Introduction to SAR/QSAR analysis: On-line chemical database and modelling environment (OCHEM)

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Public platform: <u>www.ochem.eu</u> Online manual: <u>http://docs.ochem.eu/display/MAN</u>

HelmholtzZentrum münchen German Research Center for Environmental Health

Table of Contents

1.	General concepts	3
1.1	Before we start	
1.2	User account creation and user login	6
1.3	Data browsers	8
1.4	Item profiles	9
2.	Using OCHEM	. 10
2.1	Compound properties browser	
2.2	Data structure	
2.3	Working with baskets and data export	
2.3	ToxAlert utility	
2.5	Set Compare utility	17
2.6	Model application	
3	Modeling framework	
3 3 1	Comprehensive modeling	
3.2	Development of a single model	
S	elect training set, machine learning method and internal validation options	. 28
S C	elect training set, machine learning method and internal validation options onfigure molecular descriptors	. 28 . 29
S C C	elect training set, machine learning method and internal validation options onfigure molecular descriptors onfigure the training method and start calculations	. 28 . 29 . 30
S C C L	elect training set, machine learning method and internal validation options onfigure molecular descriptors onfigure the training method and start calculations istributed model calculation	.28 .29 .30 .31
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S (((((((((((((((((((elect training set, machine learning method and internal validation options onfigure molecular descriptors onfigure the training method and start calculations istributed model calculation ave your model Model profile Applicability domain	. 28 . 29 . 30 . 31 . 32 33 35
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1. General concepts

In this chapter, we will learn the general concepts of the OCHEM interface How (and why) to register a new account What are the basic design components of OCHEM What are the basic elements of OCHEM

1.1 Before we start



Welcome to OCHEM! Your possible actions

Explore OCHEM data Search chemical and biological data: experimentally measured, published and exposed to public access by our users. You can also unload your data.

Create QSAR models Build QSAR models for predictions of chemical properties. The models can be based on the excerimental data oublished in our database.

Run predictions Apply one of the available models to predict property you are interested in for your set of compounds.

Screen compound ibraries against structural alerts for such endpoints as mutagenicity, skin sensitization, aqueous toxicity, etc.

Optimise your molecules

Optimise different properties for your molecules (e.g., reduce their toxicity or improve their ADME properties) using the state-of-the art MolOptimiser utility based on matched molecular pairs

Tutorials Check our video tutorials to know more about the OCHEM features.

Our acknowledgements

Feedback and help

User's manual Check an online user's manual

Check out the properties available on OCHEM OCHEM contains 1768800 experimental records for about 515 properties collected from 12437 ecrosore Melting Point logPow logBe LogL(veter) Construction LogD Construction Construction

 logP(+)
 logP(s)
 Water solubility
 LogL(biod)
 LogL(brain)
 <thLogL(brain)</th>
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Bedetholdor(upg) Bedetholdor(pag) Bedetholdor(pag) Cotrain/Cplasma [C50 Papp(Caco-2) Papp(MDCK) P(train) Oral absorption LIC 50 p(t1opp) ChertCplasma Chart(Cplasma Ckidery(Cplasma CwertCare

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Plasma protein binding Papp(HPBEC) Pendothelial(HPBEC) Papp(BBEC)

Pendothelial(BBEC) Papp rado(HPBEC) Pendothelial rado(HPBEC) Papp (SV-ARBEC) Pendothelial(SV-ARBEC)

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fraction unbound (fu) fraction ionized (fi) pKa VDss %Human OB

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 Call
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CYP450 reaction Vapor Pressure Boconcentration factor EC50 aquatic NOEC aquatic LOEC aquatic verte weree IC50 aquatic LC50 aquatic log(IGC50-1)

Henry's law constant Photolysis rate Kp Half-Life Photolysis HLp Photolysis quantum yield Half-Life Photolysis HLb Ah RBA EC50 EROD induction LC 50 LCLo EC50 Registerie Latest active users natalia: Miss. Natalia Golovina seconds ago

🙍 log in create a

amaziz: Mr. Ahmed Abdelaziz 38 minutes ago

- hodyna: Miss. Diana Hodyna about 1 hours ago
- nizamibilal1064: Mr. Bilal Nizami about 3 hours ago
- pbabokhov: Mr. Peter Babokhov about 7 hours ago

about 13 hours ago

Latest published models
McReynolds_5Avg model published by zfek
4 hours ago

agonists of PPARg qualitative model publish by amaziz

1 months ago

IC50 HIV model published by nizamibilal106 1 months ago

Pyrolysis Point model published by dan2097 1 months ago

 IogPow model published by itetko

 2 months ago

 Melting Point model published by romney

 5 months ago

LEL model published by novserj more than a year ago logERRBA (qualitative) model published by aveima more than a year ago The OCHEM installation is running on servers in HMGU. It can be installed on a virtual machine or on computers for better performance.

System requirements: For optimal performance, the host machine should have at least 8 CPU cores, 16GB RAM and about 100GB disc space.



OCHEM is a web-based platform. Users access it with a simple web browser, similar to the way they access services like Gmail or Facebook.

How to access OCHEM?

The public version of OCHEM is available at <u>www.ochem.eu</u>, but an inhouse installation can be run inside a company/University and be accessible inside of the intranet only.

Which browser to use?

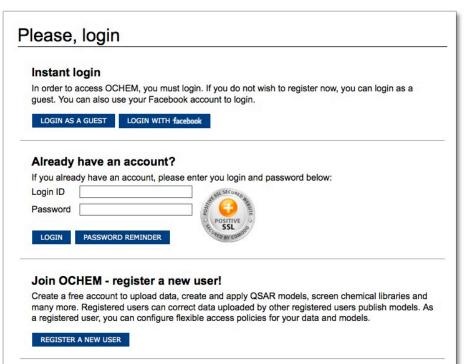
To get the best experience, it is recommended to use the latest Firefox browser. Chrome and Safari are also supported. Unfortunately, at the moment OCHEM does not fully support Internet Explorer (improves with release of new versions) and Konqueror Web browser.

User interface remarks.

OCHEM uses "tabs" extensively in the user interface. New dialogues are often conveniently opened in new tabs, as it is shown on a screenshot on the left.

Please note, that the browsers "back" button is not compatible with the tabbed interface.

1.2 User account creation and user login



In order to use the OCHEM web platform, a user has to register an account and login using this username and password.

Users can login as a guest user (with limited privileges) or as a registered user.

- **1.** The first option on the login interface is the instant login (without registration). You can log in as a guest user. For this tutorial, you are encouraged to register an account.
- **2.** If you have already created an OCHEM account, you can login using your username and password.
- **3.** The third option is to register a new account. Please, register an account if you have not done so yet.

User account Details of your personal OCHEM	account	
Registration Inform	ation	
Login*	(min. 4 characters and max. 20 characters) CHECK AVAILABILITY	
e-mail*		
Password*		
Confirm password*		
Personal Information	n	
Title*	please select Please, select a form of address!	
First name*		
Last name*		
Affiliation		
Form of organization*	please select	
City		
State		
Country		
Zip		
Phone		
Occupation		
Company		
WebSite		

In order to create a new OCHEM user account, the following information is required:

- 4. Choose your login name and check if it is available (i.e. it is not yet used by another user). Login names should be at least 4 characters long and not longer than 20 characters.
- 5. A valid e-mail address is required for the automatic notification system.
- **6.** Furthermore a password for the acount should be chosen and confirmed.
- **7.** Additional personal information like academic title, first and last name of the user and the form of organization this person is working in is required to finish the user account creation

Note:

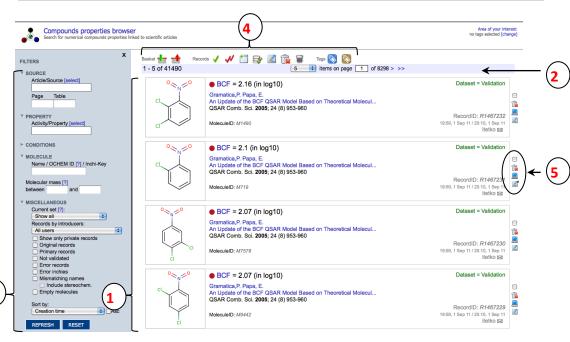
Registered users have access to more features than guest users (i.e., can upload data and develop models).

If users provide detailed information about themselves, their account will **be validated** by the OCHEM administrator. This will allow the users to run larger tasks, export more data and edit data of other validated users.

