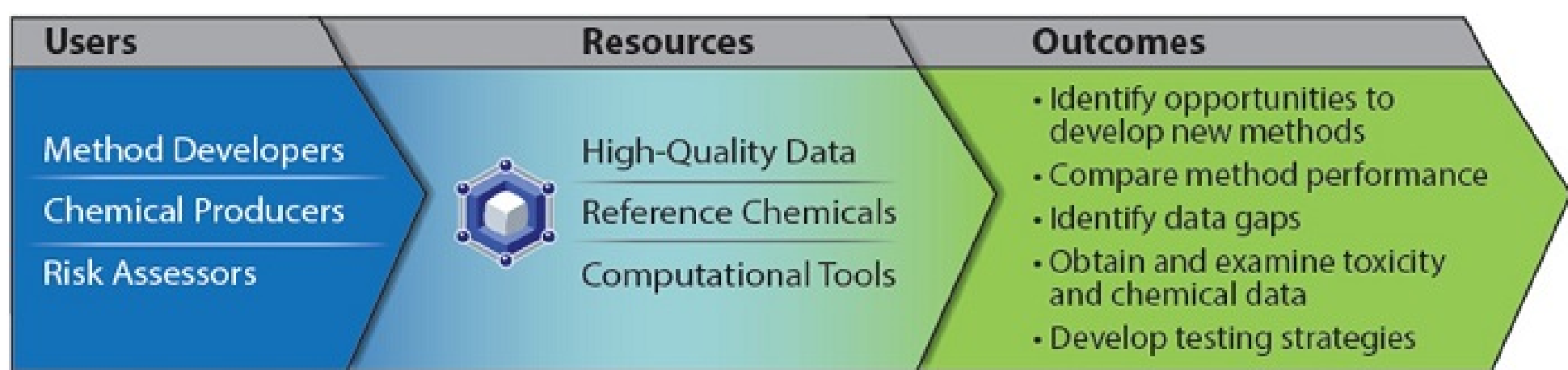


The Integrated Chemical Environment: Tools and Data to Support Toxicity Assessments

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The Integrated Chemical Environment



ICE provides free online access to:

- Curated in vivo and in vitro data related to toxicity testing
- In silico toxicity predictions and chemical property data
- Curated lists of chemicals with defined assays (reference chemical lists)
- Computational tools related to chemical characterization and predicting toxicity

ICE supports:

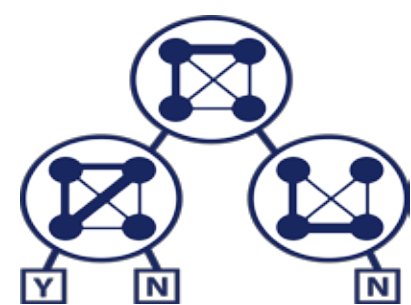
- Data integration: brings together available data, including data on formulations
- Results exploration: enables dynamic, graphical exploration with publication-quality graphics
- Data analysis: allows characterization of data using online workflows
- FAIR (findable, accessible, interoperable and reusable) data access

ICE 2.0

New Features in ICE 2.0:

- Expand your search by adding chemicals in ICE with the same QSAR-ready structures as your chemicals
- Simplified assay selection
- Updated tools

Computational Models



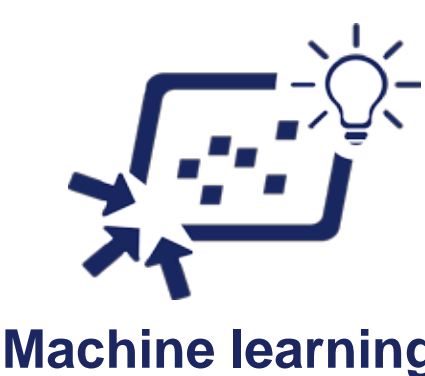
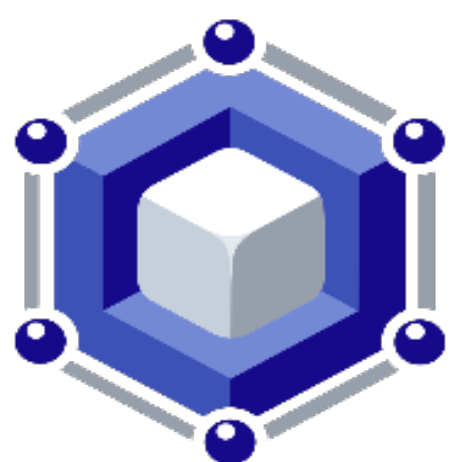
Published Data



