

High-throughput experimental and computational technologies at the National Center for Computational Toxicology

<u>Chris Grulke</u>, Antony Williams, John Wambaugh, Richard Judson, Keith Houck and Katie Paul-Friedman

National Center for Computational Toxicology, U.S. Environmental Protection Agency, RTP, NC

The views expressed in this presentation are those of the author and do not necessarily reflect the views or policies of the U.S. EPA

Overview



- The CompTox Chemicals Dashboard webbased database of 875k substances
- Provides access to "high-throughput data"
 - High-throughput bioactivity "ToxCast" data
 - High-throughput prediction "physchem, fate & transport"
 - High-throughput "exposure modeling"
 - High-throughput retrieval of data "batch searching"

Work in progress

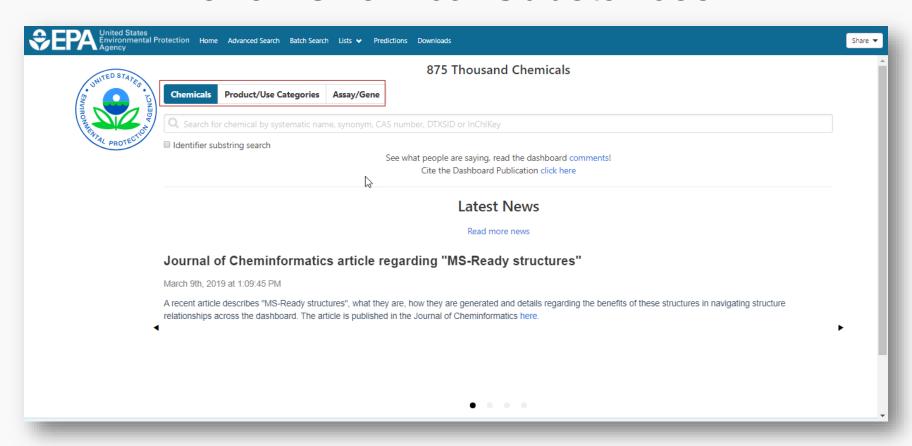
- High-throughput standardization
- High-throughput toxicokinetics
- High-throughput transcriptomics

CompTox Chemicals Dashboard



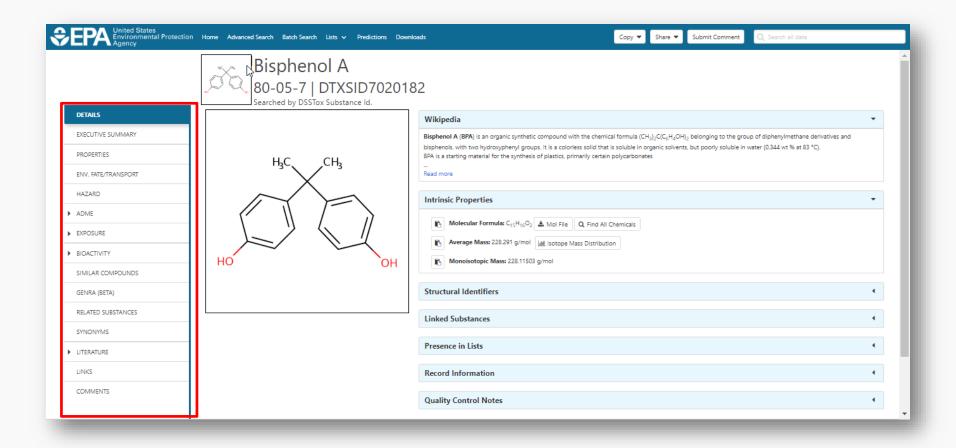


875k Chemical Substances



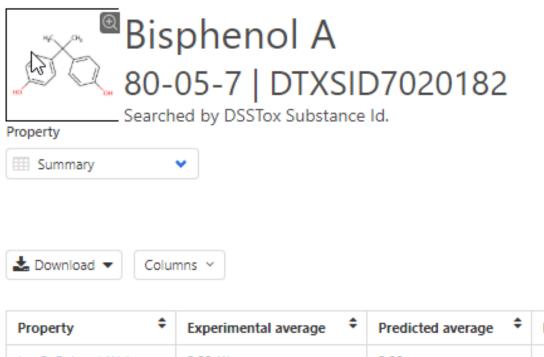
Detailed Chemical Pages





Experimental and Predicted Data





Property	Experimental average \$	Predicted average 🗘 I
LogP: Octanol-Water	3.32 (1)	3.29
Melting Point	155 (7)	139
Boiling Point	200 (1)	363
Water Solubility	5.26e-4 (1)	9.62e-4
Vapor Pressure	-	8.37e-7
Flash Point	-	190

TEST and OPERA Predictions



DETAILS

EXECUTIVE SUMMARY

PROPERTIES

ENV. FATE/TRANSPORT

HAZARD

- ADME
- EXPOSURE
- ▶ BIOACTIVITY

SIMILAR COMPOUNDS

GENRA (BETA)