1.3 Data browsers

(5

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1 - 1	0 of 10					←	\neg
2 W	V logP Chlore	form/Water	(Dimensionless / log10)	2 records	The partition coefficient between chloroform and water	midnighter 🖂	
2 W	Ø BCF		(Dimensionless / log10)	238 records	Bioconcentration factor BCF = [ConcentrationofXinOrganism]	ExpDesign 🖂 / itetko 🖂	
🗷 W	V pKa (smile:	as ob. cond.)	(Dimensionless / log10)	376 records	this pKa requires as a condition a smiles string with the i	Koerner 🖂	
2 W	Aqueou	is Solubility dd]	(Concentration / -log(mol/L))	8402 records	Solubility of chemical compounds in water (aqueous solubili	itetko 🖂 / enamine 🖂	
🜌 W	AMES		(qualitive)	6542 records	This assay measures genetic damage at the single base level	vlad121 🖂 / itetko 🖂	
🗷 W	V log(IGC50-	1)	(Concentration / -log(mmol/L))	1093 records	The toxic potency of chemicals, measured by their concentra	itetko 🖂 / mojca 🖂	
🜌 W	V CYP450 m	odulation	(qualitive)	7485 records	CYP450 modulation describes substances in terms of their sp	vkovalishyn 🖂 / charochkina 🖂	
Z W	V logS part of Aqueor	s Solubility [x]	(Concentration / mg/L)	8402 records	Logarithm of intrinsic solubility in water of non-ionized m	Anil 🖂 / itetko 🖂	
🗷 W	V LogD		(Dimensionless / log10)	1 records	The distribution coefficient of octanol/water measured at s	mojca 🖂	
🜌 W	V logPow		(Dimensionless / log10)	17351 records	s The partition coefficient is a ratio of concentrations of u	itetko 🖂	



An important user interface element in OCHEM is a browser. OCHEM has various browsers for all kinds of database entities.

Main browsers include:

- Experimental property browser (record browser)
- Molecules browser
- Properties browser
- Conditions browser
- Units browser
- Articles browser
- Journals browser
- Baskets browser
- Tags browser
- Models browser

Main elements of a browser:

- 1. Items (e.g., data records, models, articles, tasks)
- 2. Page bars can be used to navigate items
- **3.** Filters can be specified to narrow down the displayed items to a specific area
- **4.** Global toolbars can be used either to manipulate (delete, modify) several items simultaneously, or create new items
- Item toolbars can be used to perform operations on a specific item

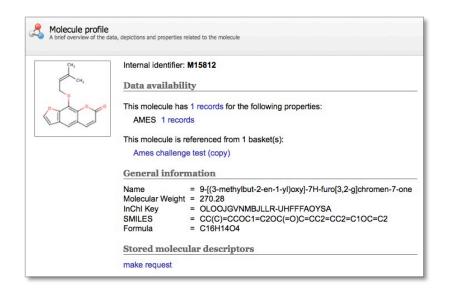
1.4 Item profiles

Overview Applica	bility domain	n					
Model name: ALogPS Public ID is 4 Predicted property: Ic Training method: AS1	ogPow	ished in Sample	e OCHEM mode	els			[OEst Correl, limit: 0.95 Variance threshold: 0 Maximum value: 100000, using L Supersab, 1000 iterations, 3 neur ensemble=100 k=0 additional par PATITION=3 SELECTION=2, PARALLE:
Data Set		#	R2	q2	RMSE	MAE	5-fold cross-valida
• Training set: ALC	OGPS 3.01	16915 records	0.951 ± 0.001	0.951 ± 0.001	0.425 ± 0.004		330 pre-filtered descrip Supersab, 1000 iterations, 3 neur
• Excluded from tra		667 records	0.62 ± 0.02	0.52 ± 0.04	2.13 ± 0.07	1.62 ± 0.05	ensemble=100 k Calculated in 14708 seco
			8 00 8 0 0				Size: 5626
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Exemplary item profiles:

- Compound property editor (record editor)
- Molecule profile
- Article profile
- Dataset ("basket") profile
- Property editor
- Unit editor



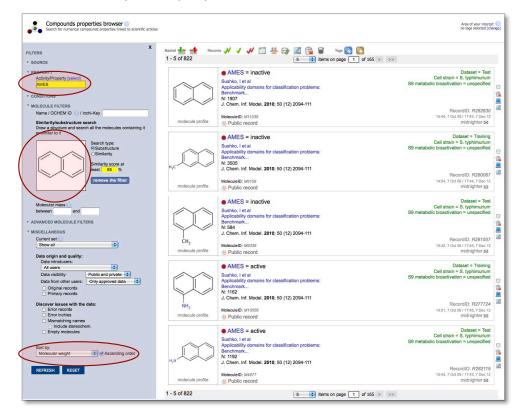


2. Using OCHEM

In this chapter, we will how to use OCHEM for several typical scenarios, including

- Search of properties using compound properties browser
- Grouping and exporting records
- Use of ToxAlert for data exploration
- SetCompare Tool for basket comparison
- Prediction of properties for molecules using models

2.1 Compound properties browser



Compound properties browser is one the main dialogues in OCHEM. It allows you to browse experimental data records using a variety of filters.

Please, try using different filters, e.g.:

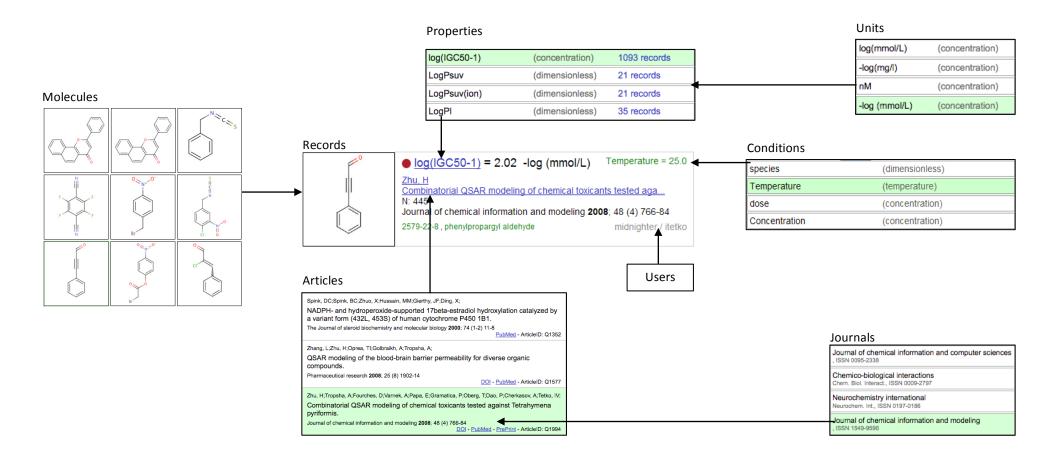
- **1.** Show only the data for the "Ames" property
- **2.** Filter the records by substructure
- 3. Sort the data by ascending molecular weight

The selected records can be put into a dataset (so called "basket"), which later can be reused for any kind of tasks, e.g. for development or validation of QSAR models.

The selected records are shown at the top of the left panel and persist until they are deselected (cleared). Clear them before selecting new records unless you would like to add them to the previous records.

Other filters of property records are available from browsers of Articles, Properties, Baskets, Models, etc.

2.2 Data structure



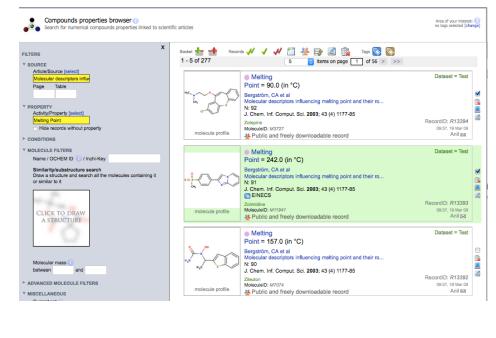
Experimental property (or "record") – a value for a property for a specific molecule published in a specific article or book.

This means that:

One molecule can have multiple records associated with it (measurements for different properties, measurements for the same property published in different articles, etc.) One article can hold multiple records for multiple properties for multiple molecules

Most of essential OCHEM operations (such as QSAR modeling) are performed on datasets of records (and not molecules)

2.3 Working with baskets and data export



6	Onlin	Online chemical database with modeling environment				
Home -	Database -	Models -				
Basket Browse, C	Compound Properties Conditions Units	properties				
Filter 1 - 1	Articles/Boo Journals ToxAlerts		•	[Create new 📸]	Show public sets	
<u>⊟ ⊠</u> 1 - 1 c		interest		created by		4 records

1. Finding interesting records

Filtering of records using filters such as article, model, property, basket or selection of records one by one, on page or all filtered

Molecule sets X Export the basket X Data export Export the selected data as an Excel, CSV or SDF file
Please, select the items that you want to export: [select all] [select unrestricted only] [select none] Structure (SMILES or SDF) RECORDID RECORDID MOLECULEID Identifier in article (N) NAMES Introducers of the records Last modifiers of the records Experimentaly measured values Experimentaly measured values Experimentaly measured values Experimentaly measured values External unique identifier Commitons of experiments Inchi-key im
Select the units to which the exported values will be converted: Melting Point Celsius
Get Excel file Get CSV file Get SDF file Get R script

3. Exporting the records

2. Exploring the baskets

2.3 ToxAlert utility

6	Online chemical database with modeling environment				
Home -	Database - Models	Moderation -			
Ex Sea mea our i	Compound proper Molecules Properties Conditions Units Articles/Books Journals	ties our possible actions L data: experimentally ed to public access by your data.			
Cre	ToxAlerts	ToxAlerts home			
Build	Pathways	View alerts			
prop	Baskets	Screen compounds against alerts			
expe	Tags	Upload new alerts			
D	Set area of interest.				