Databases



Validation Studies



ICE Integrator

Pop-up assay selection groups assays by common features/toxicity endpoints

Further filter query results and export in a variety of computer-friendly and human-readable formats

Select chemicals to query from chemical quick list and/or entering CASRN

Run tools to explore ICE data and generate predictions

Performance statistics of the model including the confusion matrix (classification) and RMSE (regression) are available to compare method performance

Machine learning tool outputs a table with the predictions from each model along with the data used for training the model

Machine Learning



Use the machine learning tool for hypothesis generation and to explore different machine learning approaches using ICE data. Selected endpoints (for classification and regression modeling) and algorithms are available through ICE to facilitate ease of use by those with limited background in computational toxicology.

Predicts endpoints for in vivo assays:

- Local lymph node assay (skin sensitization)
- Uterotrophic (estrogenic activity)
- Human skin sensitization potency

Machine learning methods available:

- cforest: conditional random forest
- rpart: recursive partitioning
- knn: k-nearest neighbor
- svmRadial: support vector machine with a radial kernel
- pls: partial least squares regression

Models: knn, rf
Metrics: ROC, Sens, Spec
Number of differences: 1
p-value adjustment: bonferroni
Machine Learning Method: knn
Confusion Matrix

Reference	Prediction	Active	Inactive
Active	Active	10	0
Active	Inactive	0	0
Inactive	Active	0	0
Inactive	Inactive	0	0

Sensitivity: 0.9385
Specificity: 0.9744
Pos Pred Value: 0.9839
Neg Pred Value: 0.9048
Prevalence: 0.6250
Detection Rate: 0.5865
Detection Prevalence: 0.5962
Balanced Accuracy: 0.9564
'Positive' Class: Active

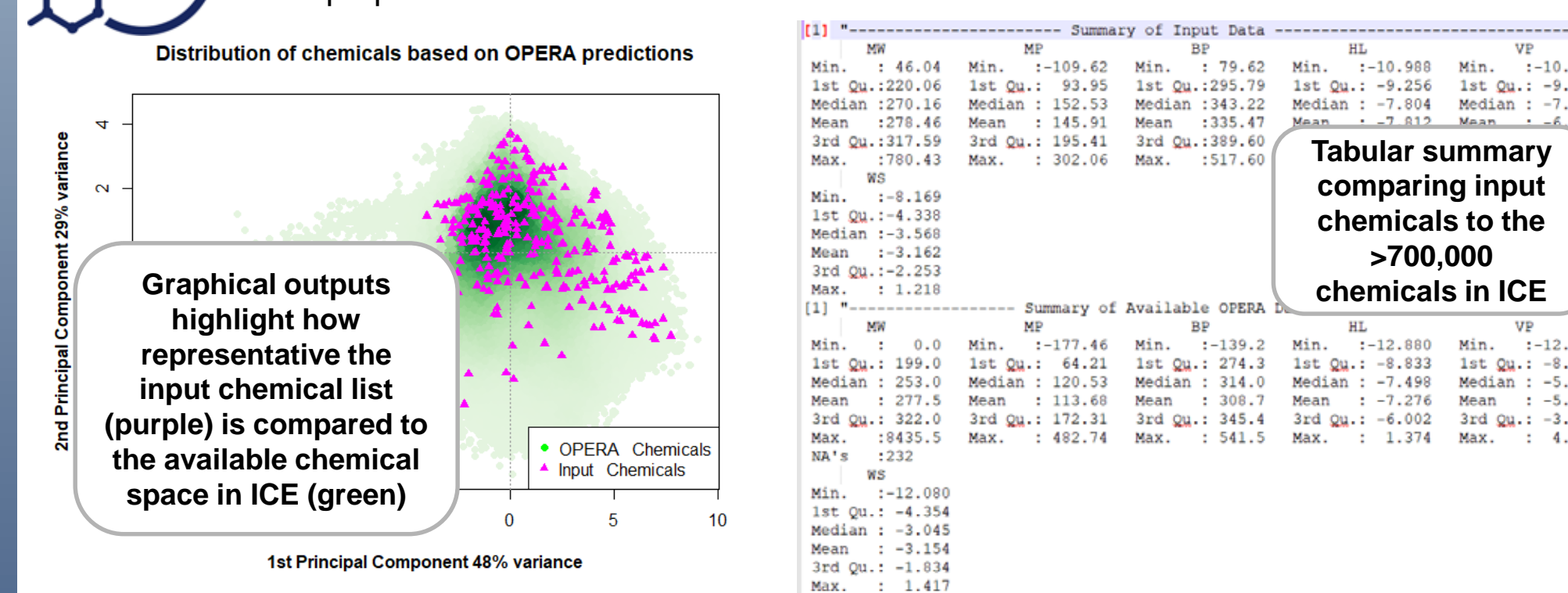
Stand-alone version available for use with custom datasets:
<https://github.com/NIEHS/Machine-Learning-Pipeline>

