

Istituto di Ricerche Farmacologiche Mario Negri IRFMN

Alessandra Roncaglioni

March 4th, 2024

AiChemist PI intro



The institute

Private non-Profit biomedical research organization (est. 1963)

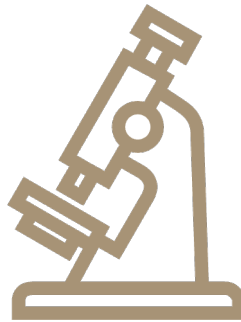
- Carrying out scientific research
- Training young people
- Disseminating the results of our research



3 locations
(MI-BG)



600 people

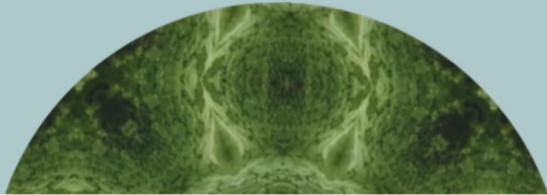


10 departments

Department of Environment Health Sciences Lab. of Environmental Toxicology and Chemistry

Head of department:
Emilio Benfenati

Environmental Health Sciences



4 labs + 1 unit

Laboratory of Chemistry and Environmental Toxicology



Head of laboratory
Alessandra
Roncaglioni

About me



- MSc Environ. Sciences (UniMI Bicocca)
 - Internship@IRFMN
- PhD Open University
 - QSARs for endocrine disruptors
- Visiting scientist (Slovenia & France)
- Head Comp Tox unit @IRFMN
- Head of Lab @IRFMN

Laboratory of Chemistry and Environmental Toxicology

What we do

Identifying and preventing the negative effects of environmental factors and contaminants on health



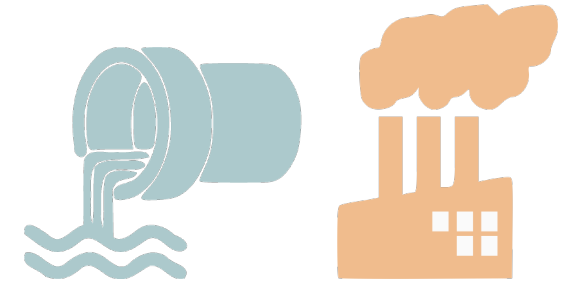
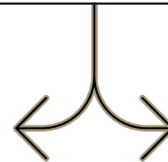
COMPUTATIONAL TOXICOLOGY

Hazard assessment:

In silico modeling (QSAR, read-across)
of (eco)tox data
(Eco)tox bioassays



RISK ASSESSMENT



ANALYTICAL MONITORING ACTIVITIES

Exposure assessment:

Environmental monitoring
Internal exposure estimation

Laboratory of Chemistry and Environmental Toxicology

What we do



RISK ASSESSMENT

Research activities in regulatory toxicology

NAMs

New approach methodologies: any non-animal technology, methodology, approach, or combination thereof that can be used to provide information on chemical hazard and risk assessment

AOPs

Develop efficient and innovative safety testing strategies through a comprehensive mechanistic understanding of cause-consequence relationships of adverse chemical effects



COMPUTATIONAL TOXICOLOGY

A variety of tools providing computer simulations of complex biological phenomena:

