

# An Introduction to the EPA CompTox Chemicals Dashboard and GenRA Read-across Module

**Antony John Williams and Grace Patlewicz** 

Center for Computational Toxicology and Exposure, US-EPA, RTP, NC

## The Charge for the Dashboard

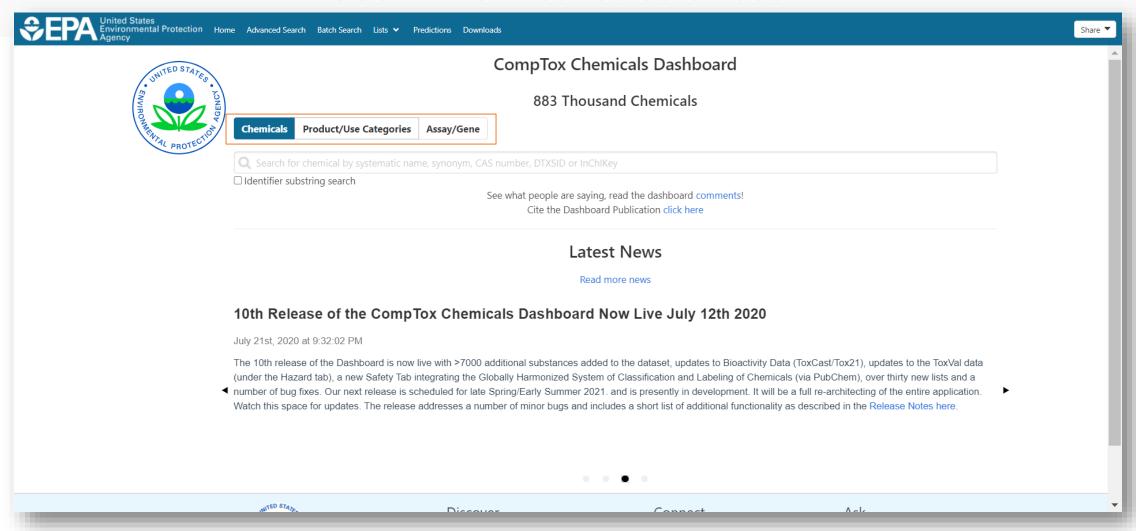


- Develop a "first-stop-shop" for environmental chemical data to support EPA and partner decision making:
  - Centralized location for relevant chemical data
  - Chemistry, exposure, hazard and dosimetry
  - Combination of existing data and predictive models
  - Publicly accessible, periodically updated, curated
- Easy access to data improves efficiency and ultimately accelerates chemical risk assessment

#### CompTox Chemicals Dashboard



#### 883k Chemical Substances



#### **BASIC Search**



**Product/Use Categories** Chemicals Assay/Gene Q Benzo(a)pyrene Benzo(a)pyrene DTXSID2020139 Benzo(a)pyrene diolepoxide 1 DTXSID9036779 Benzo(a)pyrene-7,8,9-triol,7,8,9,10-tetrahydro-, (7-alpha,8-beta,9-beta)-DTXSID00210066 Benzo(a)pyrene-1-methanol DTXSID40235374 Benzo(a)pyrene-1,6-dione, 7-methyl-DTXSID70229645 Benzo(a)pyrene-10-methanol DTXSID20235817 Benzo(a)pyrene-10-sulfonic acid, 7,8,9,10-tetrahydro-7,8,9-trihydroxy-, (7alpha,8beta,9beta DTXSID80154378 Benzo(a)pyrene-11,12-diol DTXSID70215609 Benzo(a)pyrene-11,12-diol, 11,12-dihydro-, cis-DTXSID20214501

- Type ahead search using Names, synonyms and CASRNs
- Millions of identifiers
- Substring search