	A	B	C	D	E
1	CASRN	knn	svmRadial	toPredict	PhysChem
2	95-70-5	Active	Active	Active	
3	26172-55-4	Active	Active	Active	
4	81406-37-1	Active	Active	Active	
5	119-36-8	Inactive	Inactive	Inactive	
6	87-86-5	Active	Active	Active	
7	862892-90	Active	Active	Active	
8	531-95-3	Active	Active	Inactive	
9	66422-95-1	Active	Active	Active	
10	122-57-6	Active	Active	Active	1.820
11	150114-71	Inactive	Inactive	Inactive	0.675

Chemical Space Characterization



Leverage ICE models to characterize a user-supplied chemical list, getting information on the chemical space covered based on different physicochemical properties.

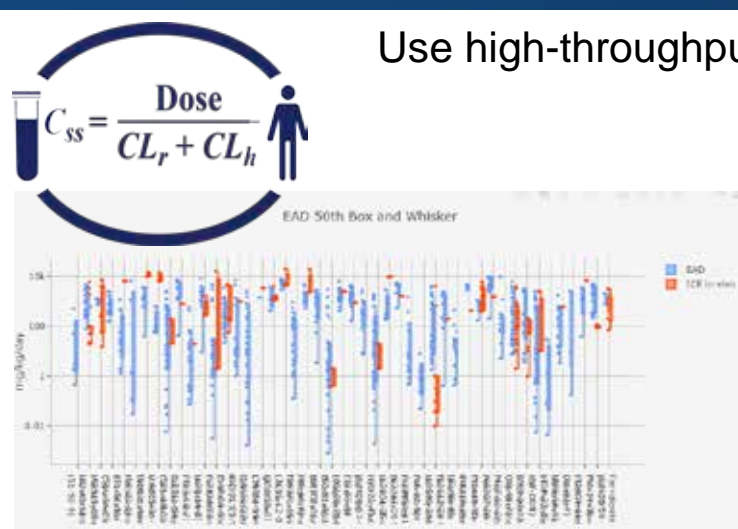


In Vitro to In Vivo Extrapolation

Use high-throughput in vitro data available from ICE to estimate external dose.

Choose from:

- One-compartment pharmacokinetic (1C PK) model including population simulation
- Two three-compartment physiologically-based pharmacokinetic (PBPK) models:
 - 3C Glu: incorporates gut glucuronidation for BPA-family compounds
 - 3C Httk: uses the httk package model
- Three-compartment models include gut, liver, and kidney
- Rat and human predictions



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