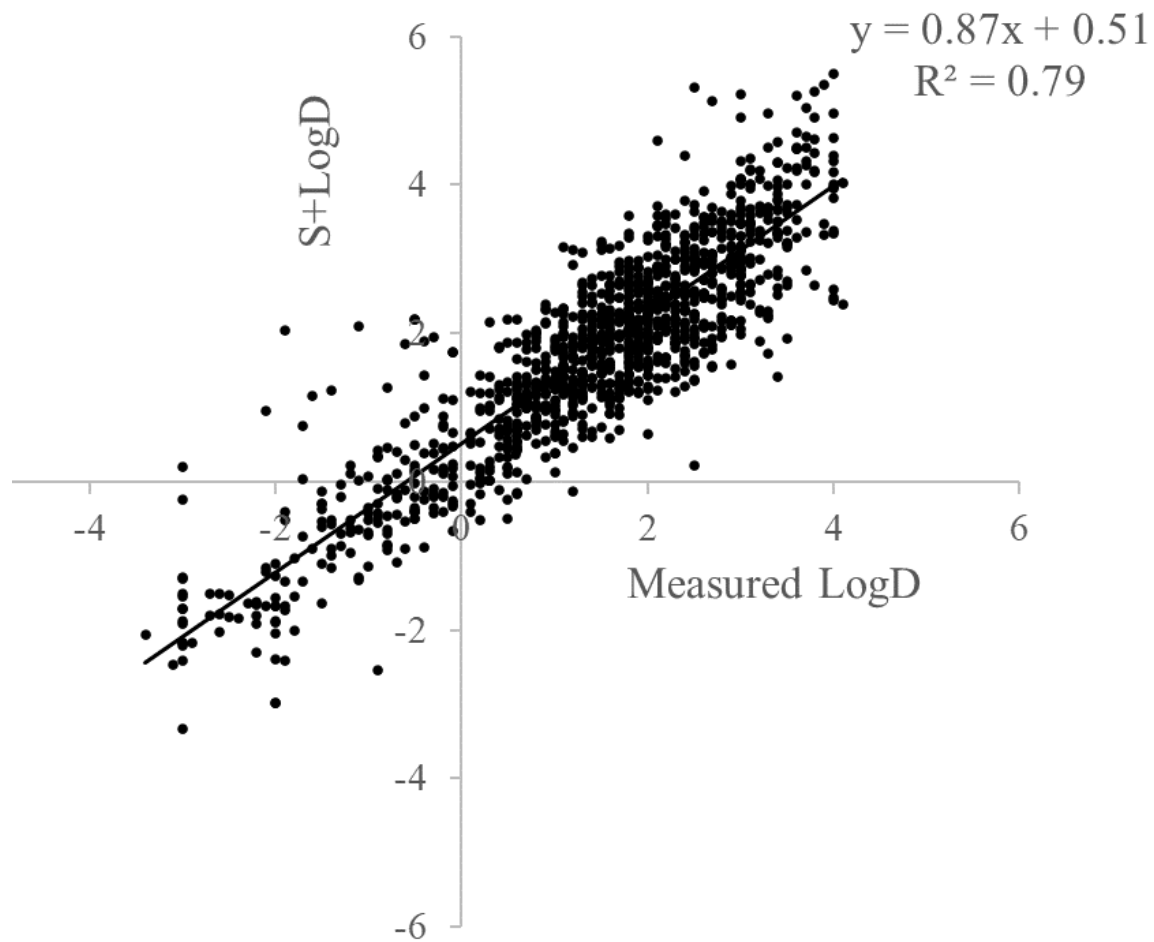
Bin definitions				
Bin	Solubility ( $\mu\text{M}$ )	Caco2 $P_{\text{app}}$ ( $10^{-6}$ cm/s)	Predicted $P_{\text{eff}}$ ( $10^{-4}$ cm/s)	HLM $CL_{\text{int}}$ ( $\mu\text{L}/\text{min}/\text{mg}$ )
A	<10	<2	<1	<15
B	10-50	2-5	1-2	15-30
C	50-90	5-10	2-3	30-80
D	>90	>10	>3	>80