Bio/Chem
Database

(Q)SAR

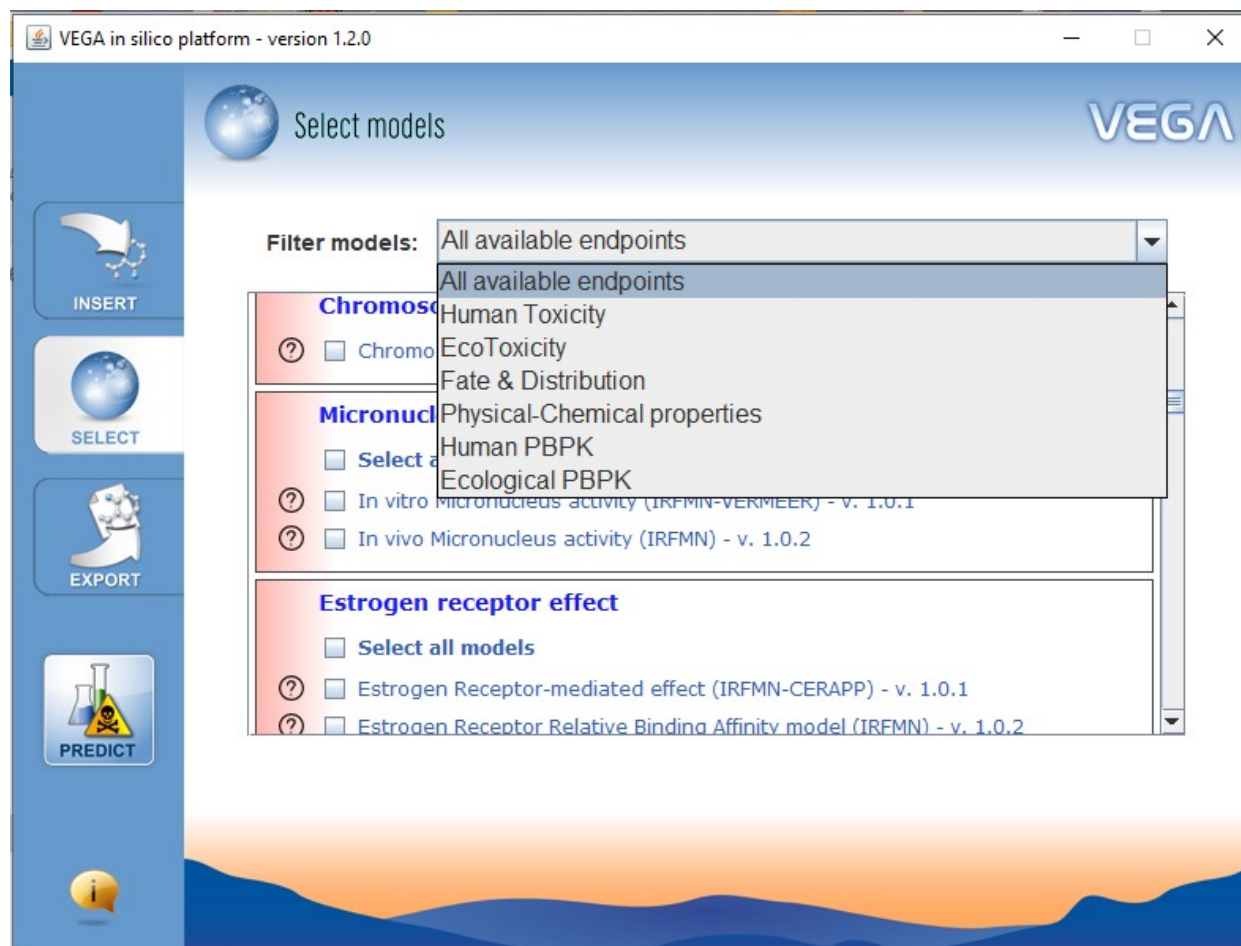
Molecular
modelling

Machine
Learning

Artificial
Intelligence

Read
across

ADME
profiling



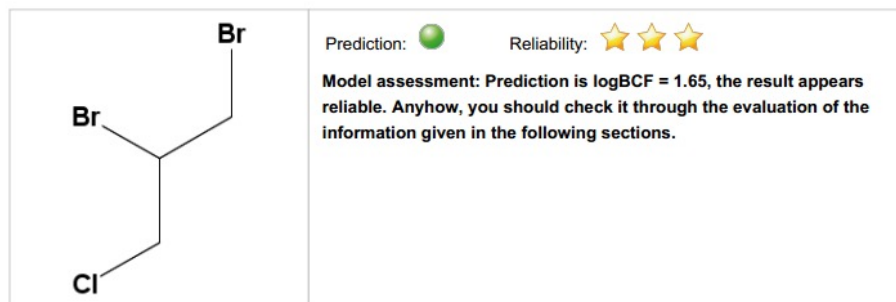
- Unified environment encoding tens of models for different effects: (eco)toxicity, environmental properties, phys-chem, etc....
 - Multiparametric applicability domain definition

Applicability Domain Index (ADI)