RELATED SUBSTANCES

SYNONYMS

LITERATURE

LINKS

COMMENTS

Toxicity Estimation Software Tool (TEST)

On this page:

- QSAR Methodologies
- What's New in Version 4.2.1?
- Prior Version History
- · System Requirements
- · Installation Instructions
- Publications
- Get Email Alerts

The Toxicity Estimation Software Tool (TEST) was developed to allow users to easily estimate the toxicity of chemicals using Quantitative Structure Activity Relationships (QSARs) methodologies. QSARs are mathematical models used to predict measures of toxicity from the physical characteristics of the structure of chemicals (known as molecular descriptors). Simple QSAR models calculate the toxicity of chemicals using a simple linear function of molecular descriptors:

Ask a Technical Expert

Got a question about our research model? Want to give us feedback? Contact a technical expert about <u>TEST</u>.

Mansouri et al. J Cheminform (2018) 10:10 https://doi.org/10.1186/s13321-018-0263-1

Journal of Cheminformatics

RESEARCH ARTICLE

Open Access

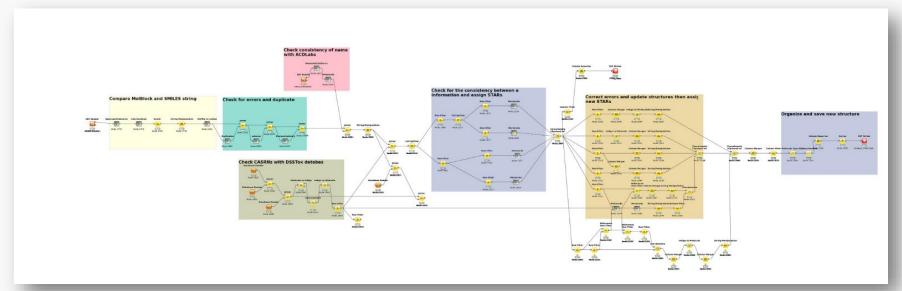
OPERA models for predicting physicochemical properties and environmental fate endpoints

Kamel Mansouri 1,2,3* , Chris M. Grulke 1, Richard S. Judson 1 and Antony J. Williams 1

Standardization Approaches



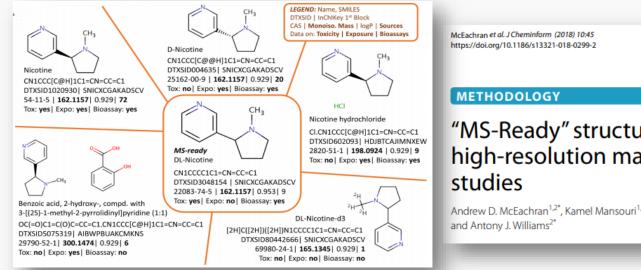
- "QSAR-ready" forms of chemicals are need for pushing through predictive models
 - OPERA predictions
 - TEST predictions
 - All QSAR models we build



Standardization Approaches



- We standardize chemicals to enable searches
- "MS-ready" forms to facilitate structure ID
 - Separate multi-component chemicals
 - Destereo, desalt, de-isotope



Sachran et al. J Cheminform (2018) 10:45
ps://doi.org/10.1186/s13321-018-0299-2

Journal of Cheminformatics

"MS-Ready" structures for non-targeted high-resolution mass spectrometry screening

Andrew D. McEachran^{1,2*}, Kamel Mansouri^{1,2,3}, Chris Grulke², Emma L. Schymanski⁴, Christoph Ruttkies⁵ and Antony J. Williams^{2*}

Tweak algorithms and bulk process all data

Open Access

Present status For every data release...



- Presently every data release requires
 - Produce standardized data (MS and QSAR-ready)
 - QSAR predictions TEST and OPERA
 - Systematic name generation
- Registration of data into databases
- Time-consuming and lots of overhead
- Not yet supporting other models for highthroughput

TEST Web Services

https://www.epa.gov/sites/production/files/2018-08/documents/webtest_users_guide.pdf





https://comptox.epa.gov/dashboard/web-test/

User's Guide for WebTEST (version 1.0) (Web-services Toxicity Estimation Software Tool)