Measured values			
Bin	Solubility	Caco-2 $P_{\text{app}}$	HLM $CL_{\text{int}}$
<b>Total No</b>	2778	2586	4794
<b>% in A</b>	21	37	28
<b>% in B</b>	11	12	12
<b>% in C</b>	12	15	20
<b>% in D</b>	55	36	40

## In-house models built in the AP Modeler™ module

- Models for solubility, HLM  $CL_{int}$  and Caco-2  $P_{app}$  were based on logarithmic data.
- Training sets used approximately 75-80% of the available measured data, with the remaining compounds used as test set.
- The test sets for the local model were chosen in the Modeler™, based on Kohonen mapping.
- The same test sets were also predicted with the global AP models for comparison.
- Each model was rebuilt at least 4 times and the deviations were around 10% or less.

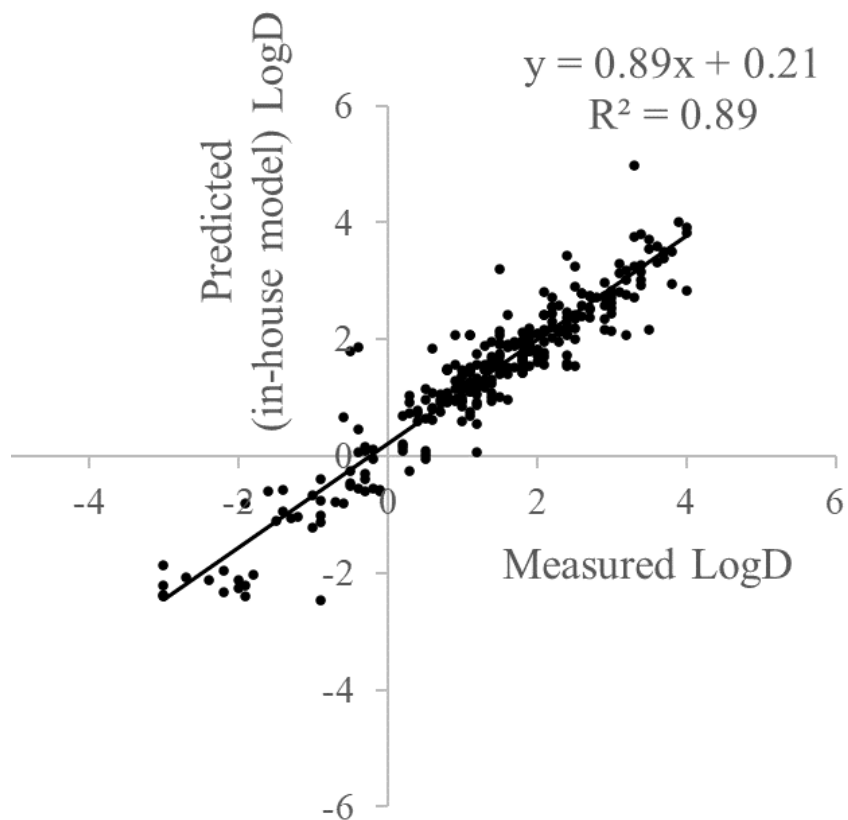
## Predicted LogD versus measured LogD using the global S+LogD model (ADMET Predictor, v 9.5) for all 1198 compounds with measured values