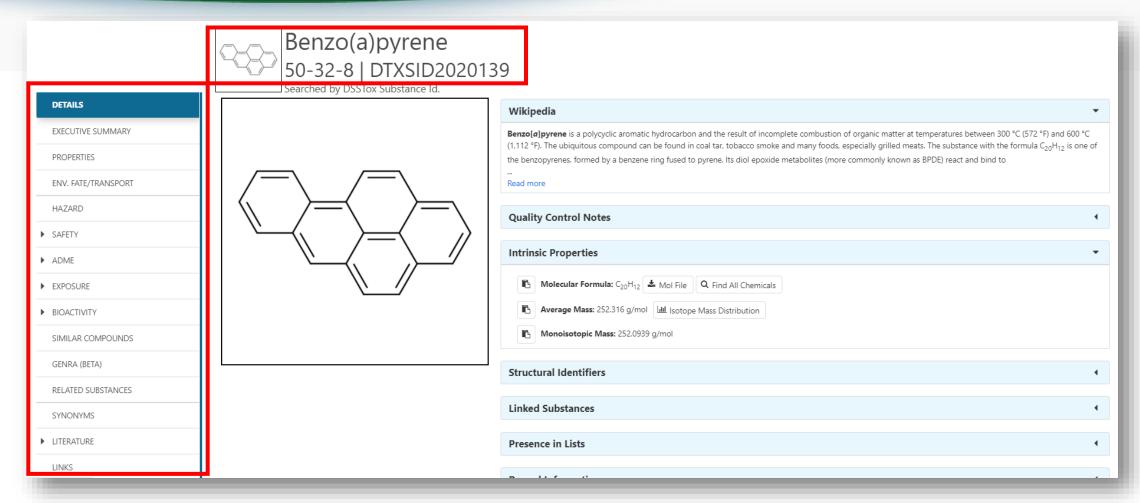
#### Search Results

Searched with 'Synonym Substring': Benzo(A)Pyrene

183 chemicals

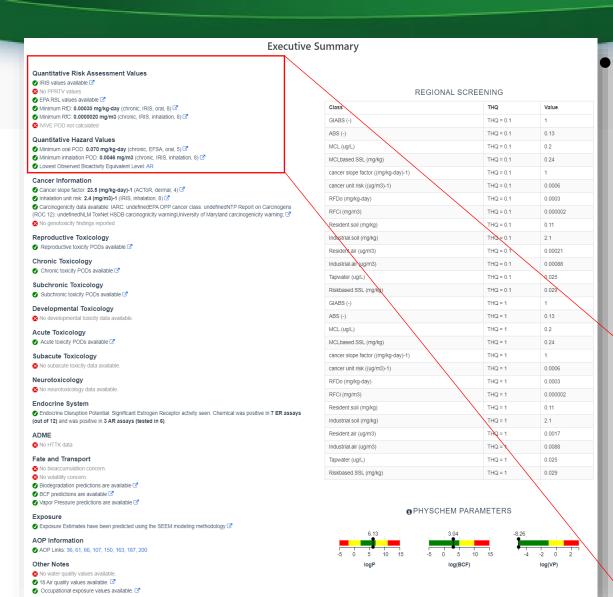
# Detailed Chemical Pages One more identifier – the DTXSID





• Chemical page: Wikipedia snippet when available, intrinsic properties, structural identifiers, linked substances