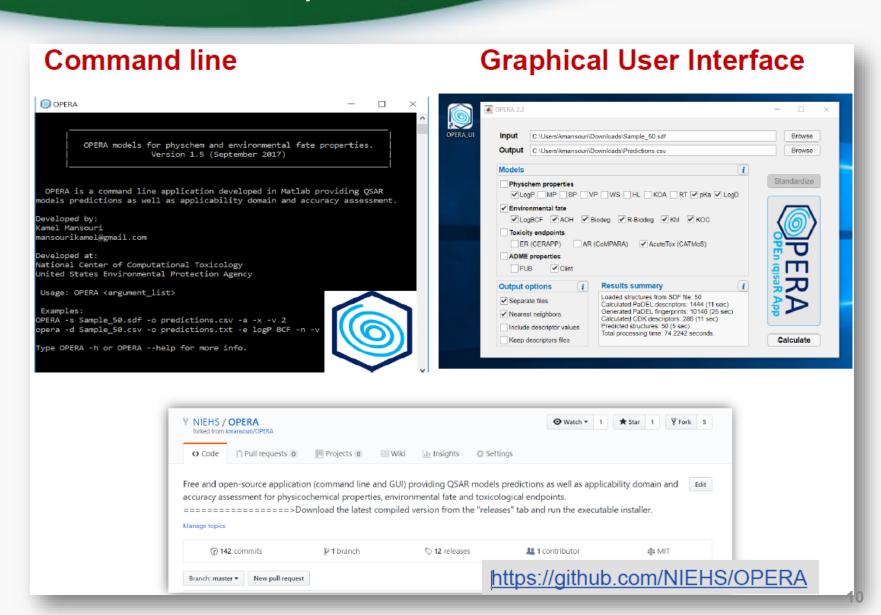
Integration to registration system required for high-throughout processing

A Web-Sei from Mole

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OPERA Standalone Application But services are required





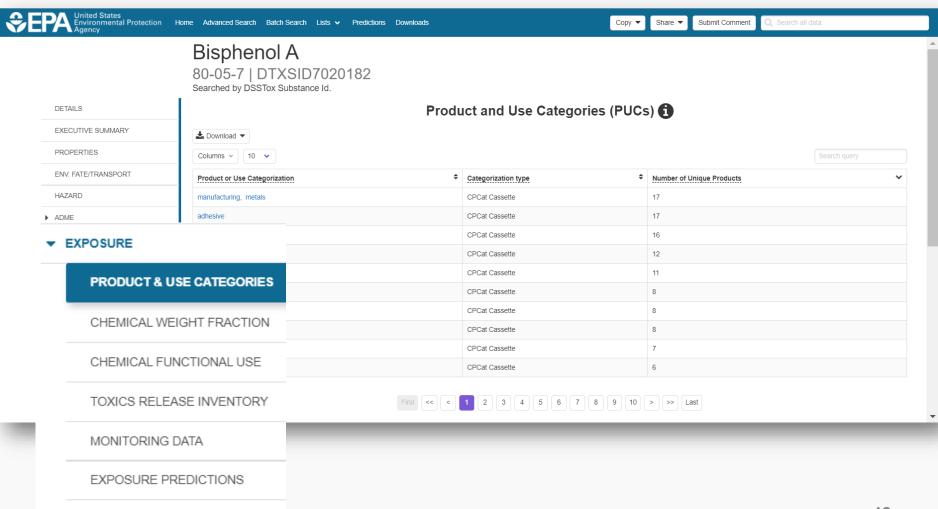


Exposure Data

Sources of Exposure to Chemicals

PRODUCTION VOLUME





Exposure Modeling



Rapid Exposure Predictions

Rapid, also called high-throughput, exposure predictions or ExpoCast provide rapid exposure estimates for thousands of chemicals. ExpoCast quickly and efficiently looks at multiple routes of exposure to provide exposure estimates. ExpoCast uses, enhances and evaluates two well-known exposure models to provide exposure predictions.

Farfield Exposure Models

Farfield exposure models are used to predict exposure from chemicals that are released into
the outdoor environment through industrial releases. ExpoCast uses "off-the-shelf" models,
USETox and RAIDAR, to estimate outdoor environment exposures. These models estimate
the average amount of chemical that gets into the air, water, and soil. The estimates from these models are used in combination with the
estimates from the nearfield models to make exposure predictions.

Nearfield Exposure Models

Nearfield exposure models provide estimates of exposure to chemicals used in consumer products and in-home products. The model used to estimate the range of total chemical exposures in a population is the EPA's Stochastic Human Exposure and Dose Simulation (SHEDS) model. There is a SHEDS high-throughput model that estimates exposure for thousands of chemicals, and a more precise traditional SHEDS model which needs more input data to make more accurate exposure predictions.