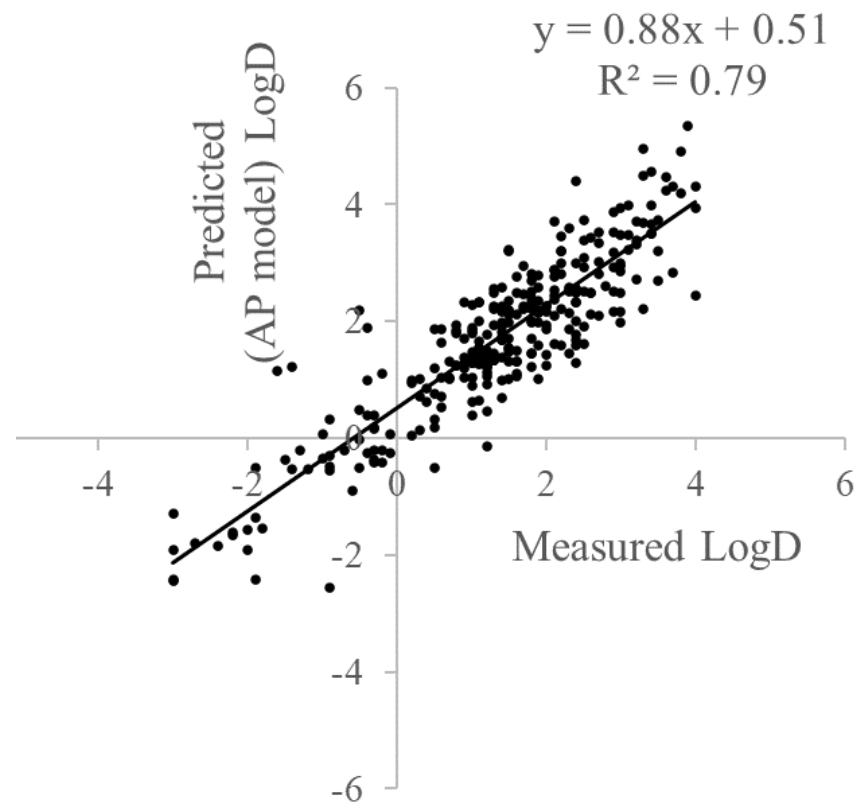


**Predicted LogD vs measured LogD using  
(A) the in-house (local) model for the test set (287 cpds) and  
(B) the S+LogD (AP) global model for the same test set.**

A)

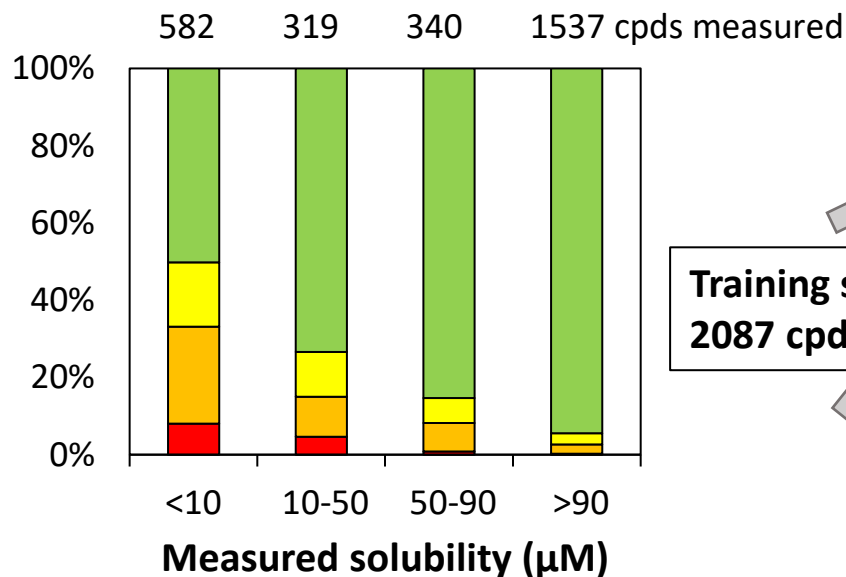


B)



# Predicted versus measured solubility (pH 7.4) for in-house compounds

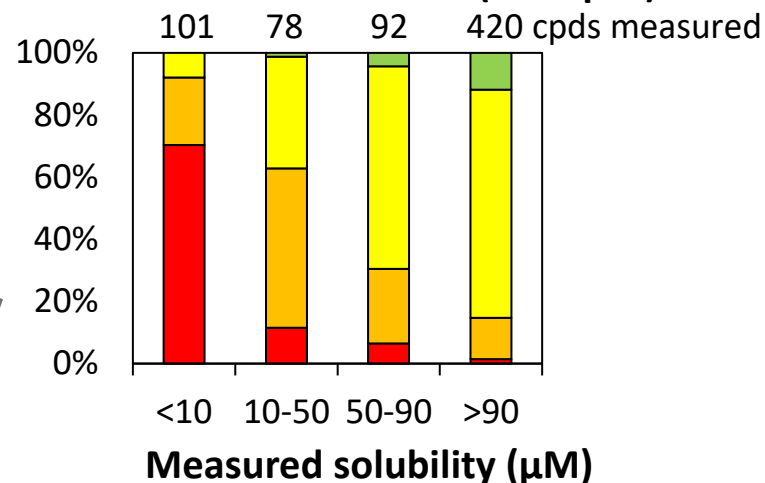
**Predictions in % of measured  
for each bin of measured values  
total: 2778 compounds (AP global model)**