Welcome to ToxAlerts!

Structural alerts (also known as "toxicophores") are molecular patterns known to be associated with particular type of toxicity. The studies performed last decade has shown that structural alerts is an efficient technique to detect potentially toxic chemicals. Screening chemical compounds against known structural alerts can be a good practice to complement the QSAR models and to help interpreting their predictions.

ToxAlerts is a platform for screening chemical compounds against structural alerts. The platform allows to search structural alerts, introduce your own alerts and screen chemical libraries for alert-hitting compounds.

View available alerts Upload new alerts Screen

Screen your molecules

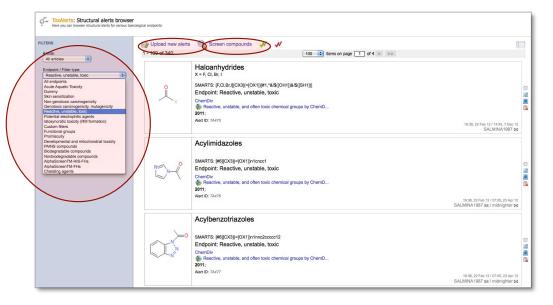
In case of any questions, ideas, or problems with the software, feel free do drop us a message. We highly appreciate any feedback from you!

Alerts are structural features that are known to be associated with a particular activity. Typical structural alerts might indicate carcinogenicity or general toxicity. Screening molecules against alerts can be simple and, most importantly, easily interpretable

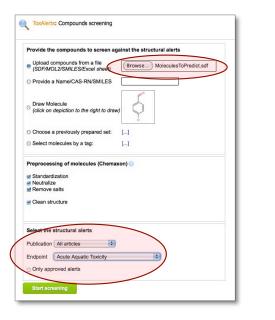
The **ToxAlert** utility allows one to screen a set of molecules against a set of structural alerts. OCHEM comes with thousands of alerts for a number of endpoints. It is also possible to introduce your own alerts

1. To get to the ToxAlert utility just select it from the menu bar.

- 2. A welcome page is the entry point for further actions, like
 - overview available alerts
 - upload new structural alerts
 - screen molecules against structural alerts



- **3.** The structural alerts browser gives an overview of the available alerts in the system.
 - Existing alerts can be filtered by their category
 - New alerts can be uploaded
 - Selected alerts can be used in a screening against a set of structures



Screening compounds against alerts

- Select the compounds you would like to screen. This can be a prepared basket from the OCHEM platform, a single structure drawn or automatically fetched by its name, or like in this example an uploaded file (SDF, Smiles, Excel).
- **5.** Optionally, you can screen against alerts for a particular endpoint or alerts from a particular publication.

$\phi^{\sigma}_{\mu} = \frac{\text{TorcAlerts: Screening results}}{\text{The compounds that instituted any alerts proped by endpoints, publications and by alerts themselves}$				
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		H, C Hydrocarbona (for Acute Aquatic Toxicity in 1992 Verhaard H.J.M.) Mexado M0777		

6. As result you can see the structures for which a particular alert was found and from which publication this alert stems.

Further information about ToxAlert screening is available in the OCHEM documentation (ToxAlerts)

2.5 Set Compare utility

R	Online chemical database with modeling environment				
Home -	Database -	Models -	Moderation -		
Sea mea our Cro Buil prop	perties. The m	Apply a r Create m Open pr View per Calculate SetComp els for predi odels can b	nodel nultiple models edictor nding tasks e descriptors pare utility	9	

The SetCompare utility allows juxtaposing two sets of chemical compounds and finding distinguishing features of each set. For example, you can compare active and inactive compounds for a particular property.

The utility helps to address the following questions:

- What are the distinguishing structural features of active compounds?
- How significant are these results statistically?
- Which are the compounds that possess a particular important structural feature?

SetCompare is accessible from the menu bar.

In the first page of the wizard two sets have to be selected. With the set comparison utility two sets can be examined with respect to common structural alerts and common descriptors.

ne SetCompare utility is experimental. It allows you to compare two sets of molecu ease, provide the two sets available options below.	es based on their structural reatures.
Select the compounds in the first set:	2 Select the compounds in the second set:
Upload compounds from a file (DSDF/MOL2/SMILES/Excel sheet)	Upload compounds from a file (SDF/MOL2/SMILES/Excel sheet)
Provide a Name/CAS-	Provide a Name/CAS- RN/SMILES
Draw Molecule (click on depiction to the right to draw)	Draw Molecule (click on depiction to the right to draw)
⊖ Choose a previously prepared [] set:	⊖ Choose a previously prepared [] set:
Select molecules by a tag: []	○ Select molecules by a tag: []

2.6 Model application

and property name:

Welcome to Apply a model	ions	Check out the properties available on OCHEM
Explore OC Bearch chemical	2 C 3 S 3 S 3 S 3	OCHEM contains 1666815 experimental records for about 43 properties collected from 23262 sources
View pending t View publisher SetCompare u MolQhimiser SetCompare u MolQhimiser Descriptors sto Run predictions Apply one of the available models roou are interested in for your set o Screen your compound libraries ag or such endpoints as mutagenicity aqueous toxicity, etc.	asks cccess by tasks ality riptors rage al to predict property f compounds. h ToxAlerts gainst structural alerts , skin sensitization,	Melting Point logPow logS CYP450 modulation LC50 aquatic log(IGC50-1) Boiling Point AMES DMSO Solubility LogKoc BCF MProperty 9 chembl_IC50 chembl_Ki chembl_pIC50 chembl_KB deembl_Rd deembl_Rdt chembl_EC50 chembl_Active chembl_DEC50 chembl_pEC50 chembl_pKd
DCHEM features. Dur acknowledgement:	3	Dummy pIC50 pKi pEC50 pKd UMLS PROPERTY
Feedback and h	elp	Caco2 permeability
User's manual Check an online user's manual		Caco2 apparent permeability Caco2 permeability rate

or by article id:

apply the model

oredicts log(IGC50-1) using T Pyriformis tutorial dataset (training) (656) ASNN validated by T Pyriformis tutorial dataset (test) (437)

Getting to model application

1. Open the model applier browser from Models > Apply a model.

Models applier browser

- The models applier browser lists all the models (public and private, developed by the user). It shows the model name, the predicted properties, used training set, used machine learning method and the creation date.
- The 🕙 icon links to model export.

2. Please, find the model you want to apply and click "apply the model".

Note:

You can check and apply multiple models simultaneously.

Select a model from the list

(refresh)

🗉 🗐 🖷 🙀 Aqueous toxicity - a demo mode

Model name or model ID:

creation time

1 - 6 of 6

2014-02-13

Models visibility: Public and private Corder by:

ovide the compound(s) to p	redict
ease provide compounds for which everal options are available:	you want to predict the target property
Upload compounds from a file (SDF/MOL2/SMILES/Excel sheet)	Browse MoleculesToPredict.sdf
Provide a Name/CAS-RN/SMILES	
Draw Molecule (click on depiction to the right to dra	w)
Choose a previously prepared set:	[]
Select molecules by a tag:	[]
ditional options	
ediction scenario: Use predictions	only
Disable prediction cache	2

	file (Excel, CSV or SDF) ility domain charts>	Accuracy estimates for the set log(IGC50+1) for 192 compounds RMSE = 0.54 to 0.6 MAE = 0.42 ± 0.05
- 15 of 192	15 tems c	n page 1 of 13 > >>
CIC molecule profile	log(IGC50-1) (Aqueous toxicity - a demo model) = 2.1 -log(mn	nol/L) ± 1.15 (ASNN-STDEV = 0.30, estimated RMSE = 0.59)
	log(IGC50-1) (Aqueous toxicity - a demo model) = 0.83 -log(m	mol/L) ± 1.15 (ASNN-STDEV = 0.46, estimated RMSE = 0.59) OUTOFAD
	log(IGC50-1) (Aqueous toxicity - a demo model) = 1.9 -log(mn	no/L) ± 1.15 (ASNN-STDEV = 0.58, estimated RMSE = 0.59) OUT OF AD
	log(IGC50-1) (Aqueous toxicity - a demo model) = 0.2 -log(mn	nol/L) ± 1.15 (ASNN-STDEV = 0.26, estimated RMSE = 0.59)

Application of regression model:

- **3.** Now it is time to provide the compounds you would like to predict. There are several possibilities:
 - Upload structures (e.g., in SD-format)
 - Provide SMILES for a single molecule
 - Draw a structure in a visual structure editor
 - o Use an earlier created basket
 - Select a certain set of records / molecules by a tag

Selecting the prediction scenario and disabling the prediction cache are additional options for the prediction process.