Pictured Above: Farfield Exposure Examples



Pictured Above: Examples of Nearfield Exposure,

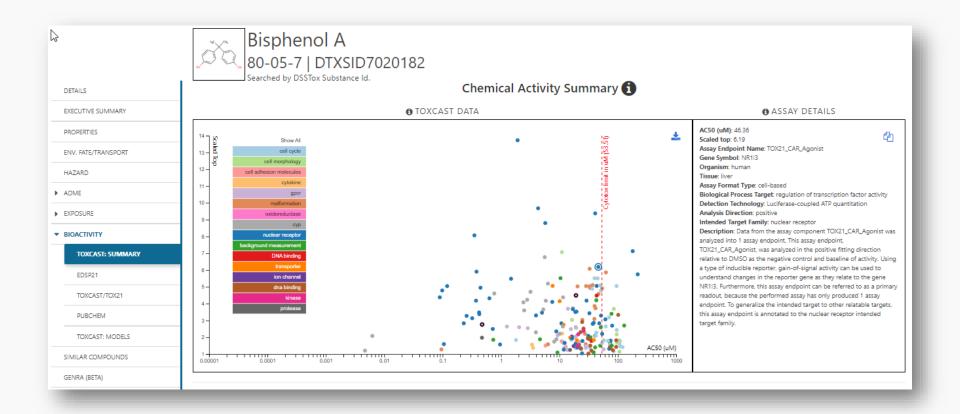


In vitro bioactivity data

In Vitro Bioassay Screening

ToxCast and Tox21

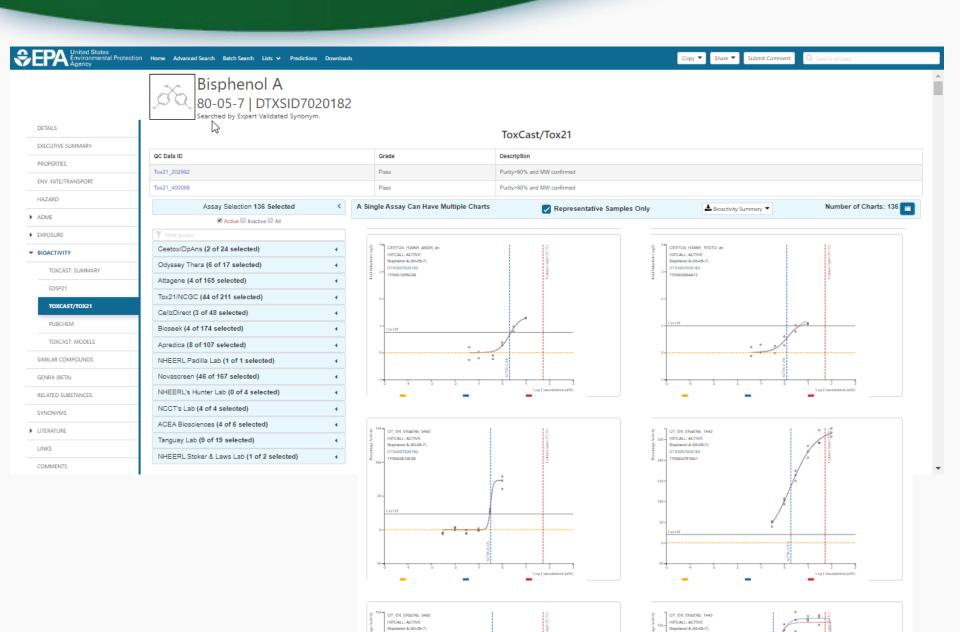




In Vitro Bioassay Screening

ToxCast and Tox21





Bioactivity: Downloadable Data

https://www.epa.gov/chemical-research/exploring-toxcast-data-downloadable-data



Exploring ToxCast Data: Downloadable Data

The results after processing through the Pipeline are available on the <u>ToxCast Dashboard</u>, and for most users EPA recommends accessing the data there.

- ToxCast Chemicals
- <u>ToxCast Assays</u>

ToxCast Data and Information

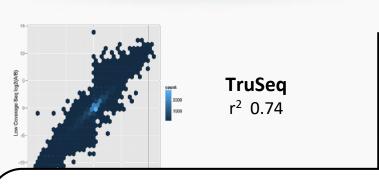
- ToxCast & Tox21 Summary Files. Data for a single chemical endpoint pair for thousands of chemicals and assay endpoints for 20 variables such as the activity or hit call, activity concentrations, whether the chemical was tested in a specific assay, etc.
 - o <u>Download ToxCast Summary Information</u>
 - Download ReadMe
- ToxCast & Tox21 Data Spreadsheet. A spreadsheet of EPA's analysis of the chemicals screened through ToxCast and the Tox21 collaboration which includes EPA's activity calls from the screening of over 1,800 chemicals.
 - Download Data
 - Download ReadMe
- ToxCast Data Pipeline R Package. The R computer programming package used to process and model all EPA ToxCast and Tox21 chemical screening data. The files include the R programming package as well as documents that provide overviews of the data analysis pipeline used and the R package. Users will need experience with R to use these files.
 - <u>Download Package</u>
 - TCPL Overview

Resources

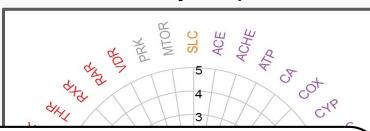
- <u>Toxicity Forecaster (ToxCast)</u>
 <u>Fact Sheet</u>
- ToxCast Publications
- ToxCast Citation
- About ToxCast

Transcriptomics Data will Deliver Terabytes of Data for Analysis



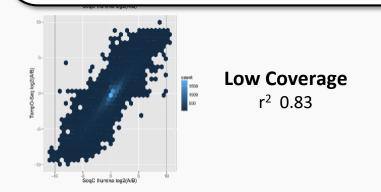


MOA Analysis Pipeline





- Large scale screen of 1,000 chemicals (ToxCast I/II) in single cell type this summer
- Additional screens across multiple cell types/lines
- Additional reference chemicals and genetic perturbations (RNAi/CRISPR/cDNA)





Currently capable of assigning to >40 MOAs based on transcriptional responses



"High-Throughput Searching"

Batch Searching



 Singleton searches are useful but people generally want data on LOTS of chemicals!