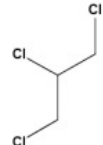
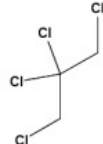
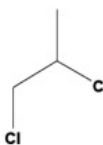
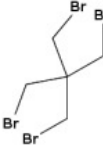
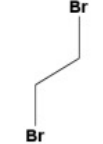
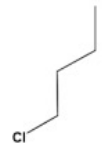
1. Prediction Summary



Prediction for compound 1 (Molecule 1)



Compound: 1
Compound SMILES: C(C(CBr)Br)Cl
Experimental value: -
Prediction: 1.65 [log(L/kg)]
Prediction: 45 [L/kg]
Prediction of model 1 (HM): 1.75 [log(L/kg)]
Prediction of model 2 (GA): 1.61 [log(L/kg)]
Structural Alerts: -
Calculated LogP: 2.96 [log units]
Reliability: Compound is in model Applicability Domain
Remarks for the prediction:
none

	CAS: 96-18-4 Dataset id: 255 (training set) SMILES: <chem>C(C(CCl)Cl)Cl</chem> Similarity: 0.997 Experimental value: 0.96 [log(L/kg)] Predicted value: 1.14 [log(L/kg)]
	CAS: 13116-53-5 Dataset id: 256 (training set) SMILES: <chem>C(C(CCl)(Cl)Cl)Cl</chem> Similarity: 0.966 Experimental value: 1.57 [log(L/kg)] Predicted value: 1.37 [log(L/kg)]
	CAS: 78-87-5 Dataset id: 254 (training set) SMILES: <chem>CC(CCl)Cl</chem> Similarity: 0.963 Experimental value: 0.57 [log(L/kg)] Predicted value: 0.9 [log(L/kg)]
	CAS: 3229-00-3 Dataset id: 253 (training set) SMILES: <chem>C(C(CBr)(CBr)CBr)CBr</chem> Similarity: 0.944 Experimental value: 2.62 [log(L/kg)] Predicted value: 2.88 [log(L/kg)]
	CAS: 106-93-4 Dataset id: 249 (training set) SMILES: <chem>C(CBr)Br</chem> Similarity: 0.859 Experimental value: 0.35 [log(L/kg)] Predicted value: 0.84 [log(L/kg)]
	CAS: 109-69-3 Dataset id: 250 (training set) SMILES: <chem>CCCCl</chem> Similarity: 0.858 Experimental value: 1.15 [log(L/kg)] Predicted value: 1.05 [log(L/kg)]

Applicability Domain Index (ADI)

- ✓ Visualization of similar substances
- ✓ Similarity index
- ✓ Descriptors range
- ✓ Atom centered-fragment

- ✓ Check of the descriptor sensitivity
- ✓ Uncertainty

- ✓ Fragments for outliers
- ✓ Prediction Accuracy
- ✓ Prediction Concordance

3.2 Applicability Domain: Measured Applicability Domain Scores



✓	Global AD Index AD Index = 1 Explanation: predicted substance is into the Applicability Domain of the model.
✓	Similar molecules with known experimental value Similarity index = 0.981 Explanation: strongly similar compounds with known experimental value in the training set have been found.
✓	Accuracy (average error) of prediction for similar molecules Accuracy index = 0.19 Explanation: accuracy of prediction for similar molecules found in the training set is good.
✓	Concordance with similar molecules (average difference between target compound prediction and experimental values of similar molecules) Concordance index = 0.384 Explanation: similar molecules found in the training set have experimental values that agree with the target compound predicted value.
✓	Maximum error of prediction among similar molecules Max error index = 0.2 Explanation: the maximum error in prediction of similar molecules found in the training set has a low value, considering the experimental variability.
✓	Atom Centered Fragments similarity check ACF matching index = 1 Explanation: all atom centered fragment of the compound have been found in the compounds of the training set.
✓	Descriptors noise sensitivity analysis Noise Sensitivity = 0.912 Explanation: predictions has a good response to noise scrambling, thus shows a good reliability.
✓	Model descriptors range check Descriptors range check = true Explanation: descriptors for this compound have values inside the descriptor range of the compounds of the training set.

Symbols explanation:

- ✓ The feature has a good assessment, model is reliable regarding this aspect.
- ⚠ The feature has a non optimal assessment, this aspect should be reviewed by an expert.
- ✗ The feature has a bad assessment, model is not reliable regarding this aspect.



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Thank you for your attention!