#### "Executive Summary"



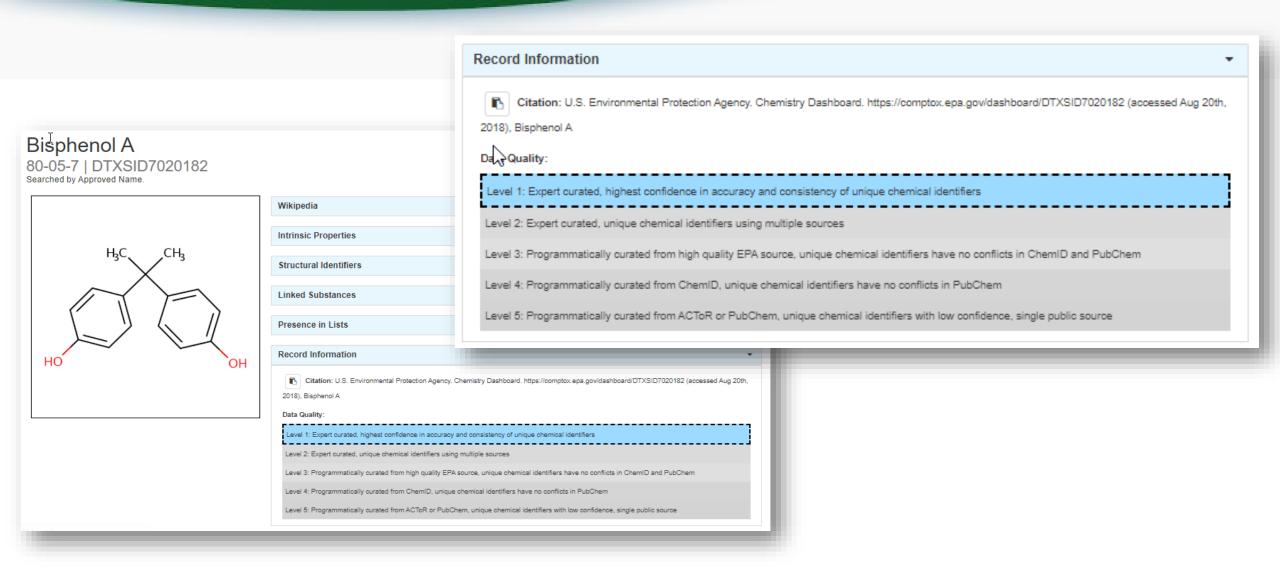


- Overview of toxicity-related info
  - Quantitative values
  - Info re. toxicology subsets
  - Physchem. and Fate & Transport
  - Adverse Outcome Pathway links
  - *In vitro* bioactivity summary plot



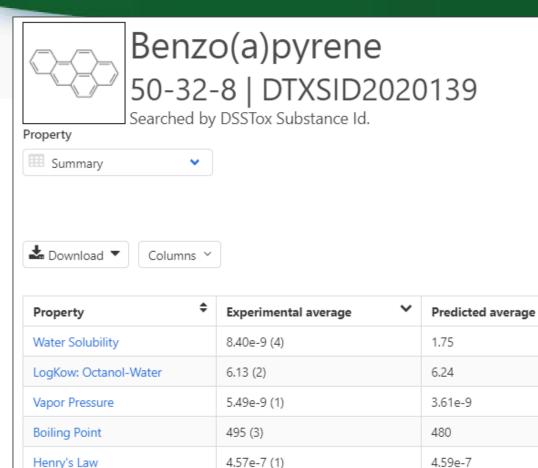
#### Record Information Quality Flags





#### **Experimental and Predicted Data**





177 (8)

189

53.9

234

1.28

Melting Point

Flash Point

Density

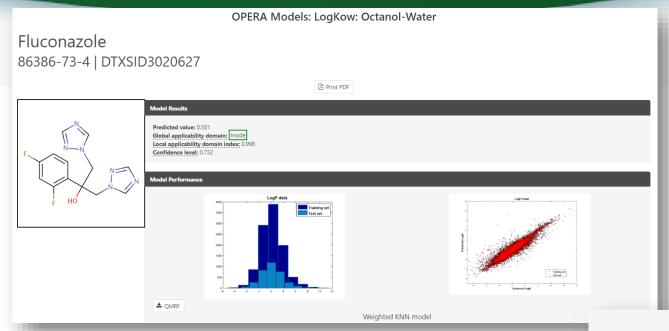
Surface Tension

- Physchem and Fate & Transport experimental and predicted data
- Data can be downloaded as Excel, TSV and CSV files

- Predictions: multiple algorithms
  - EPI Suite: Estimation Program Interface
  - ACD/Labs (commercial)
  - TEST: Toxicity Estimation Software Tool
  - OPERA: OPEn structure—activity/ property Relationship App

## **OPERA Reports**

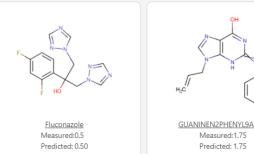




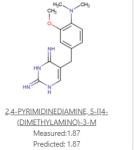
#### Weighted KNN model

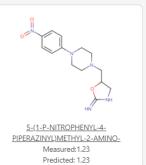
5-fold	CV (75%)	Trainir	ng (75%)	Test (25%)		
Q2 RMSE		R2	RMSE	R2	RMSE	
0.850	0.690	0.860	0.670	0.860	0.780	

