

Hands-on Workshop on eNanoMapper tools + services

Workshop:	Hands-on Workshop on eNanoMapper tools + services		
DATE / PLACE:	29-30 September 2016 / National Technical University of Athens		
TIME:	10:00-11:30		

TITLE:	Extracting knowledge from data using the JaqPot Modelling Tool
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ABSTRACT

The workshop will offer hands-on work on the development of nanoQSAR models based on data available from the data.enanomapper.net server, making use of the eNanoMapper computational infrastructure from NTUA¹ that extends the OpenTox API². The focus is on the use case of predicting cell association of metal oxide Nanoparticles, based on experimental data, publically available in the publication of Gajewicz et. al.³.

Participants will go through the workflow of constructing a model from a dataset drawn from the eNanoMapper database into the Jaqpot platform, getting predictions based on the model and evaluating its efficiency through model validation. Finally, users will work on the creation of predictive models using statistical and machine learning algorithms.







1. Login & Authorization to the Jaqpot and eNanoMapper services

Users access the Jaqpot homepage by NTUA <u>http://jaqpot.org</u> (Figure 2), currently as a test instance as is under development. To gain access to the services, users should click the Sign in button, to be transferred to the screen in Figure 3.



Figure 2 Homepage of eNanoMapper NTUA Modelling interface



Figure 3 Login page





Log in to Jaqpot Quattro using OpenTox credentials. You can get an OpenTox user account by clicking on "Create Account" on the start page (Figure 2) at: <u>http://old.opentox.org/join_form</u> (Figure 4), login with your own credentials or just type "**guest**" in both fields. Successful entry of credentials leads to the homepage for registered users (Figure 5).

Open To	X	Site Map Accessib	ility Contact	Log in Join		Search Site
Home Applications [Downloads Tutorials	OpenTox Blog	Development	OpenToxipedia	Meet	
About Reading Room	FP7 REACH Mailing	glist FAQ				
Note: Please do not use a Personal Details Full Name = Enter full name, eg. John Smith.	ny accents, umlauts or	special characters	(e.G.: ê,á,ò,ü,	3) in this form		
User Name =						
Enter a user name, usually some	hing like 'jsmith'. No spaces or :	special characters. Usernam	nes and passwords a	re case sensitive, make	sure the caps lock key is not enabled. This is	the name used to log in.
E-mail = Enter an email address. This is ne	ecessary in case the password is	lost. We respect your priva	cy, and will not give	the address away to ar	ny third parties or expose it anywhere.	
	Figure 4 Open	ox Registratio	n page (htt	p://www.op	pentox.org/join form)	



Figure 5 Starting page for registered users





2. eNanoMapper Jaqpot modelling services – A walkthrough

In this section we will have a brief walkthrough of the eNanoMapper Jaqpot modelling interface. Please note that this a test User interface for the Modelling API (Application programming interface), which can be examined by more proficient users who would like to use the Web Services exposed through it achieve more automated modelling workflow. Users in order to can visit http://jaqpot.org:8080/jaqpot/swagger/ to view the documentation of the API and experiment with its possibilities using the Swagger interface hosted there. As the eNanoMapper Jaqpot modelling interface leaves the testing phase and reaches finalization, a manual for its functionality will be released.



Figure 6 Jaqpot menu tree

The menu tree of the Jaqpot modelling interface is shown in Figure 6. Only the items of interest to this workshop will be described; please also note that this is test release and tools are under development.

The modelling Actions are (see Figure 6):

- Dataset Creation from • an existing dataset in the data.enanomapper.net database, allowing selection a subset of substances or properties and execution of descriptor calculations when image files or Crystallographic data are included in the dataset, resulting to the creation of a local dataset.
- Modelling using a dataset in the data.enanomapper.net database or the local dataset and applying



Select substances, properties and Figure 7 "My resources" menu

"My resources" menu





one of the available algorithms, currently:

- o Regression
 - ocpu-Im: Linear Regression by R offered through OpenCPU
 - weka-mlr: Linear Regression by Weka
 - weka-pls: Partial Least Squares by Weka
 - weka-svm: Support Vector Machines by Weka
 - *python-pls-vip*: Partial Least Squares with "Variable Importance in Projection" scores by Python
 - python-lasso: Lasso Regression by Python
 - *python-lm*: Linear regression by Python
- o Classification
 - weka-pls: Partial Least Squares Classification by Weka
 - python-id3-mci: ID3 classification algorithm
 - Bernoulli Naive Bayes
 - Generalised Naive Bayes
 - Multinomial Naive Bayes
 - CMI Decision Tree
 - Id3 with MCI

Users can export their model in PMML format (explained in following chapter) and distribute it online using the unique URI assigned to it for easy deployment to other systems/web services. More detailed descriptions on the algorithms, their implementation and their parameters can be found in the openly available Deliverable 4.3 of eNanoMapper.

- **Predict** by applying a model to a dataset with matching properties and receive a dataset with predictions and Domain of Applicability calculations, if that choice has been made.
- Validate model to review its efficiency by applying it to a dataset.

Available choices are:

- External validation: apply the model to another dataset
- Cross validation: a model validation technique for assessing how the results of a statistical analysis will generalize to an independent data set, more info here: <u>https://en.wikipedia.org/wiki/Cross-validation_(statistics)</u>
- Split validation: split the data into training and test, train a model and apply it to the test dataset.
- Experimental design
 - o With data available
 - O No data available
- **Other:** Interlaboratory testing and Read Across will soon exit the test phase and be released.

The resources that are available to the user are:

- QSAR
 - o **Model**: a repository of models created by the user and example models
 - o Algorithm: a list of the available algorithms offered by Jaqpot
- Dataset
 - o A repository of datasets created by the user and example datasets
- **Reports, Bibtex, Features**: future capabilities





3. Model training and use for predictions

The workflow that will be followed in the three modelling cases is shown in Figure 8.



Figure 8 Workflow of the modelling cases

We will work on a dataset investigating the Cytotoxicity of Metal Oxides Nanoparticles by Gajewicz et.al.³.

Please note that the dataset we will work on has undergone processing prior to modelling. The following steps guide the user to performing the modelling operations outlined in Figure 8.



Figure 9 Train a model Action





Jaqpot Actions -	My resources -	Welcome back, filipposd O
	a second contractor	
Select da	taset:	
(a		
Example datase	ets:	
Name	Title	Teachater (
E20vuMNNTHsp	Gajewicz et al - 10 Metal Oxide NPs	10 MeOx NPs with 29 descriptors, used for predicting HaCaT toxicity.
KIL42dC8JSmp	Walkey et al - 56 Gold NPs	56 Gold NPs with 25 PhysChern descriptors, used for predicting cellular interaction.
PVrY25q3vd5O	Walkey et al - 56 Gold NPs	56 Gold NPs with 76 protein corona descriptors, used for predicting cellular interaction.
UhysSLaF345pKl	Walkey et al - 84 Gold NPs	84 Gold NPs with 76 protein corona descriptors, used for predicting cellular interaction.
class-dummy	Dummy Classification Dataset	This dataset contains classification data
interlab-dummy	Interlab testing dataset	This dataset contains data for interlab testing (3 measurements + uncertainty)
All Datasets:		
Name	Title	Ferentiation
2Mb0LIPm0PsvW3OliavkC	Walkey et al - 28 Gold NPs	28 Gold NPs with 25 PhysChem descriptors, used for predicting cellular interaction
AmMoEDDVEH	ovard	
WITHWISPPPTEIU	ungou	igus
Previous 1 Next		



0	
Train model	
Choose Algorithm	
Regression	Classification
CLasso Regression	OBernoulli Naive Bayes
Linear Regression	Generalised Naive Bayes
PLS - with VIP scores	CMI Decision Tree
MLR - Weka Implementation	OID3 Decision Tree
PLS - Weka Implementation	⊡ld3 - with MCI
SVM - Weka (LibSVM) Implementation	Multinomial Naive Bayes
CR LM Algorithm	OPLS - Weka Implementation
Previous 1 Next	Previous 1 Next

Users should access the start page for registered users and select the "Actions \rightarrow Train" option, as shown in Figure 9, where the Gajewicz et al - 10 Metal Oxide NPs dataset should be selected (Figure 10). The screen of Algorithm selection follows (Figure 11). Here Lasso Regression should be selected.