- 4. Click on next button to start the application Wait until the calculations are completed. Again, if the task is taking long time, it is possible to fetch results anytime later from the pending tasks browser (Models > Pending tasks menu).
- **5.** Now the calculations have finished. The results include prediction values, and accuracy estimates for each predicted compound. Additionally, there is an estimate for the overall prediction accuracy for the set.
- 6. Predictions can be exported to an Excel or CSV sheet, SDF files or R scripts.
- 7. Prediction results of each single structure are listed in the browser. There are the predicted values themselves, distance to model value and an estimation of the accuracy (RMSE), as well as the originally measured value if it is known.

OCHEM predictor - results Here you can browse the predictions for your compounds and export them in a variety of formats	
Export results in a file (Excel, CSV or SDF) Sorting none	Accuracy estimates for the set AMES for 449 compounds Accuracy = 80%
	15
AMES (Ames levenberg) = inactive (95.0% a	ccuracy)
AMES (Ames levenberg) = inactive (95.0% a	ccuracy)
AMES (Ames levenberg) = inactive (95.0% a	CCUFACY) CACHED

Application of classification models

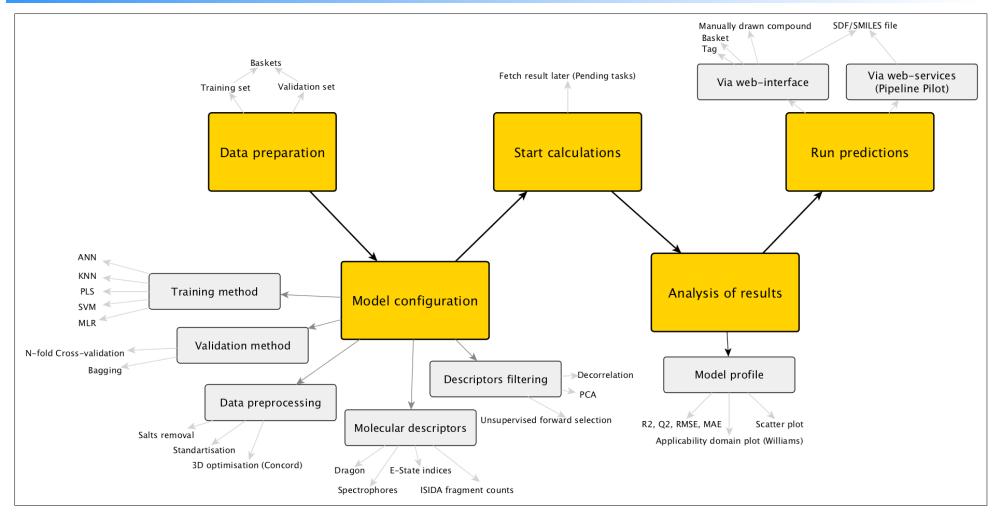
For application of classification models, the same steps have to be done as for a regression model.

8. The result browser shows the predicted class together with an estimation of the accuracy. In this example case, the model was applied to a single drawn structure.

3 Modeling framework

In this chapter, we learn how to upload data and develop QSAR models

Development of multiple models using Comprehensive Modeling (CM) Development of single models



The basic steps of a QSAR modeling lifecycle: prepare data, configure model, train the model, analyse results and use the model to predict new compounds

3.1 Comprehensive modeling

Home *	Database *	Models •			
_	Welcome to		model model	ions	
	plore OC	Create m Open pr	ultiple models edictor	entally	
mea	sured, publisi users. You ca	View per View pub	nding tasks olished tasks	iccess by	
Build		MolOptin	bare utility niser e descriptors brs storage	al	

Comprehensive modeling The comprehensive modeling feature allows you to simultaneously run multiple models with different machine learning methods, molecular descriptors and validation protocols Please note that running multiple models may require significant computational resources and time. Select the training and validation sets: Training set (required): T. pyriformis train [details] Add a validation set The model will predict this property. log(IGC50-1) using unit: -log(mmol/L) \$ Select the methods you want to use for the modeling Model validation Method Descriptors **Descriptor selection** [all] [none] [all] [none] [all] [none] [all] [none] CDK Unsupervised forward selection ASNN (with Library mode) Dragon v.6 (all blocks) OEstate and ALogPS Simple pairwise decorrelation (r < 0.95)</p> 5-fold cross-validation (stratified) KNN Bagging with 64 models LibSVM ISIDA Fragments (Length 2 - 4) +add a custom template GSFrag +add a custom template Mera and Mersy PLS Chemaxon descriptors Inductive Descriptors +add a custom template Adriana Spectrophores QNPR (SMILES - length 1 - 3 threshold 5)
Two simple descriptors (MW+Number of carbons) +add a custom template Show advanced options>> Considering the selection above, 6 models will be created

1. The "comprehensive modelling" feature accessible via the "Models" menu is an advanced feature that allows you to easily create multiple models based on different descriptor sets and training methods.

With this feature, you can create dozens of models simultaneously and directly compare their performance.

- **2.** In the following dialog, first you should select your training set.
- **3.** You can see a set of predefined configuration templates for several training methods, molecular descriptors, descriptor selection methods and model validation.

The checked methods will be applied using the "all against all" principle. On the following screenshot, we selected methods, descriptor sets, descriptor selection method and validation method, which results into six models.

We selected only six models for the reason of speed. Normally, you can run dozens or hundreds of models, depending on available calculation resources.

4. Now we are ready to launch all models.

Please	wait		
) ngth 2 - 4)	Starting model 3 out of Cancel +add a c	s	ek la
Create	mprehensive modeling multiple models simultaneously		
Success! 6 models	have been started, 2 already existen	nt models have been sl	<mark>kipped.</mark>
You can tra	ack the status of your models in the b	pasket summary page.	
Basket browser Browse, Compare or Jo			
Filter by name	: T. pyriformis [Create new 🛅] 🥫	Show public sets	
1 - 2 of 2			
1 - 2 of 2	T. pyriformis validation	449 records	

5. Please, wait until OCHEM starts the necessary calculation tasks.

6. When done, you are forwarded to the success page, from which you can directly go to the **models summary page**.

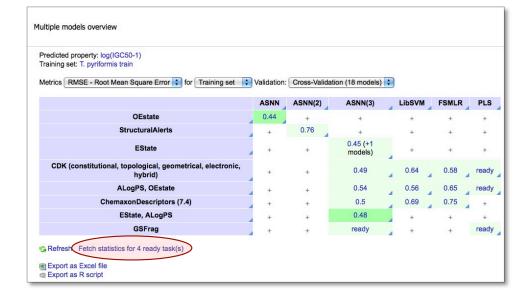
7. The models summary page built for a particular basket is also available via Basket browser (menu Database > Baskets), by clicking icon for your basket.

	property: log(IGC50-1) set: T. pyriformis train						
Metrics	RMSE - Root Mean Square Error 🛟 for Training set 🛟 Vali	dation: Cr	oss-Validation	(18 models)			
		ASNN	ASNN(2)	ASNN(3)	LibSVM	FSMLR	PLS
	OEstate	0.44	+	+	+	+	+
	StructuralAlerts	+	0.76	+	+	+	+
	EState	+	+	0.45 (+1 models)	+	+	+
CDK (c	onstitutional, topological, geometrical, electronic, hybrid)	+	+	0.49	0.64	0.58	in queue
	ALogPS, OEstate	+	+	0.54	0.56	0.65	in queue
	ChemaxonDescriptors (7.4)	+	+	0.5	0.69	0.75	+
	EState, ALogPS	+	+	0.48	+	+	+
	GSFrag	+	+	in queue	+	+	in queue
Refres							

8. The models summary page shows all the models (ready and pending) for the selected basket.

The models are grouped by methods, descriptors and validation protocols. Currently, we see that our four models are still running.

You can return to this page at any time to check the status of your models or click "refresh" to update the dialog. Normally, the creation of multiple models takes a while.