#### Nearest Neighbors from the Training Set



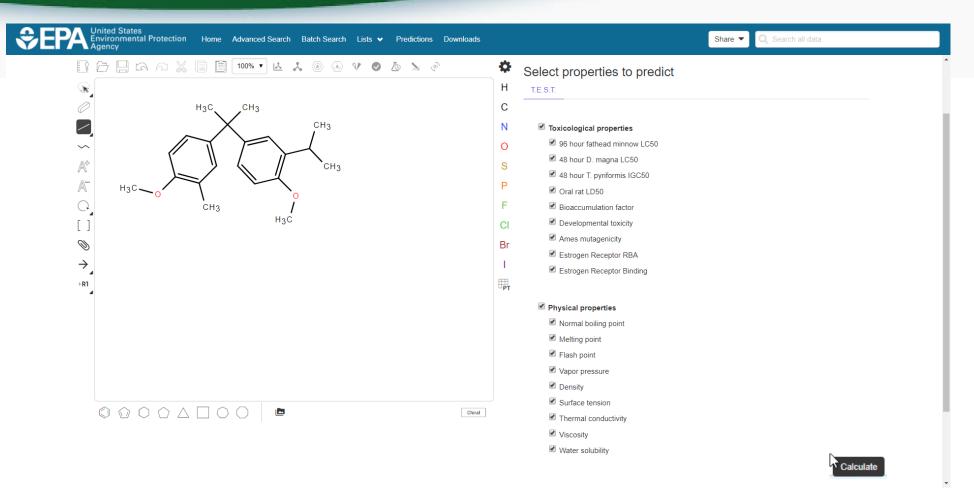






#### Real-Time Predictions



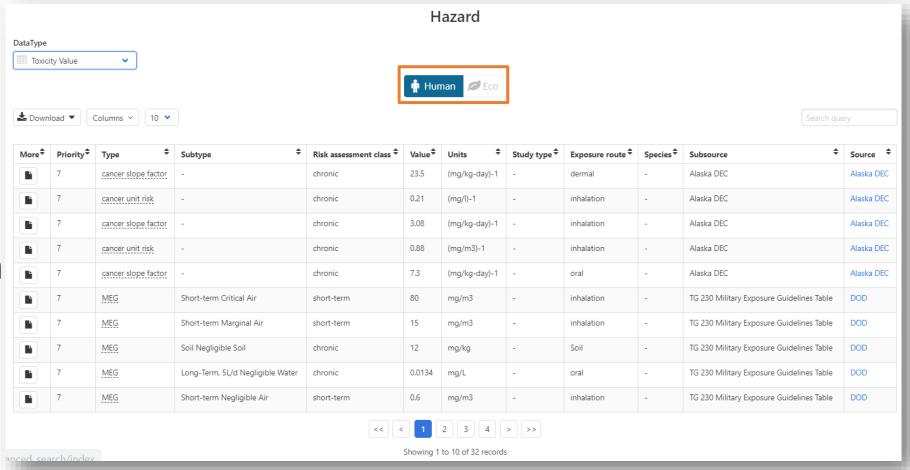


#### Chemical Hazard Data



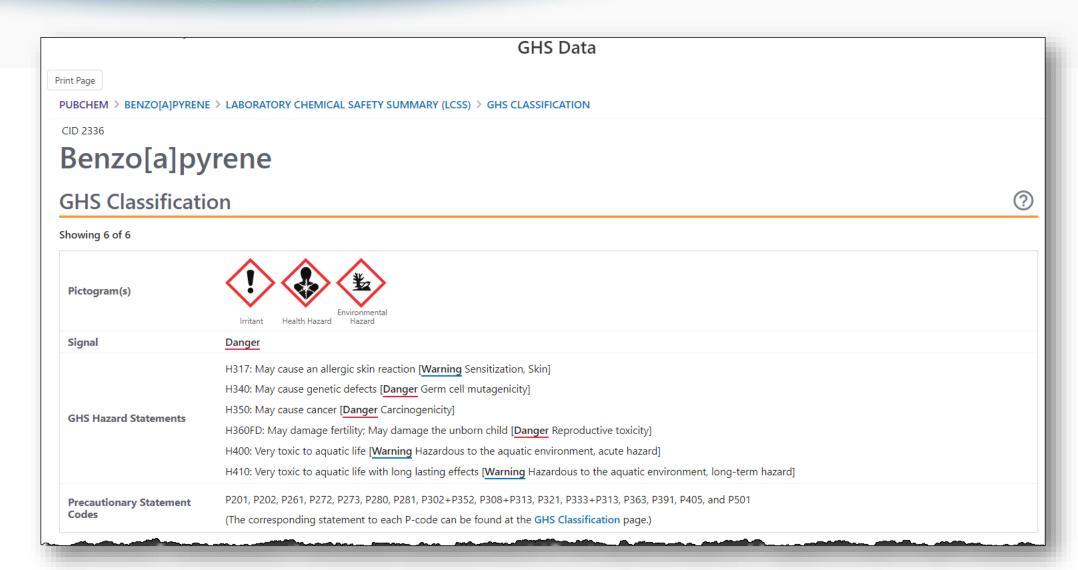
#### **ToxVal Database**

- >50k chemicals
- >770k tox. values
- >30 sources of data
- ~5k journals cited
- ~70k citations



## Safety Data





#### Sources of Exposure to Chemicals





#### Benzo(a)pyrene 50-32-8 | DTXSID2020139

Searched by DSSTox Substance Id.