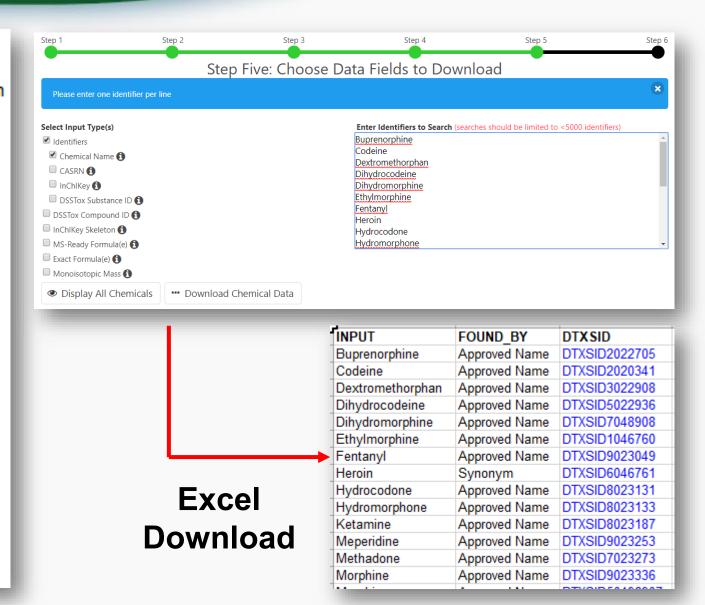
Typical questions

- What is the list of chemicals for the formula C_xH_yO_z
- What is the list of chemicals for a mass +/- error
- Can I get chemical lists in Excel files? In SDF files?
- Can I include properties in the download file?

Batch Search Names



Buprenorphine Codeine Dextromethorphan Dihydrocodeine Dihydromorphine Ethylmorphine Fentanyl Heroin Hydrocodone Hydromorphone Ketamine Meperidine Methadone Morphine Morphinone Naloxone Naltriben 0xycodone Oxymorphone Propoxyphene Sufentanil Tramadol



Add Other Data of Interest



Chemical Identifiers

- ✓ DTXSID
- Chemical Name
- ☐ DTXCID **(**)
- ✓ CAS-RN
- ✓ InChlKey <a>f
- ☐ IUPAC Name 🚹

Structures

- ☐ Mol File 🚯
- SMILES 1
- ☐ InChI String **1**
- ✓ MS-Ready SMILES
- QSAR-Ready SMILES (1)

Intrinsic And Predicted Properties

- Molecular Formula 6
- Average Mass < 1</p>
- ✓ Monoisotopic Mass

 ⑥
- TEST Model Predictions
- OPERA Model Predictions

INPUT	DTXSID	CASRN	MOLECULAR FO	MONOISOTOPIC	MS READY SMI
	DTXSID202		C29H41NO4		[H]C12CC3=C4C
	DTXSID202		C18H21NO3		[H]C12CC3=C4C
Dextrometh	DTXSID302	125-71-3	C18H25NO		[H]C12CC3=C(C=
Dihydrocode	DTXSID502	125-28-0	C18H23NO3	301.1677936	[H]C12CC3=C4C(
Dihydromor	DTXSID704	509-60-4	C17H21NO3	287.1521435	[H]C12CC3=C4C(
Ethylmorph	DTXSID104	76-58-4	C19H23NO3	313.1677936	[H]C12CC3=C4C
Fentanyl	DTXSID902	437-38-7	C22H28N2O	336.2201635	CCC(=O)N(C1CC
Heroin	DTXSID604	561-27-3	C21H23NO5	369.1576228	[H]C12CC3=C4C
Hydrocodon	DTXSID802	125-29-1	C18H21NO3	299.1521435	[H]C12CC3=C4C
Hydromorph	DTXSID802	466-99-9	C17H19NO3		[H]C12CC3=C4C
Ketamine	DTXSID802	6740-88-1	C13H16CINO	237.0920418	CNC1(CCCCC1=
Meperidine	DTXSID902	57-42-1	C15H21NO2	247.1572289	CCOC(=0)C1(CC
Methadone	DTXSID702	76-99-3	C21H27NO	309.2092645	CCC(=O)C(CC(C)
Morphine	DTXSID902		C17H19NO3	285.1364935	[H]C12CC3=C4C
Morphinone	DTXSID501	467-02-7	C17H17NO3	283.1208434	[H]C12CC3=C4C
Naloxone	DTXSID802	465-65-6	C19H21NO4	327.1470582	[H]C12CC3=C4C
Naltriben	-	-	-	-	-
	DTXSID502		C18H21NO4		[H]C12CC3=C4C(
	DTXSID502		C17H19NO4		[H]C12CC3=C4C
	DTXSID102		C22H29NO2		CCC(=O)OC(CC1
Sufentanil			C22H30N2O2S		CCC(=O)N(C1=C
Tramadol	DTXSID908	27203-92-5	C16H25NO2	263.188529	COC1=CC=CC(=