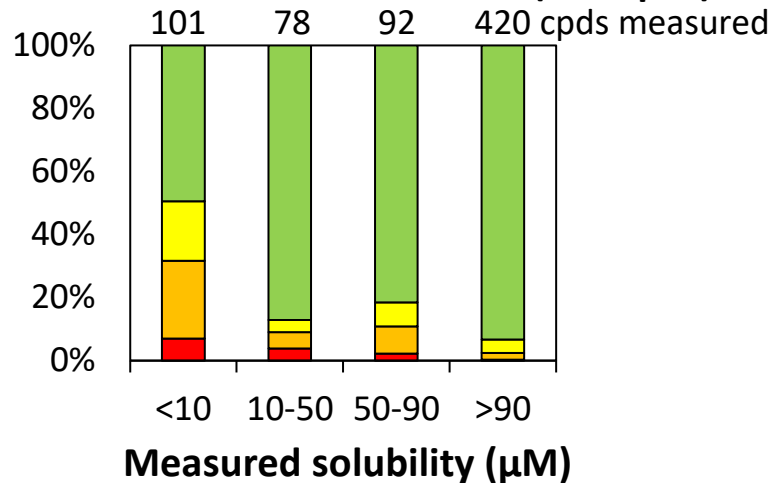
- pred >90 μM
- pred 50-90 μM
- pred 10-50 μM
- pred <10 μM

**Training set:  
2087 cpds**

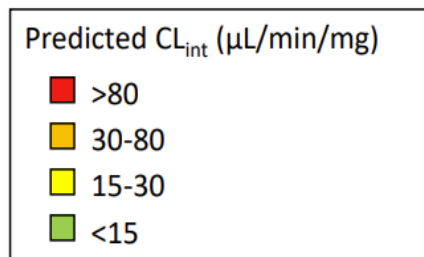
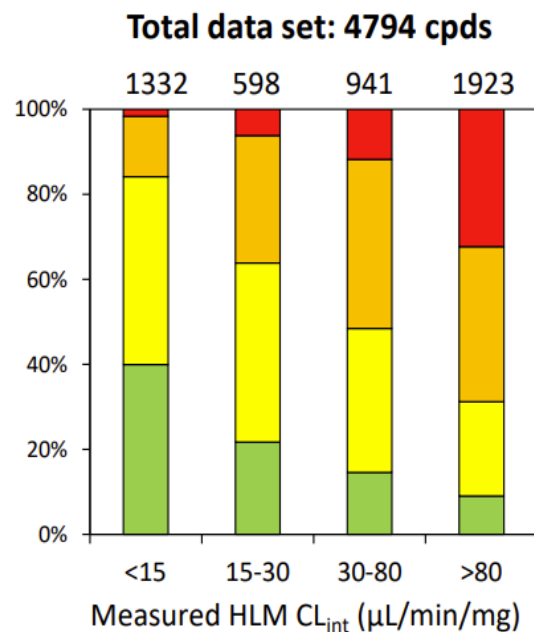
**Local model: test set (691 cpds)**