Chemical Weight Fractions 1





Columns Y 10 V

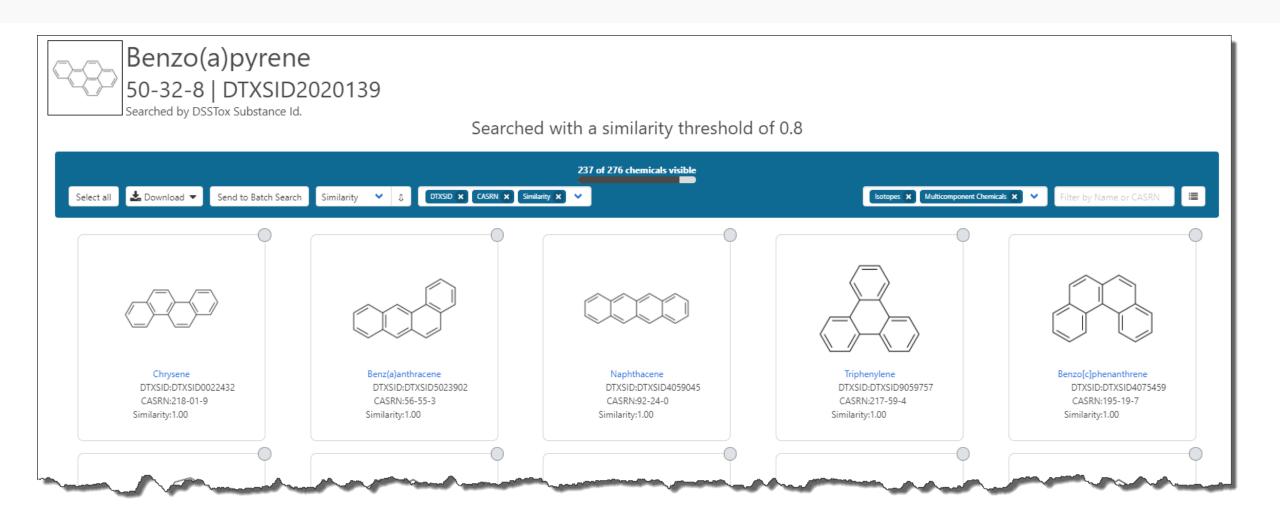


Product Name	Product Use Category	Minimum Weight Fraction	Maximum Weight Fraction	Data Type	Source
m-525-1-5x pah mixtures 0.5 mg/ml for method 525	Not Yet Categorized:			MSDS	SIRI
mm6125 surface conditioner	Not Yet Categorized:			Health Product Declaration	Health Product Declaration Collaborative
monolithic membrane 6125 (mm6125) / monolithic membrane	Not Yet Categorized:			Health Product Declaration	Health Product Declaration Collaborative
organic potablewatr pw 32_component h:reg semi-volatile 690	Not Yet Categorized:	0.00	1.00e-3	MSDS	SIRI
polynuclear aromatic hydrocarbon mixture_ep84627	Not Yet Categorized:			MSDS	SIRI
prestone(r) power steering fluid	engine maintenance: auto fluids and additives			MSDS	CPCPdb
r-12 shield tite wet surface coating	Not Yet Categorized:	0.00	0.500	MSDS	SIRI
sea tar 1010_ 0028	Not Yet Categorized:			MSDS	SIRI
supelpreme-hc kit pah mix_ 48909	Not Yet Categorized:			MSDS	SIRI
supelpreme-hc pah mix 1ml_ 48905	Not Yet Categorized:			MSDS	SIRI