Access to other data



Chemical Identifiers

- ✓ DTXSID
- Chemical Name
- ☐ DTXCID **(**)
- ✓ CAS-RN
- ✓ InChlKey <a>1
- ☐ IUPAC Name 🚹

Structures

- ☐ Mol File 🚯
- SMILES 1
- InChl String 6
- ✓ MS-Ready SMILES
- QSAR-Ready SMILES 6

Intrinsic And Predicted Properties

- Molecular Formula 6
- Average Mass <a>6
- Monoisotopic Mass 6
- ☐ TEST Model Predictions **1**
- OPERA Model Predictions

Pre-predicted TEST and OPERA data

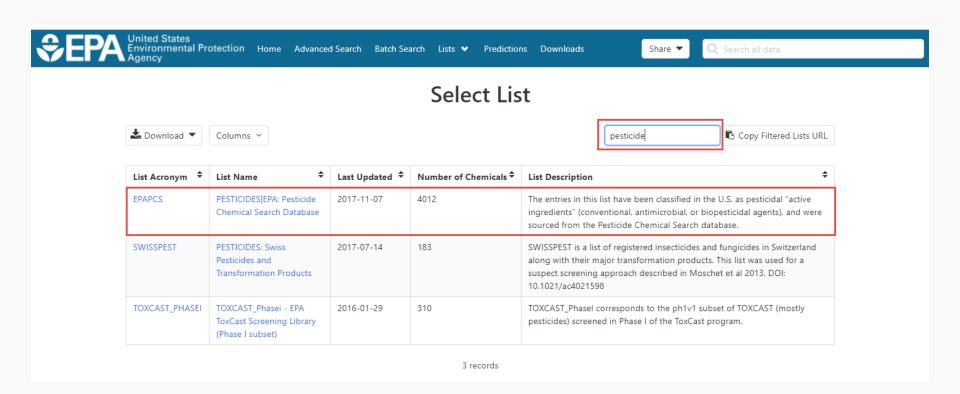
С	D	Е	F	G	Н		J	K
DTXSID	PREFERRE	BIOCONCE	BOILING_P	48HR_DAPI	DENSITY_C	DEVTOX_T	96HR_FATI	FLASH_PC
DTXSID7020009	Acetonitrile	2.04	74.44	0.0064565	0.92	0.54	0.0075858	22.46
DTXSID5020023	Acrolein	-	52.443	0.0006427	0.841	0.641	0.0001581	8.913
DTXSID5020029	Acrylonitrile	-	107.213	0.0002877	0.901	0.642	0.0002193	19.044
DTXSID8020044	Allyl alcohol	1.87932	83.788	0.000975	0.858	0.719	0.0011912	21.313
DTXSID8020090	Aniline	5.3827	192.271	3.184E-05	1.108	0.529	0.000869	98.93
DTXSID8020173	bis(Chlorom	3.47536	102.976	0.001205	1.233	0.308	0.0017824	35.353
DTXSID8020250	Carbon tetra	17.2982	91.367	0.000104	1.601	0.777	0.00014	21.094
DTXSID1020273	Chlorine	-	-	-	-	-	-	-
DTXSID4020292	Chloroaceta	-	73.604	0.0011169	1.162	0.797	0.0002897	49.401
DTXSID9020297	(2-Chlorobe	-	278.468	4.808E-06	1.286	0.859	2.559E-06	164.565
DTXSID4020298	Chlorobenze	108.393	150.5	6.699E-05	1.123	0.42	0.0001762	49.082
DTXSID1020306	Chloroform	6.71429	61.07	0.0008395	1.425	0.794	0.0007079	24.417
DTXSID6020307	Chloromethy	2.29087	69.641	0.0038726	1.126	0.426	0.0048306	16.663
DTXSID6020351	(E)-Crotona	-	97.879	0.0004764	0.871	0.6	0.0001318	19.542
DTXSID2020422	1,1-Dichloro	8.97429	33.566	0.0006138	1.214	0.464	0.0007638	7.323
DTXSID6020515	N,N-Dimethy	0.677642	166.494	0.0048195	0.947	0.94	0.0141906	77.204
DTXSID1020516	1,1-Dimethy	-	97.361	0.002958	0.914	0.515	0.0185353	21.051
DTXSID1020566	Epichlorohyo	3.12608	102.31	0.000743	1.214	0.66	0.0005023	21.312
DTXSID8020599	Ethyleneimir	-	28.496	-	-	0.73	-	-2.929
DTXSID0020600	Ethylene oxi	0.724436	34.921	0.0043551	0.968	0.758	0.0037497	15.445
DTXSID6020646	Furan	4.88652	73.387	0.0020845	0.997	0.394	0.0021232	-1.928



So what about Pesticides??