9. To calculate statistics for all the completed models, press the "fetch statistics for ready models".

redicted property: log(IGC50-1) raining set: T. pyriformis train						
Metrics RMSE - Root Mean Square Error 🛟 for Training set 🛟 Valid	lation: Cro	oss-Validation	(18 models)			
	ASNN	ASNN(2)	ASNN(3)	LibSVM	FSMLR	PLS
OEstate	0.44	+	+	+	+	+
StructuralAlerts	+	0.76	+	+	+	+
EState	+	+	0.45 (+1 models)	+	+	+
CDK (constitutional, topological, geometrical, electronic, hybrid)	+	+	0.49	0.64	0.58	0.58
ALogPS, OEstate	+	+	0.54	0.56	0.65	0.56
ChemaxonDescriptors (7.4)	+	+	0.5	0.69	0.75	+
EState, ALogPS	+	+	0.48	+	+	+
GSFrag	+	+	0.62	+	+	0.73
Refresh						

10. We can see all four of our models are ready. The numbers in the cells ("metrics") show the root mean square error.

In this particular case, we can immediately observe that neural network models (ASNN) have lower errors (RMSE) than e.g. the PLS models.

	ed property: log(IGC50-1) set: T. pyriformis train							
Metrics	RMSE - Root Mean Square Error	for Training set 🛟 Valid	ation: Cro	oss-Validation	(18 models)			
	RMSE - Root Mean Square Error MAE - Mean Absolute Error		ASNN	ASNN(2)	ASNN(3)	LibSVM	FSMLR	PLS
	R2 Q2 Model size		0.44	+ 0.76	+	+	+	+
	EState		+ +	+	+ 0.45 (+1 models)	+	+	+
CDK	(constitutional, topological, geomet	rical, electronic, hybrid)	+	+	0.49	0.64	0.58	0.5
	ALogPS, OEstate		+	+	0.54	0.56	0.65	0.5
	ChemaxonDescriptors	; (7.4)	+	+	0.5	0.69	0.75	+
	EState, ALogPS		+	+	0.48	+	+	+
	GSFrag		+	+	0.62	+	+	0.7

11. It is also possible to display other statistical parameters, such as R2 or Q2, using the drop-down box.

Predicted property: log(IGC50-1) raining set: T. pyriformis train						
Aetrics RMSE - Root Mean Square Error 🛟 for Training set 🛟 Valid	dation: Cr	oss-Validation	(18 models) 🛟			
	ASNN	ASNN(2)	ASNN(3) Dele	ete 7 matching	models	
OEstate	0.44	+	+ Exp	ort XML for 7	matching n	nodels
StructuralAlerts	+	0.76	+ Cre	ate 2 missing	models	/
EState	+	+	0.45 (+1 models)	+	+	+
CDK (constitutional, topological, geometrical, electronic, hybrid)	+	+	0.49	0.64	0.58	0.5
ALogPS, OEstate	+	+	0.54	0.56	0.65	0.5
ChemaxonDescriptors (7.4)	+	+	0.5	0.69	0.75	(+
EState, ALogPS	+	+	0.48	+	+	+
GSFrag	+	+	0.62	+	+	0.73
Refresh						

- **12.** You can perform row-wise or column-wise batch operations, e.g., delete the models or create new models.
- **13.** You also can create new models individually by pushing "+" sign in the "missing" cells.

3.2 Development of a single model

Select training set, machine learning method and internal validation options

Select the training and validation sets: Training set (required): T. Pyrformis tutorial dataset (training) (details) Vide a validation set The model will predict this property: (gifCoSCo-1) using unit: [og(mmol/L]]]] (Force the learning methods: (Force the learning methods: (For the learning method: (For t	
Validation set #1: T Pyriformis tutorial dataset (test) [x] [details] Add a validation set The model will predict this property: [ag((3CSG-1) using unit: dog(mmol/L) • Choose the learning method:	Select the training and validation sets:
The model will predict this property: log(IGC50-1) using unit:	Validation set #1: T Pyriformis tutorial dataset (test) [x] [details]
log(IGC50-1) using unit: log(mmu/L) Choose the learning methods: Stagestied modeling methods: O ASIN (Associative Neural Networks) O FSMLR (Fast Stagewise Mittiplic Linear Regression) NUK (Kosarest Nightbors) O LibSVM wrapper with qui-descarch parameter optimisation MLR (Multiple Linear Regression) PLS (Multiple Linear Regression) WEKA-J48 (Weka-based implementation of C4.5 decision tree) Bayesian Regression Consensus model (experimental) Bayesian Regression) KRR (Kernel Ridge Regression) Naive Bayes classifier by Weka Naive Bayes classifier by Weka Maive day additation Validation Validation Validation Validation Veloc> Staffilted cross-validation Validation Validation Staffilted cross-validation Metaco <	Add a validation set
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To start the model development process, open "Model > Create a model" from the menu panel. The first page of the model creation "wizard" asks you to select a training set, external validation sets (optional), a machine learning algorithm and an internal model validation technique.

- **1.** Select the training set and optionally one or more external validation sets that you have prepared before by clicking on the [...] label and the "Add a validation set" link.
- OCHEM supports two dozen state of the art machine-learning methods. For this tutorial, we will use defaults for most of the configurable options. Thus, we will select associative neural networks (ASNN) to train the model.
- You can choose between n-fold cross-validation, bagging and no validation at all. A 5-fold cross-validation is most commonly used.
- **4.** Model configurations can also be imported from earlier model building processes to follow the same protocol.

The model creation process is organized as a "wizard" guiding you through the model configuration process. So click "Next" to navigate forward.

Pre-processing of the molecules includes four options: standardization of some chemical groups for consistency, neutralization of ions, removal of salts and cleaning of given meta-information in the structure file.

5. We will use the default recommended configuration and employ all the available pre-processing options.

Configure molecular descriptors

elect the molecular descriptors	
elect the molecular descriptors	Predictions by OCHEM's featured models Melling Point - 3D (Dragon 6 + Corina) Melling Point - 2D (ALGOPS 22.1 + OESTATE) OCH430 modulation - statle OCH430 modulation - statle OCH430 modulation - statle Additional descriptor types OCSIstle MolPint - 5.4 (F630/3D) Oragon + 5.4
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Forbid NaN and Infinite descriptor values	s correlation coefficient <i>R</i> larger than 0.95

Selection of molecular descriptors is an important step that can significantly contribute to the quality of the model.

6. For this tutorial, we will use the default selection – E-State descriptors and ALogPS.

Several descriptors and descriptor packages are available on the OCHEM platform, ranging from simple 1D to sophisticated 3D descriptors. If a 3D descriptor is selected, a structure optimization method can be selected in the next step (not shown here).

Furthermore, the output of already existing models can be used as input for the new model to train. I.e. a predicted logP value (if not available) can be used as a molecular descriptor. This functionality is referred to as "feature nets".

The next dialog allows filtering out redundant and correlated descriptors.

7. Again, for the purpose of this tutorial, we will use the default values, which include simple filters like pairwise decorrelation.

It is also possible to select the list of desired descriptors manually (advanced).

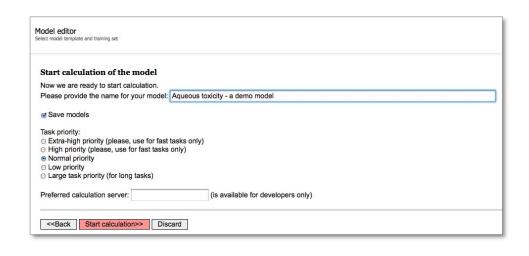
<<Back Next>>

Configure the training method and start calculations

Configure ANN me	ethod 🕕	Configure ANN me	ethod 🕕
Training method:	SuperSAB 🛟	Training method:	SuperSAB
Number of neurons in hidden layer	3	Number of neurons in hidden laver	Momentum SuperSAB
Learning iterations	1000	Learning iterations	RPROP QuickProp
Ensemble	64	Ensemble	Differential equations
Disable ASNN	8	Disable ASNN	QuickProp II
Additional Parameters separated by comma)	PARTITION=3,SELECT	Additional Parameters (separated by comma)	PARTITION=3,SELEC

Each machine learning method (e.g., neural networks in our case, KNN, MLR, PLS, etc.) requires additional configuration options. For neural networks, we can configure the training algorithm, the number of neurons, learning iterations and the number of networks in the ensemble.

8. We will not experiment here now and will use the default options, which are often a good starting point.



- **9.** Finally, we are ready to start calculations. Before starting, please provide the name for your future model.
- **10.** Specifying the priority of the calculations is optional and defaults to "normal".
- **11.** Please click "start calculations" to start the model training process.