## Similarity Searching





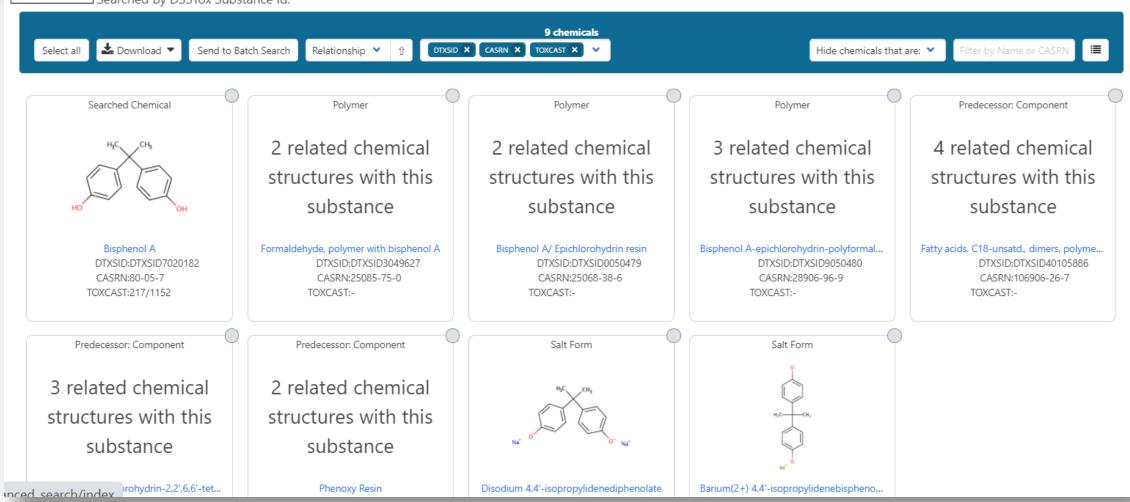
#### Related Substances





#### Bisphenol A 80-05-7 | DTXSID7020182

Searched by DSSTox Substance Id.

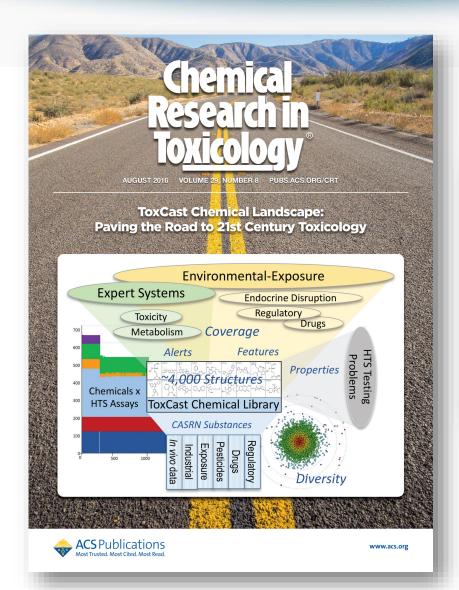




# Bioactivity Data

#### **ToxCast**





# ToxCast Chemical Landscape: Paving the Road to 21st Century Toxicology

Ann M. Richard<sup>\*†</sup>, Richard S. Judson<sup>†</sup>, Keith A. Houck<sup>†</sup>, Christopher M. Grulke<sup>†</sup>, Patra Volarath<sup>‡</sup>, Inthirany Thillainadarajah<sup>§</sup>, Chihae Yang<sup>||⊥</sup>, James Rathman<sup>⊥#</sup>, Matthew T. Martin<sup>†</sup>, John F. Wambaugh<sup>†</sup>, Thomas B. Knudsen<sup>†</sup>, Jayaram Kancherla<sup>▽</sup>, Kamel Mansouri<sup>▽</sup>, Grace Patlewicz<sup>†</sup>, Antony J. Williams<sup>†</sup>, Stephen B. Little<sup>†</sup>, Kevin M. Crofton<sup>†</sup>, and Russell S. Thomas<sup>†</sup>