**Global AP model: test set (691 cpds)**



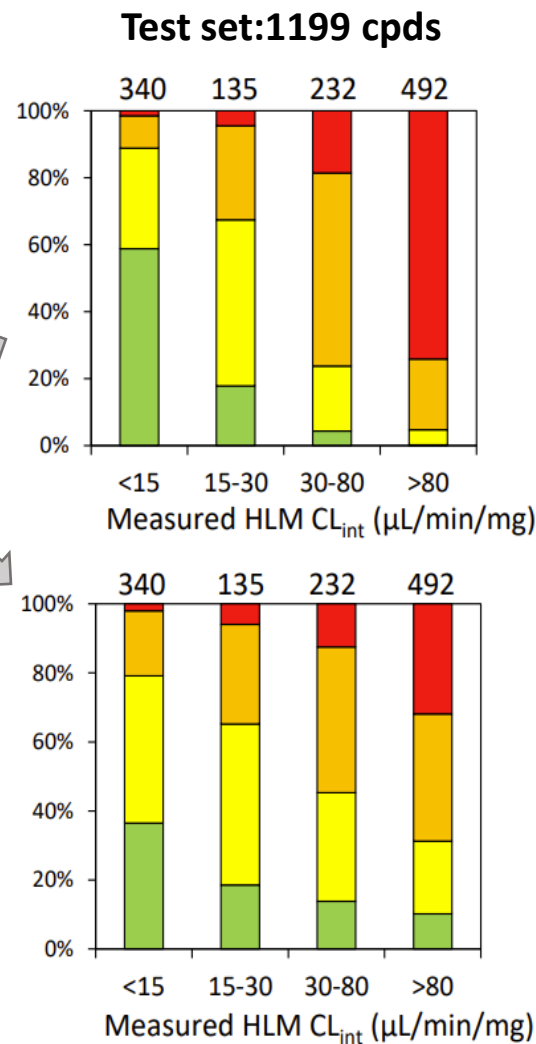
# Predicted versus measured HLM $CL_{int}$ for in-house compounds



Training set:  
3595 cpds

Local model

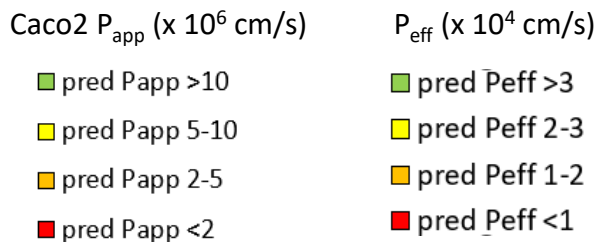
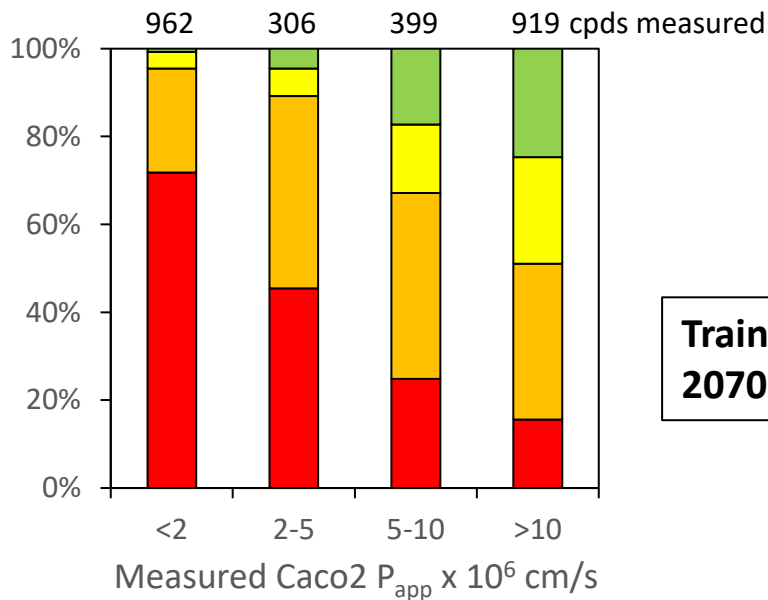
Global (AP)  
model



# Predicted versus measured permeability for in-house compounds

No Caco2 model available in AP v9.5

Predicted  $P_{eff}$  (AP model) in % of measured Caco2  $P_{app}$  for each bin: all 2586 compounds