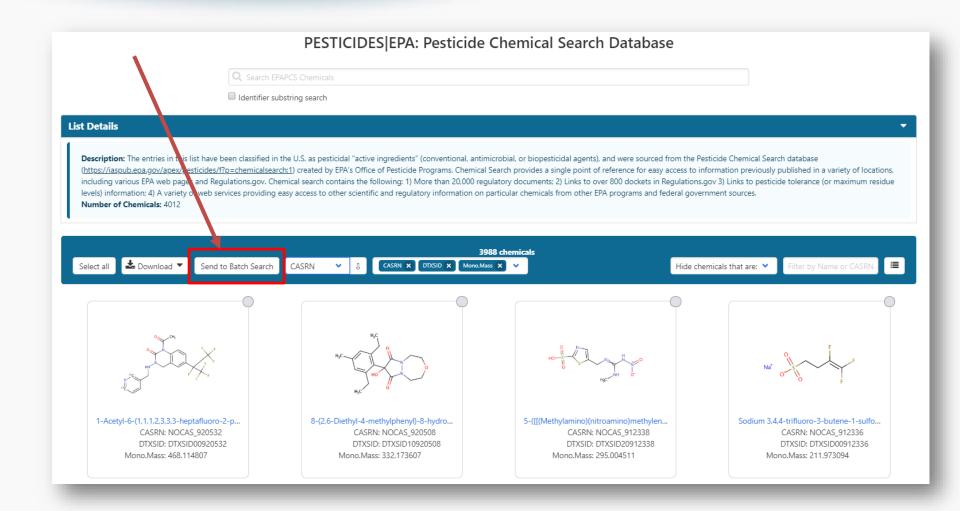
Curated List of Pesticides





Curated List of Pesticides





Send to batch and select....



Intrinsic And Predicted Properties
Molecular Formula 🐧
Average Mass 🐧
Monoisotopic Mass 6
TEST Model Predictions 6
OPERA Model Predictions
Metadata
Curation Level Details 🚯
☐ NHANES/Predicted Exposure 1
Data Sources 🚯
Include ToxVal Data Availability 6
Assay Hit Count 🚯
Number of PubMed Articles 6
☐ PubChem Data Sources 1
CPDat Product Occurrence Count 🚯
□ IRIS (1)
PPRTV 1

3988 Pesticide....Batch Search



- A few seconds to assemble
 - ToxCast data #actives/#assays and % active
 - # articles in PubMed
 - Links to IRIS or PPRTV reports
 - TEST or OPERA predictions
 - Availability of exposure data, presence in consumer products, EXPOCAST predictions

Α	В	С	D	E	F	G	H		J	K L	M
DTXSID	PREFERRED_NAME	EXPOCAST_MEI	EXPOCAST	NHANES	TOXVAL_DATA	TOXCAST_%_ACT	TOXCAST	#PUBMED_	PUBCHEM_	CPDAT_COUNTRIS_LINK	PPRTV_LINK
DTXSID2021105	Pentachloronitrobenzene	1.14e-07	Υ	Υ	Υ	11.8	99/839	69	96	164 Y	-
DTXSID4022527	Propylparaben	1.4e-05	Υ	Υ	Υ	13.77	99/719	201	121	1476 -	-
DTXSID4024064	Dinex	8.29e-08	Υ	-	Υ	42.13	99/235	-	35	5 Y	-
DTXSID0032493	Triadimenol	1.73e-08	Υ	-	Υ	10.54	98/930	163	74	83 -	-
DTXSID4032667	Esfenvalerate	1.7e-06	Υ	-	Υ	11.45	98/856	483	45	198 -	-
DTXSID6020561	Endrin	1.29e-07	Υ	-	Υ	14.02	98/699	284	16	98 Y	Υ
DTXSID6025355	Glutaraldehyde	2.03e-05	Υ	_	Υ	14.35	98/683	6515	139	1144 -	-
DTXSID8032417	Isofenphos	1.87e-08	Υ	_	Υ	16.28	98/602	30	42	60 -	-
DTXSID6032352	Chlorpyrifos-methyl	1.07e-07	Υ	Υ	Υ	11.27	97/861	72	50	116 -	-
DTXSID8020620	Fenthion	8.99e-08	Υ	Υ	Υ	11.56	97/839	354	100	99 -	-
DTXSID2020189	FD&C Blue No. 1	0.000178	Υ	-	Υ	13.72	97/707	174	49	672 -	-
DTXSID7044843	Erythrosin B	6.3e-07	Υ	-	-	24.25	97/400	14843	51	7 -	-
DTXSID5041778	Chloropropylate	1.05e-07	Υ	-	Υ	40.93	97/237	-	36	12 -	-
DTXSID5023900	Benomyl	1.11e-07	Υ	-	Υ	11.23	96/855	476	91	105 Y	-
DTXSID9020247	Carbaryl	5.61e-08	Υ	Υ	Υ	11.51	96/834	1135	117	245 Y	-
DTXSID8024109	Flutolanil	1.63e-08	Υ	-	Υ	11.4	95/833	6	59	80 -	-
DTXSID1023998	Cypermethrin	1.62e-06	Υ	Υ	Υ	10.78	94/872	1148	148	246 -	-
DTXSID2024242	Paclobutrazol	9.19e-08	Υ	-	Υ	11.11	94/846	139	-	40 Y	-
DTXSID1020807	2-Mercaptobenzothiazole	4.7e-05	Υ	-	Υ	12.82	94/733	111	181	86 -	Υ