#### View Author Information ✓

**○ Cite this:** Chem. Res. Toxicol. 2016, 29, 8, 1225–1251

Publication Date: July 1, 2016  $\scriptstyle \vee$ 

https://doi.org/10.1021/acs.chemrestox.6b00135

Article Views

6687

36

Altmetric

244

Citations





Add to Export

LEARN ABOUT THESE METRICS

## ToxCast Chemicals and Assays



#### **Select List**



List Acronym	List Name	Last Updated 🕏	Number of Chemicals	List Description
TOXCAST_PH1V2	TOXCAST_ph1v2 - EPA ToxCast Screening Library (ph1v2 Subset)	2016-01-25	293	TOXCAST_ph1v2 is the ph1v2 subset of TOXCAST, a reprocured subset of Phase I (ph1v1) chemicals moved into Phase II and later testing phases of the ToxCast program.
TOXCAST_PHASEI	TOXCAST_Phasei - EPA ToxCast Screening Library (Phase I subset)	2016-01-29	310	TOXCAST_Phasel corresponds to the ph1v1 subset of TOXCAST (mostly pesticides) screened in Phase I of the ToxCast program.
TOXCAST_PH2	TOXCAST_ph2 - EPA ToxCast Screening Library (ph2 Subset)	2016-01-25	768	TOXCAST_ph2 is the ph2 subset of TOXCAST, added in Phase II of the ToxCast program to increase chemical diversity and coverage of chemicals of concern to EPA programs.
TOXCAST_E1K	TOXCAST_e1k - EPA ToxCast Screening Library (e1k Subset)	2016-01-25	799	TOXCAST_e1k is the e1k subset of TOXCAST, selected for screening in endocrine-related assays.
TOXCAST_PHASEII	TOXCASST_PhaseII - EPA ToxCast Screening Library (Phase II Subset)	2016-01-29	1864	TOXCAST_PhaseII is the full set of chemicals screened in Phase II of the ToxCast program, consisting of TOXCAST_ph1v2, ph2 and e1k sublists.
TOXCAST_PH3	TOXCAST_ph3 - EPA ToxCast Screening Library (ph3 subset)	2018-04-11	2678	TOXCAST_ph3 is the ph3 subset of TOXCAST, added to the most recent Phase III of the ToxCast program to further increase chemical diversity and coverage of chemicals of concern to EPA programs.
TOXCAST_PHASEIII	TOXCAST_PhaseIII - EPA ToxCast Screening Library (Phase II Subset)	2017-04-11	4584	TOXCAST_PhaseIII is the full set of chemicals available for screening in Phase III of the ToxCast program, consisting of the majority of chemicals screened in Phase II and newly added ph3

# ToxCast covers a lot of biology but not all ToxCast is growing over time.

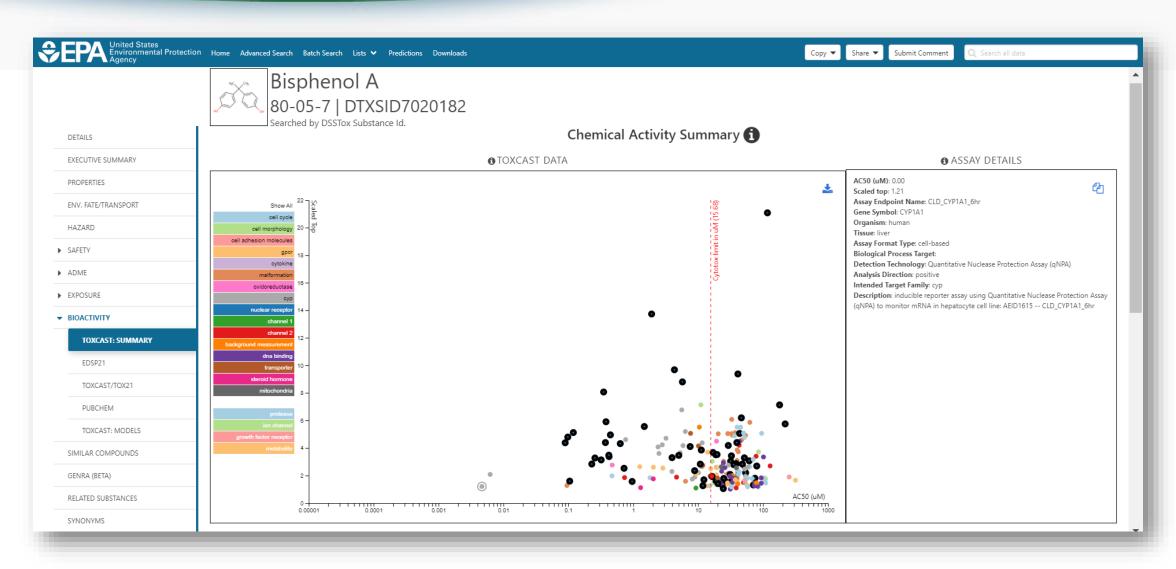


Invitrodb version 3.3 (released August 2020) contained 17 different assay sources, covering (at least) 491 unique gene-related targets with 1600 unique assay endpoints.