Distributed model calculation

Model editor	
Run model builder	Running the teacher - Task started
< <back next="">></back>	

All tasks types	All tasks s	tatuses 🗘	[Refr	esh] [Delete	all matching tacks]	Refresh every minute		
- 1 of 1					-			
Task type / Time started	Model / Task name	Property / Set	Method	Status	Priority	Details		
Model training 2014-02-13 13:30:08	Aqueous toxicity - a demo model	log(IGC50-1) T Pyriformis tutorial dataset (training)	ASNN	assigned	normal 🛟	Processing task CrossValidatio [more>>]	terminate	10

All tasks types	All tasks statuses \$	[Refresh] [Delete	all matchir	ig tasks]	Refresh every	minute	
- 1 of 1							
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Model training 2014-02-13 13:30:08	Aqueous toxicity - a demo model	log(IGC50-1) T Pyriformis tutorial dataset (training)	ASNN	ready	normal 🛟	-	recalculate

A waiting-screen that shows you the status of the calculations. The training process is automatically distributed to several internal calculation units, but still for large datasets it can take a while to complete (from minutes to weeks).

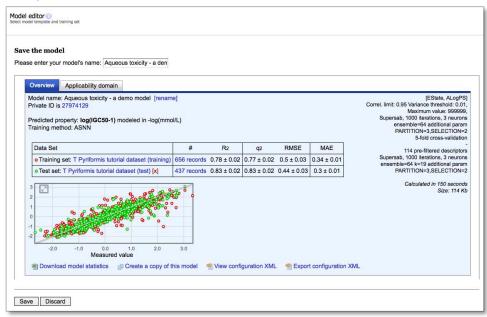
12. Although we could have waited, we will opt to click "fetch result later" to get an overview of currently submitted tasks to the system.

The next screen is the list of currently pending tasks, also accessible from menu "Model > View pending tasks". This list displays all tasks that are currently running on the system or have been finished, but not yet fetched by the user.

13. Here you can observe the status, terminate running tasks or fetch ready tasks. Please, click "refresh" or check the box for Refresh every minute" to actualize the page.

14. When the task has finished, please click the green check button or the model name link to fetch the model and investigate the statistics of the model.

Save your model



If the calculation was successful, you can see the profile of the ready model. Before saving the model the profile can be investigated further:

15. The model profile shows information about the training configuration (used descriptors, machine-learning method and predicted property. It shows the training process statistics like the training set size and correlation coefficients (R2, Q2) and deviation measures of the predicted values to the observed values (RMSE, MAE). On the interactive plot training results of single structures can be inspected. E.g. the calculated descriptor values and the predicted value.

This important dialog is explained in more detail in a chapter on its own. For now please save your model with a meaningful name.

You have successfully built a prediction model on the OCHEM platform

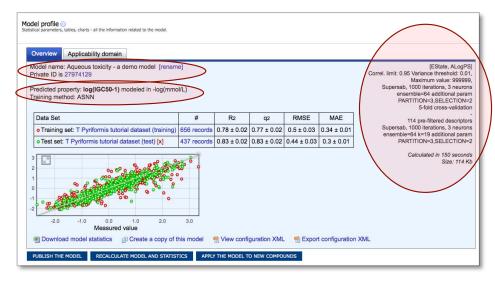


Your model has been saved

Thank you for your cooperation.

Your next possible actions are: Apply your model View your model's properties **16.** After saving the model, you can directly apply it to predict new compounds. Before running predictions, we will investigate the model profile page in more detail.

3.3 Model profile



ecord	Model data	
1 of 1	Model data	
OFT		
-0	log(IGC50-1) = 2.08 (in -log(mmol/L)) Predicted value: 2.16 (in -log(mmol/L))	Dataset = Training
1	ASNN-STDEV: 0.26	
Ĭ.	Zhu, H	
	Combinatorial QSAR modeling of chemical toxicants tested aga N: 140 J. Chem. Inf. Model. 2008; 48 (4) 766-84	
	5 0.000 m m model 200, 10 (1) 10 01	i.tetko 🖂
mole	cule profile open in browser	

The model profile contains all the information related to the performance of the model: **statistical parameters**, **interactive scatter plot** for the single structures, links to the **data sets** used for the model and various operations, like export of the model, application of the model to new compounds, etc.

There is information about the **model name** and **public visibility**, the **predicted property** and the **training method**, used **descriptors** and **validation method**, **training duration** and **model size**.

Typical model statistics are shown in a summary table (5). There are the data sets (training set and given external validation sets) with direct links to their profiles or the records respectively showing the set size. For each set the coefficient of determination $_{R}R^{2''}$, cross-validated R^{2} called $_{R}Q^{2''}$, root mean squared error "RMSE" and mean absolute error "MAE".

Furthermore with the interactive scatterplot, showing predicted versus measured values, single records can be further investigated, e.g. comparison of predicted and measured value or inspection of calculated descriptor values for this structure.

Each point on the scatter plot is clickable and will open the "model point profile" containing the details of the respective compound from the training or the validation set.

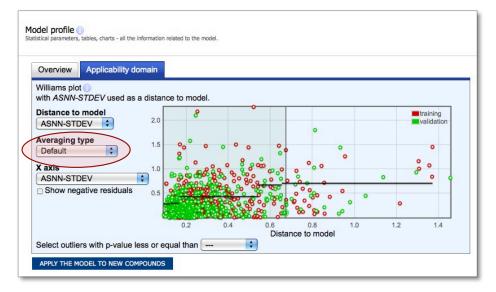
This is a powerful feature that allows you to investigate outliers "under microscope". What are the prediction values, molecular descriptor values, the respective publication, the user who introduced this record? You can track this individually for each compound.

eight 1. tive*0.5] neuron	Correl. limit: 0.95 Variance threshol Maximum value: 99 [AMES with weig	uracy of	ieve accu	dels to ac						
	[OE Correl. limit: 0.95 Variance threshok Maximum value: 99 [AMES with weig (classes weights: [inactive*0.5, active]Levenberg, 1000 iterations, 3 ne ensemble=100 k=0 additional p PARALLE				I>in silico<∕I> m	∕ domain for <	Applicability	blished ir	ents	del name: Ames leve perimental measurem blic ID is 1 edicted property: AME ining method: ASNN
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1019 K		Hit rate	active	inactive	II/Predicted→	Rea	Hit rate	active	inactive	ow ROC curves
		0.78	220	789	inactive		0.75	504	1512	inactive
		0.81	947	225	active		0.801	1876	467	active
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	Calculated in 2614 Size:	0.78	220 947 0.81	789 225 0.78	active Precision		0.75	504 1876 0.788	1512 467 0.764	Real↓/Predicted→ inactive active Precision

For classification models, the model profile shows different statistical parameters. These are:

- accuracy, balanced accuracy, MCC, AUC,
- ROC curves
- Confusion matrices, where you can see the number of false positives, false negatives and so on.

3.4 Applicability domain



Each prediction in OCHEM is complemented with an accuracy estimate.

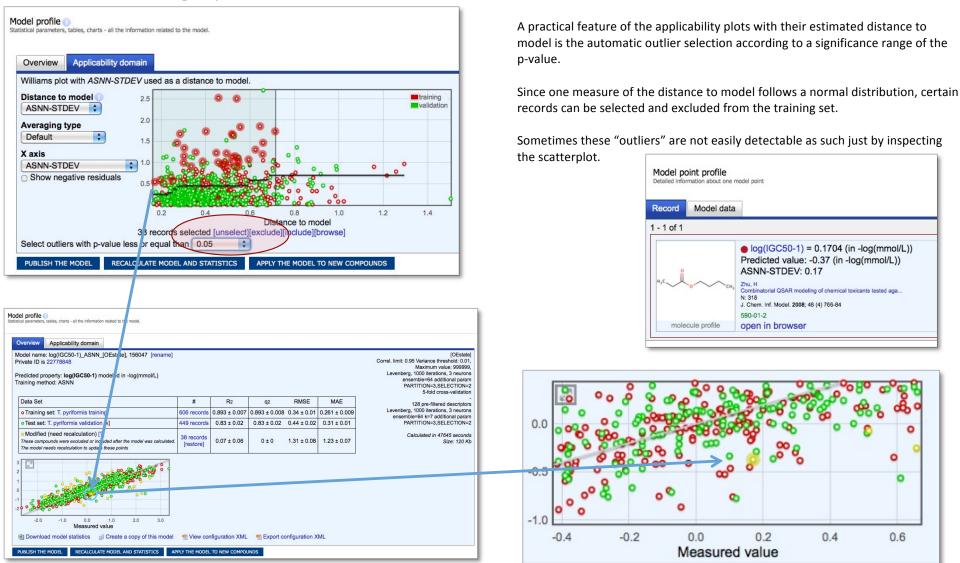
The key concept used for the accuracy estimation is so called **distance to model** (DM). DM is any measure of prediction uncertainty correlated with the prediction accuracy.