Other packages available from NCCT

High-Throughput Toxicokinetics

https://cran.r-project.org/web/packages/httk/index.html



Authors: Robert G. Pearce, R. Woodrow Setzer, Cory L. Strope, Nisha S. Sipes, John F.

Wambaugh

Title: httk: R Package for High-Throughput Toxicokinetics

httk: High-Throughput Toxicokinetics

Functions and data tables for simulation and statistical analysis of chemical toxicokinetics ("TK") as in Pearce et al. (2017) < doi:10.18637/jss.v079.i04. Chemical-specific in vitro data have been obtained from relatively high throughput experiments. Both physiologically-based ("PBTK") and empirical (e.g., one compartment) "TK" models can be parameterized for several hundred chemicals and multiple species. These models are solved efficiently, often using compiled (C-based) code. A Monte Carlo sampler is included for simulating biological variability (Ring et al., 2017 < doi:10.1016/j.envint.2017.06.004) and measurement limitations. Calibrated methods are included for predicting tissue:plasma partition coefficients and volume of distribution (Pearce et al., 2017 < doi:10.1007/s10928-017-9548-7). These functions and data provide a set of tools for in vitro-in vivo extrapolation ("IVIVE") of high throughput screening data (e.g., Tox21, ToxCast) to real-world exposures via reverse dosimetry (also known as "RTK") (Wetmore et al., 2015 < doi:10.1093/toxsci/kfv171).

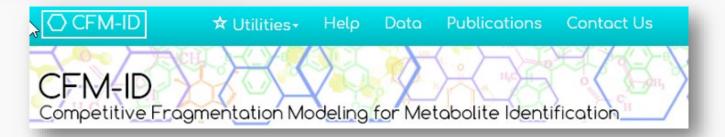


Work in Progress

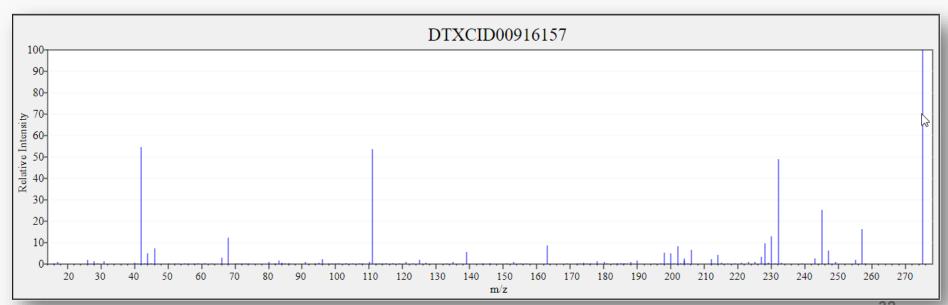
Predicted Mass Spectra

http://cfmid.wishartlab.com/





 MS/MS spectra prediction for ESI+, ESI-, and EI to support candidate ranking for mass/formula searches



High-Throughput Spectral Prediction CFM-ID Predicted Library



Predictions already generated for >700k structures

Linking in silico MS/MS spectra with chemistry data to improve identification of unknowns

Andrew D. McEachran , Ilya Balabin, Tommy Cathey, Thomas R. Transue, Hussein Al-Ghoul, Chris Grulke, Jon R. Sobus & Antony J. Williams

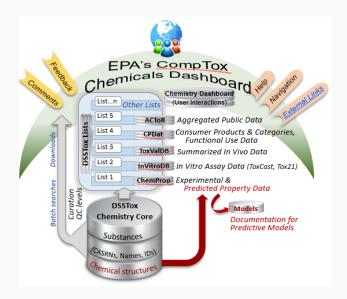
Scientific Data 6, Article number: 141 (2019) | Download Citation

 Ideally CFM-ID spectra will be generated for all chemicals added to our registration system

Conclusion



- High-throughput data generation comes in many forms at the National Center for Computational Toxicology
 - High-throughput bioactivity screening ToxCast/Tox21
 - High-throughput transcriptomics data
 - High-throughput standardization (MS and QSAR-ready processing)
 - High-throughput QSAR prediction (e.g. TEST and OPERA models)
 - High-throughput Mass Spec fragmentation prediction (CFM-ID)
- Pipelining and versioning of data is in scope as future work



Acknowledgements





EPA-RTP

- An enormous team of contributors from NCCT, especially the IT software development team
- Our curation team for their care and focus on data quality
- Multiple centers and laboratories across the EPA
- Many public domain databases and open data contributors