For details on the algorithms and their parameters please consult this document: http://www.enanomapper.net/deliverables/d4/150801_eNanoMapper-D4_3final.pdf .

In Figure 12 the appropriate parameters are given, as shown in the screen (all settings are kept to default). Here *Scaling* was not selected and no *Domain of Applicability* (DoA) calculations were requested.

•			
Algorithm Title: python-lasso		Select variables : Select Input variable(s) and endpoint Select PMML Upload PMML file Select endpoint only (all other variables will be used as input variables)	
Title:	Lasso Regression	Select Input variable(s) and endpoint:	
Parameters: Alpha:	1.0	Input variable(s) Select All	Endpoint C ZETA POTENTIAL
Fill in the model	e title and description of the produced		C Aspect ratio Y Chemical hardness O logLC50 Core core repulsion energy Core
Model description: Lasso Regression - <u>Galewicz</u>		Standard enthalpy of formation of metal oxide nanocluster DHcf	C Schuurmann MO shift alpha Shift
		Select scaling method: None Select domain of applie None Train	: cability method:

Figure 12 Train: Algorithm parameters – selection of input-output variables

After an intermediate Task page (Figure 13) the result of the Train action is given, which is the page of the model (Figure 14), which contains full information on the model.

Task:	Training on algorithm: py	hon-lasso #T9zu8G8EgZnA		
Status:	COMPLETED			
Туре:	TRAINING			
Date:	09/27/16			
Result: Description:	See result			
Training	task using algorithm python-lasso			
0				

Figure 13 Train: Training on algorithm task page





	е: +QCxjR7KQBAkP6pII/Bik
:	Lasso Regression - Gajewicz
scription:	
	Lasso Regression - Gajewicz
Fransformations:	http://app.jaqpot.org:8080/jaqpot/services/model/owTS847vOtDz7v8P23vZ
Algorithm: python-lasso Features:	
Required Featur	res
Dependent Fea	tures
Independent Fe	atures
Predicted Featu	ires
Representati	ion:
PMML	

Figure 14 Train: Model page with details

A very important aspect here is the generation of the PMML representation of the model, which allows the seamless integration of models into various platforms. This model's PMML file is accessible by users who have logged in (at least as guests) at http://jaqpot.org/m_detail?name=QCxjR7KGBAkP6pllfBtk.

Predictions will be made by clicking the Predict button and selecting the **Gajewicz et al - 8 Metal Oxide NPs** Dataset (Figure 15). The dataset with predictions is given next (<u>http://jaqpot.org/predicted_dataset?name=5RCO6mRmFU5hiz4zMIAv&model=QCxjR7KGBA</u> <u>kP6pIIfBtk</u>, Figure 16).





Choose method: Select dataset. Insert values.				
Select da	ataset for predic	ction:		
Example Data	sets:			
interlab-dummy	Interlab testing dataset	This dataset contains data for interlab testing (3 measurements + uncertainty)		
interlab-dummy2	Interlab testing dataset	This dataset contains data for interlab testing (2 measurements + uncertainty)		
kE0RiswkaCrg	Walkey et al - 28 Gold NPs	28 Gold NPs with 76 protein corona descriptors, used for predicting cellular interaction.		
ICUNPD99xGjgkU	Walkey et al - 84 Gold NPs	84 Gold NPs with 25 PhysChem descriptors, used for predicting cellular interaction.		
vNxjUZMyv33fTM	Gajewicz et al - 18 Metal Oxide NPs	18 MeOx NPs with 29 descriptors, used for predicting HaCaT toxicity.		
xbR5AMG1rOBc	Gajewicz et al - 8 Metal Oxide NPs	8 MeOx NPs with 29 descriptors, used for predicting HaCaT toxicity.		
yzsAXE5rLPzz	Walkey et al - 28 Gold NPs	28 Gold NPs with 25 PhysChem descriptors, used for predicting cellular interaction.		
All Datasets:				
Name	The	These distant		
5p3qHIvdBAvU	another corona dataset	This dataset contains corona data		
RfbLMSC3fdXO	fdsfsd	fsdsd		