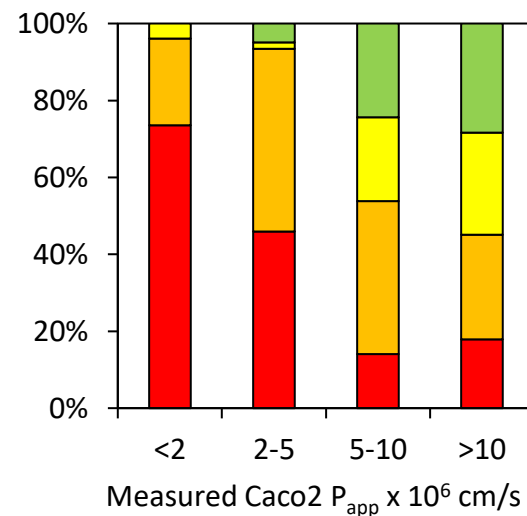
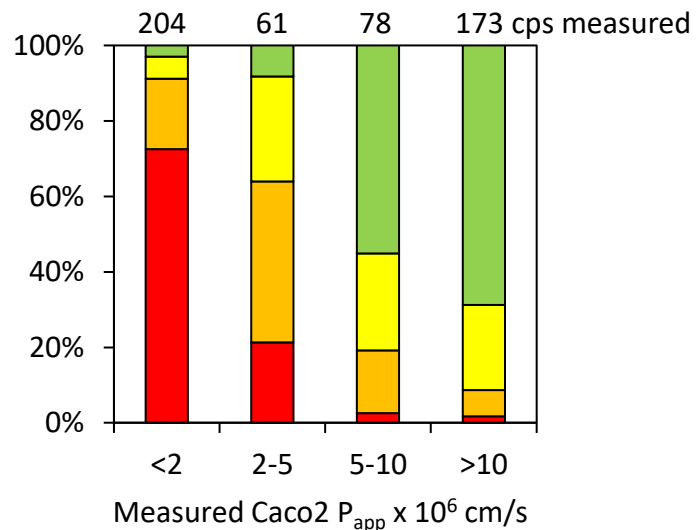


Local model  
Predicted vs Measured  
Caco2  $P_{app}$

Training set:  
2070 cps

Global model  
Predicted  $P_{eff}$  vs Measured  
Caco2  $P_{app}$

Test set: 516 cpds



# The AP model for prediction of P-gp substrate (Yes/No) evaluated against measured in-house data

387 in-house compounds were tested in the A to B +/- P-gp inhibitor assay.

A total of 264 compounds (68% of tested) had a +/- inhibitor ratio >1.5.

103 in-house compounds were tested in the ABBA assay.

A total number of 94 compounds (91% of tested) had an ABBA ratio >2.

Compounds with measured values above these ratios were considered to be true P-gp substrates.

A confusion table demonstrating the AP performance in predicting the tested compounds was constructed and resulted in an accuracy of 0.75-0.86

Caco2 assay setup	Precision	Sensitivity	Specificity	Accuracy
+/- P-gp inhibitor >1.5	0.75	0.96	0.30	0.75
ABBA >2.0	0.93	0.93	0.22	0.86