Usually, the prediction accuracy falls as DM grows, which is shown on so called **accuracy averaging plots** (shown on the left). The accuracy can be averaged in several manners: via bin-based averaging (used for regression models) and sliding window averaging (used for classification models).



With the Applicability Domain (AD) user interface different display options of the AD can be selected. There is the "Distance to model" type, the "averaging type", sliding "window size" and the label show at the x-axis.

Smart outlier detection using AD plots



Model export

ease, select the iter	ns that you want to e	xport:	
elect all] [select nor	ie]		
Structure (SMILES	or SDF)		
CASRN			
RECORDID			
MOLECULEID			
Identifier in article (NAMES	N)		
Introducers of the n	ecords		
Last modifiers of th			
Publication IDs			
Error messages			
Predicted values			
Experimentaly mea			
DM (distance to mo Conditions of exper			
DESCRIPTORS	inents		
External unique ide	ntifier		
Comments			
Inchi-key			
Get Excel file	Get CSV file	Get SDF file	Get R script

- It is possible to export the data related to your model by clicking "Download model statistics in Excel format". The appearing dialog allows you to select detailed info for the training and validation set – the molecular structures, identifiers, predicted and measured values, prediction accuracies, etc.
- **2.** You can export this data in Excel, CSV, SDF or R formats. For this tutorial, please try to export an Excel file.

4 Working with data

In this chapter, we learn how to upload data using

- Batch data upload
- Advanced basket management

4.1 Batch data upload

6	Online chemical database with modeling environment						
Home •	Database • Models •						
Exy Sea mea our o	Units Articles/Books Journals	Ir possible actions data: experimentally ed to public access by your data.					
Build prop expe	Tags Set area of interest User-related changes	ns of chemical ased on the our database.					
Ru Appl	My data exports	ls to predict property					
you	Batch data upload	of compounds.					



Although you can introduce each record individually, this is usually not practically feasible. Instead, it is convenient to upload hundreds of thousands of records from external files, e.g. Excel or SD files. This can be done using the "Batch data upload" utility.

In this tutorial, we are going to upload about a thousand records for aquatic toxicity (namely, growth inhibition concentration for T. pyriformis).

1. Select the "Batch data upload" item in the "Database" submenu of the main OCHEM menu. You will open the first page of the "Batch upload wizard".

- **2.** Select your provided SD-file in the "Upload file" field. The tool supports SDF and XLS file formats. To remind you, this file contains about 1,000 measured values for the growth inhibition assay.
- **3.** Make sure you select "make the uploaded records hidden" to avoid data conflicts with the other course participants.
- **4.** Hit "Upload" to continue.

	1	-0.16			
CC1-CC-CC-C1 3			-log(mmoi/L)	120055-09-6	
	2	0.87	-log(mmol/L)	2430-16-2	
1=CC=CC=C1 3	3	0.12	-log(mmol/L)		4-Phenyl-1-butanol
0)C1=CC=CC=C1 4	4	0.06	-log(mmol/L)	1565-75-9	
C1=CC=C(N)C=C1 5	5	0.97	-log(mmol/L)	39905-50-5	
OC1=CC=C(N)C=C1 6	6	1.38	-log(mmol/L)	39905-57-2	
=CC=C(N)C=C1	7	0.22	-log(mmol/L)	99-88-7	
=CC=C(N)C=C1 8	8	1.07	-log(mmol/L)	104-13-2	
=CC=CC=C1	9	0.42	-log(mmol/L)		(2-BROMOETHYL)BENZEN
=CC=C1N	10	-0.16	-log(mmol/L)		2-methylaniline
	D)C1=CC=CC=C1 C1=CC=C(N)C=C1 OC1=CC=C(N)C=C1 =CC=C(N)C=C1 =CC=C(N)C=C1 =CC=C(N)C=C1 =CC=CC=C1	D)C1=CC=CC=C1 4 C1=CC=C(N)C=C1 5 OC1=CC=C(N)C=C1 6 =CC=C(N)C=C1 7 =CC=C(N)C=C1 8 =CC=CC(N)C=C1 9	D)Cl=CC=CC=C1 4 0.06 Cl=CC=C(N)C=C1 5 0.97 DCl=CC=C(N)C=C1 6 1.38 =CC=C(N)C=C1 7 0.22 =CC=C(N)C=C1 8 1.07 =CC=CC(C1) 9 0.42	O)C1=CC=CCC=C1 4 0.06 -log(mmol/L) C1=CC=C(N)C=C1 5 0.97 -log(mmol/L) OC1=CC=C(N)C=C1 6 1.38 -log(mmol/L) =CC=C(N)C=C1 7 0.22 -log(mmol/L) =CC=C(N)C=C1 8 1.07 -log(mmol/L) =CC=C(N)C=C1 9 0.42 -log(mmol/L)	O)C1=CC=CC=C1 4 0.06 -log(mmol/L) 1565-75-9 C1=CC=C(N)C=C1 5 0.97 -log(mmol/L) 39905-50-5 OC1=CC=C(N)C=C1 6 1.38 -log(mmol/L) 39905-57-2 =CC=C(N)C=C1 7 0.22 -log(mmol/L) 99-88-7 =CC=C(N)C=C1 8 1.07 -log(mmol/L) 10-13-2 =CC=CC(C)C=C1 9 0.42 -log(mmol/L) 10-13-2

5. The second page of the wizard is the file review page with "column remapping" tool. Here you can preview the first few lines of your uploaded file and see which columns were recognized by the system. On this page, you also have the possibility to reassign column names and select/deselect columns for upload.

Note:

Column headers are colour coded. Green means recognized by the system, red means not recognized. The property is dark green if it is already in the system.

Columns can be remapped by clicking on the column header

) ☑ UNIT (log(IGC50-1))	Known colun	nn
-log(mmol/L)	Property	
-log(mmol/L)	Condition	
-log(mmol/L)		4-Phenyl-
-log(mmol/L)	1565-75-9	

Select	column	name	
unit	\$		
	ОК	Cancel	

6. For example, the column holding the data values is named "UNIT ..." in the uploaded file. We need to specify that these values represent the unit for the "Aqueous toxicity" property. Click on the red unrecognized "Unit ..." column header and select "Known column" from the popup menu. Then select unit from list of known columns.

_pyriformis_all.sdf							
MOLECULE	SMILES	✓ N	✓ log(IGC50-1)	∎unit	CASRN	☑ NAME	
1 2 3 11 11 0 0 0 0 9	CCC(0)CC1=CC=CC=C1	1	-0.16	-log(mmol/L)	120055-09-6		
	OCCCCCCC1=CC=CC=C1	2	0.87	-log(mmol/L)			
	OCCCCC1=CC=CC=C1	3	0.12	-log(mmol/L)		4-Phenyl-1-butanol	
1 2 3 11 11 0 0 0 0 9	CCC(C)(0)C1=CC=CC=C1	4	0.06	-log(mmol/L)			
		5	0.97	-log(mmol/L)			
	CCCCCCOC1=CC=C(N)C=C1	6	1.38	-log(mmol/L)			
	CC(C)C1=CC=C(N)C=C1	7	0.22	-log(mmol/L)			
	CCCCC1=CC=C(N)C=C1	8	1.07	-log(mmol/L)			
1239900009	. ,	9	0.42	-log(mmol/L)		(2-BROMOETHYL)BENZENE	
1238800009	CC1=CC=CC=C1N	10	-0.16	-log(mmol/L)		2-methylaniline	
reen titles indicate reco property, condition or icking on the green bu you have irrelevant co	sing, the stub unpublished article ognized columns, red titles indi another column type like name	icate e, val	errors. Please cli ue or molecule, t	ck on the red o hen select the	matching enti	ty and confirm your selection	by

Batch Upload 3.0 - Entity remapping Review and remap the properties, conditions, units, articles and baskets involved in the data upload
Database entities remapping
Property: log(IGC50-1)
Values
Unit: -log(mmol/L), min value: -2.6656, max value:
3.34
Article: unpublished
Molecule set: default
submit

7. Note that the column header has changed from red to green (recognized unit), the header name is now just unit, and the checkbox in the column header is checked, indicating that the column will be processed by the tool.

8. Click the "Upload this sheet" button to proceed to page three of the wizard.

- **9.** The third page of the wizard is the "entity remapping" page. You can review and change some aspects of the uploaded data (property, unit used for data upload, article, etc.)
- **10.** Since no article has been specified in the data sheet, a stub "unpublished" was put instead of the article. Final corrections can be done here, e.g. correcting the unit, selection of a certain article or renaming of the basket.
- **11.** With "submit" the data can be uploaded to the database. In this case the data will be introduced by default as hidden data and is only visible to the current user (recommended option for the tutorial exercise).

Note:

To upload data originally published in an article click on the "Unpublished" link in the "Article" section of the page.