	Search:			
Compounds	 International Advancements and an exception of ColorAddition (ColorAddition), and ColorAddition (ColorAddition), and ColorAddition (ColorAddition), and ColorAddition) 			
Al2O3	1.9911716068			
Cr2O3	2.37343366203			
Fe2O3	2.2659755716			
La2O3	2.8371148572			
NiO	2.5591079205			
SnO2	2.39895124314			
WO3	2.59304489739			
Y2O3	2.2234674521			
Previous 1 Next				

Figure 16 Predict: Predictions by model





4. Model validation

A very important step before accepting mathematical models for nanomaterials, and models in general, is model validation, defined as *"the process of deciding whether the numerical results quantifying hypothesized relationships between variables, obtained from regression analysis, are acceptable as descriptions of the data"* (https://en.wikipedia.org/wiki/Regression_validation).

The validation process can be performed on the Jaqpot platform by clicking the Validate button on the model's page (<u>http://jaqpot.org/m_detail?name=QCxjR7KGBAkP6pIIfBtk</u>, Figure 14), selecting an appropriate dataset, not the one used for training, but one that still contains the same properties and substances. Here the **Gajewicz et al - 8 Metal Oxide NPs** Dataset should be selected (Figure 15).

This leads to a report page http://jaqpot.org/report?name=6TgaNqGYrETZw9X (Figure 17).

-				
Title:	External validation report			
Description:				
External validation with model:http://app.jaqpot.org:8080/jaqpot/servic es/mode/QCx/R7/KGBAkP6pII/Btk and				
Algorithm Type:	REGRESSION			
F-Value:	13.12			
Number of				
variables:	2			
RMSD:	0.14			
R^2 (OECD):	0.81			
R^2 Adjusted (if applicable):	0.73			
StdError:	0.17			

Report: #6TgaNqGYrETZw9X

Figure 17 Model validation report

Please not the following on the definition of R^2

(https://en.wikipedia.org/wiki/Coefficient_of_determination):

In statistics, the **coefficient of determination**, denoted R^2 or r^2 and pronounced "R squared", is a number that indicates the proportion of the variance in the dependent variable that is predictable from

the independent variable. If $oldsymbol{y}$ is the mean of the observed data:

$$ar{y} = rac{1}{n}\sum_{i=1}^n y_i$$

then the variability of the data set can be measured using three sums of squares formulas:

• The total sum of squares (proportional to the variance of the data):

$$SS_{
m tot} = \sum_i (y_i - ar y)^2,$$





• The sum of squares of residuals, also called the residual sum of squares:

$$SS_{
m res} = \sum_i (y_i - f_i)^2 = \sum_i e_i^2$$

The most general definition of the coefficient of determination is

$$R^2 \equiv 1 - rac{SS_{
m res}}{SS_{
m tot}}$$

5. Exercise: Explore algorithms

It is now your exercise to:

- navigate the Jaqpot platform
- try out different models
- validate your models
- evaluate model performance

What have you found?





REFERENCES

- 1. NTUA eNanoMapper Web services, available at http://jaqpot.org, described at http://jaqpot.org"/http://jaqpot.org, describe
- 2. Hardy, B. et al, Collaborative Development of Predictive Toxicology Applications, Journal of Cheminformatics 2010, 2:7.
- Gajewicz, A.; Schaeublin, N.; Rasulev, B.; Hussain, S.; Leszczynska, D.; Puzyn, T.; Leszczynski, J. Towards understanding mechanisms governing cytotoxicity of metal oxides nanoparticles: Hints from nano-QSAR studies. Nanotoxicology 2014, 5390 (April 2016), 1–13 DOI: 10.3109/17435390.2014.930195.