Precision: True predicted Yes/All predicted Yes

Sensitivity: True predicted Yes/All measured Yes

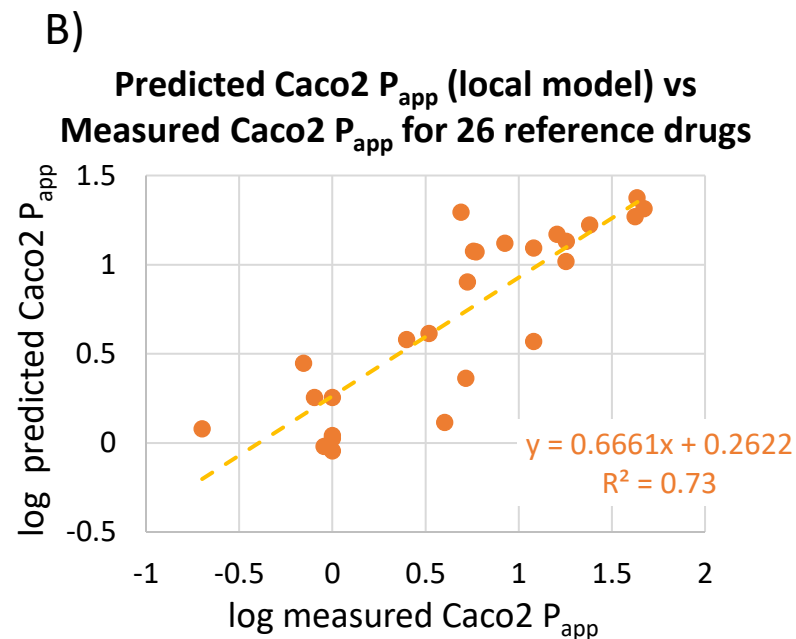
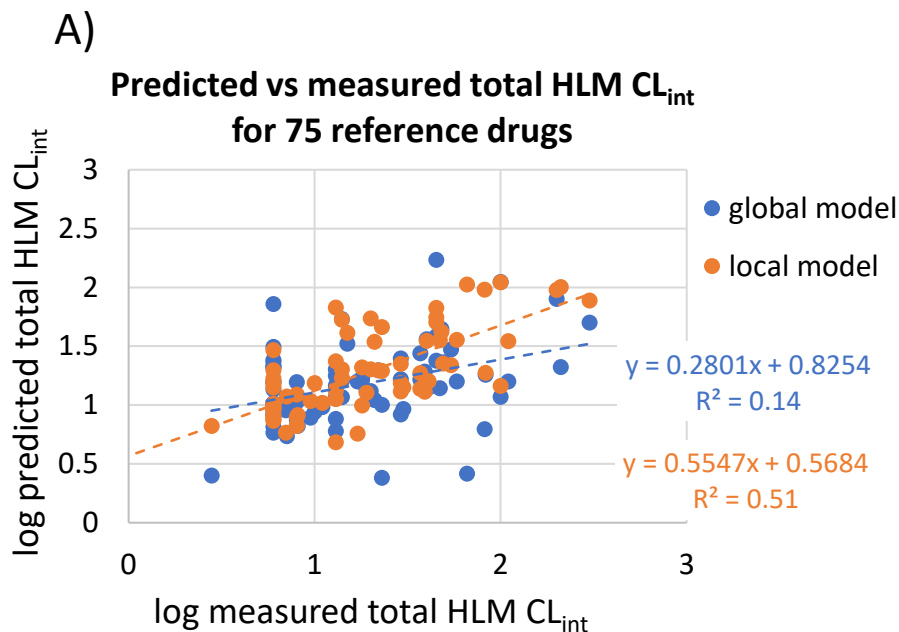
Specificity: True predicted No/All measured No

Accuracy: True total predictions (predicted Yes and No/Total measured)

The predictions picked up almost all P-gp substrates but did not pick up all negatives. The overall prediction accuracy was good but the ABBA assay had very few negatives (only 9% of compounds tested). As the +/- inhibitor assay dataset is unbalanced, the confusion table may “overestimate” precision and accuracy.

# Predicted HLM $CL_{int}$ and Caco2 $P_{app}$ for reference drugs

Measured data are the mean of at least 3 independent experiments



# Summary

Prediction outcome for the ADMET Predictor (v 9.5) models (global models) and the in-house models (local models) built with the AP Modeler™ module.

Assay	Total number of compounds	Number of compounds in Training set	Number of compounds in Test set	R <sup>2</sup>		
				Global AP		Local model
				All compounds	Test set	Test set
LogD	1198	911	287	0.79	0.79	0.89
Solubility*	2778	2087	691	0.26	0.2	0.59
HLM CL <sub>int</sub> *	4794	3595	1199	0.53	0.5	0.72
Caco2 P <sub>app</sub> *	2586	2070	516	NA	NA	0.61

\* Model based on logarithmic data

# Take-home message

**NOTE!** The dataset used here comprises mostly protease inhibitors and polymerase inhibitors, while global models are normally built on a chemically more diverse set of compounds.

Predictions can almost always be **improved by building local models** on good quality in-house data (such as a chemical series from a specific project). The separate AP Modeler™ module can be used by non-modelers to build useful local models.

However, global models can also be useful, especially when there is insufficient in-house data, i.e. when starting new projects. AP is especially useful for companies that do not have dedicated in-house modeling groups. AP can save time and money by helping to prioritize virtual compounds for synthesis and/or testing.

Reasons for improved predictions with models built with in-house data:

- **Better representation of in-house structures.**
- **All data based on the same assay conditions.**

Biological assay systems can vary quite a lot (e.g. here: the same batch of pooled HLM)

However, when using local model predictions to guide synthesis of better compounds, structures move away from the chemical space used in the model building. **This often makes it necessary to rebuild models, including the new structures in the training set.**



# Thanks to

- The former chemists and DMPK staff at Medivir
- Medivir for allowing us to use the dataset