Batch upload preview bro	wser	
Summary: All rows in the sheet	Count: 1093	
Status: valid,	Count: 1093	
Filter by row number:	and row type: all	Batch operation
1 - 10 of 1093	10 items on page 1 of 1	10 > >>
Row 1 •Save •Skip	log(IGC50-1) = -0.16 (in -log(mmol/L)) eADMET T_pyriformis_all.sdf N: 1	
ĺ	120055-09-6 MoleculeID: M6569	RecordID: R- eadmet st Only visible to eadm
Row 2 •Save •Skip		

Row 9 ●Save ●Skip		 log(IGC50-1) = 0.42 (in -log(mmol/L)) eADMET T_pyriformis_all.sdf N: 9 	
		(2-BROMOETHYL)BENZENE MoleculeID: M12429	RecordID:
		Proclucity, P12423	eadmet 🖾 Only visible to ea
Row 10 •Save •Skip	H ₂ N	 log(IGC50-1) = -0.16 (in -log(mmol/L)) aADMET T_pyr(formis_all.sdf N: 10 	
		2-methylaniline MoleculeID: M9899	RecordID: F
		Proteculery, 19935	eadmet 🖾 Only visible to ea
- 10 of 1093		10 ; items on page 1 of 110 > >	>
Proceed with	upload)		

Batch upload 3.0 - finished Your upload has been finished					
Batch upload results Batch upload is finished. You ca	n download the detailed	pload report.			
Summary:					
All rows in the sheet	Count: 1093				
Status: valid, saved_valid	Count: 1093				
				 New Batch Upload	Download Excel file

Depending on the size of the uploaded set, the process may take from seconds to hours for completion (e.g. more than 50000 data points).

12. The fourth page of the wizard is the data preview browser. Here you can review your records and determine any errors in the data upload process.

The page holds information on the total number of records to be uploaded, the number of valid, and erroneous or duplicated records among them. You can select or deselect individual records from the upload.

13. Since all records being uploaded are valid, continue the upload by clicking the big "Proceed with upload" button.

The upload itself is the slowest part in the process. It may take from seconds (for a hundred records) to several hours (for a large dataset of tens of thousands of records).

14. The final page of the batch data upload wizard gives some statistics about the uploaded data. You can review the uploaded data in the "Experimental property browser" or download a detailed report.

Note:

For your convenience, the uploaded data are automatically put into a newly created basket.

4.2 Advanced data management

R	Onlin		nical database odeling environment
Home -	Database -	Models -	
	Baskets		

Basket browser	sule set	
Filter by name:	[Create new **] _Show public sets	
1 - 11 of 11		
- 🗟 🖬 🕵	Selected records	0 records
0 1 2 6	T pyriformis all.sdf	1093 records

T. D. offerencie to testanist states		
T Pyriformis tutorial datase	34	
basket name is good and length is 29	-	
Description (optional):		
Actions		
Create a copy of this	basket	Split the basket into two sets
Create a primary recon		Transform the basket using OScript
Add or delete particular		Export this basket into Excel, CSV or SDF
Discretize the nume	ical values	
Statistics of the basket		
Statistics of the basket		
Properties		Records Unique compounds
		1093 records 1087 compounds
log(IGC50-1)		nemistry) 1087 (1087) compounds
log(IGC50-1) Total Compounds (ignorin	ng stereo-che	
Total Compounds (ignorin	-	
. ,		
Total Compounds (ignorin		
Total Compounds (ignorin Articles Count T_pyriformis_all.sdf 10 fags	93 Count	
Total Compounds (ignorin	2 Count	

OCHEM allows combining experimental records into reusable sets. Such sets of records are referred to as **baskets**. Baskets have names and each user can virtually have as many baskets as required. Baskets are typically used as training or validation sets for the development of QSAR models.

Basket browser and basket profile

 Review a list of all your baskets in the basket browser accessible from Database > Baskets menu

The list should contain the basket created during the batch upload process from the previous tutorial.

- **2.** To open the profile of a particular basket, click on its name.
- **3.** The profile shows you brief information on the basket size, its content, articles, properties and tags. Here, you can also rename your basket or perform a number of advanced operations on it.

Please give your basket a name of your choice (e.g., "T. Pyriformis tutorial dataset") and click save.

Splitting the basket

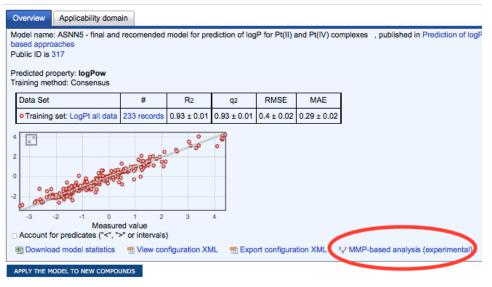
4. Click the link to randomly split your basket into two subsets.

Basket splitter	5. Enter names for the two new baskets to be created
You are going to split the basket T Pyriformis tutorial dataset into two new baskets. Provide the basket names Basket 1: T Pyriformis tutorial dataset (training)	 Select the percentage by which the molecules are divided randomly in the training and validation set.
Basket 2: T Pyriformis tutorial dataset (test) Select the splitting method © Random splitting Size of the validation set, in percentages: 40 % O Y-based splitting (not implemented yet) Your original basket will be preserved. Split the basket	7. Click "split the basket" and wait for a couple of seconds.
The basket has been splitted successfully!	 After notification that the new baskets have been created, you are forwarded to the basket browser. The process is complete and you can see two new baskets in the basket browser. We will use these baskets further in the model development tutorial.
Basket browser () Browse, Compare or Join molecule set	
Filter by name: [Create new 1] Show public sets 1 - 15 of 20	15 items on page 1 of 2 > >>
😑 🗟 😰 Records in trash	0 records
Elected records	644 records 5 pending models
T Pyriformis tutorial dataset (test)	437 records
T Pyriformis tutorial dataset (training)	656 records
E 🗟 🖉 🙀 T Pyriformis tutorial dataset	1093 records

5. MMP Analysis

In the previous chapters, you have learned to perform the most basic steps of a QSAR modelling. In this chapter, we will review advanced features:

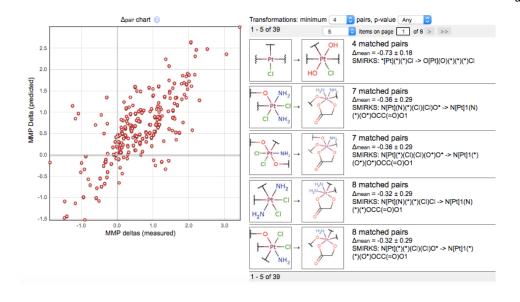
Interpretation of models using Molecular Matched Pairs

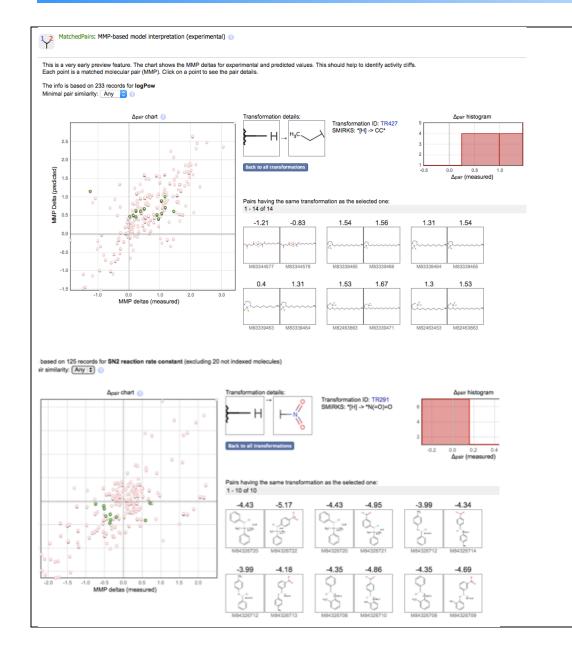


Matched Molecular Pair analysis to interpret models.

1. Click on the highlighted region to open the MMP plot

2. Points near to the diagonal corresponds to the MMPs which were learnt. Points in 2nd and 4th quadrants correspond MMPs for which even the sign was incorrectly predicted.





3. Clicking on each point will show the transformation responsible for each MMP. In the shown case the point in the 4th quadrant does not agree with qualitative changed of property (logP) for other transformations. It is likely to be an error, indeed addition of two carbon atoms CC increases logP – this is not an expected behavior.

4. The same approach can be used to interpret chemical reactions. Here the two groups of pairs of reactions correspond to two types of reactios.

Acknowledgement

I thank Dr. Sushko, Dr. Novotarskyi, Mr. Körner and other members of my team for their work on the

development of the OCHEM platform and preparation of the earlier version of this tutorial.