Assay source	Long name	Truncated assay source description	Some rough notes on the biology covered
ACEA	ACEA Biosciences	real-time, label-free, cell growth assay system based on a microelectronic impedance readout	Endocrine (ER-induced proliferation)
APR	Apredica	CellCiphr High Content Imaging system	Hepatic cells (HepG2)
ATG	Attagene	multiplexed pathway profiling platform	Nuclear receptor and stress response profile
BSK	Bioseek	BioMAP system providing uniquely informative biological activity profiles in complex human primary co-culture systems	Immune/inflammation responses
NVS	Novascreen	large diverse suite of cell-free binding and biochemical assays.	Receptor binding; transporter protein binding; ion channels; enzyme inhibition; many targets
ОТ	Odyssey Thera	novel protein:protein interaction assays using protein-fragment complementation technology	Endocrine (ER and AR)
TOX21	Tox21/NCGC	Tox21 is an interagency agreement between the NIH, NTP, FDA and EPA. NIH Chemical Genomics Center (NCGC) is the primary screening facility running ultra high-throughput screening assays across a large interagency-developed chemical library	Many – with many nuclear receptors
CEETOX	Ceetox/OpAns	HT-H295R assay	Endocrine (steroidogenesis)
CLD	CellzDirect	Formerly CellzDirect, this Contract Research Organization (CRO) is now part of the Invitrogen brand of Thermo Fisher providing cell-based in vitro assay screening services using primary hepatocytes.	Liver (Phase I/Phase II/ Phase III expression)
NHEERL_PADILLA	A NHEERL Padilla Lab	The Padilla laboratory at the EPA National Health and Environmental Effects Research Laboratory focuses on the development and screening of zebrafish assays.	Zebrafish terata
NCCT	NCCT Simmons Lab	The Simmons Lab at the EPA National Center for Computational Toxicology focuses on developing and implementing in vitro methods to identify potential environmental toxicants.	y Endocrine (thyroid - thyroperoxidase inhibition)
TANGUAY	Tanguay Lab	The Tanguay Lab, based at the Oregon State University Sinnhuber Aquatic Research Laboratory, uses zebrafish as a systems toxicology model.	Zebrafish terata/phenotypes
NHEERL_NIS	NHEERL Stoker & Laws	The Stoker and Laws laboratories at the EPA National Health and Environmental Effects Research Laboratory work on the development and implementation of high-throughput assays, particularly related to the sodium-iodide cotransporter (NIS).	Endocrine (thyroid - NIS inhibition)
UPITT	University of Pittsburgh	The Johnston Lab at the University of Pittsburgh ran androgen receptor nuclear translocation assays under a Material Transfer Agreement (MTA for the ToxCast Phase 1, Phase 2, and E1K chemicals.	Endocrine (AR related)

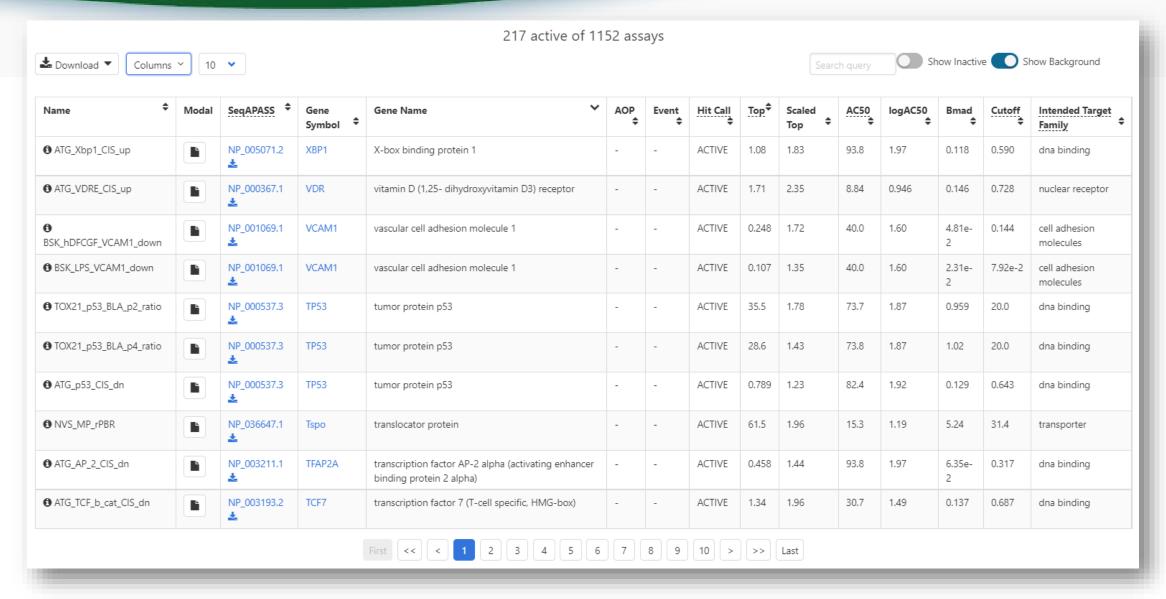
#### ToxCast/Tox21 Data





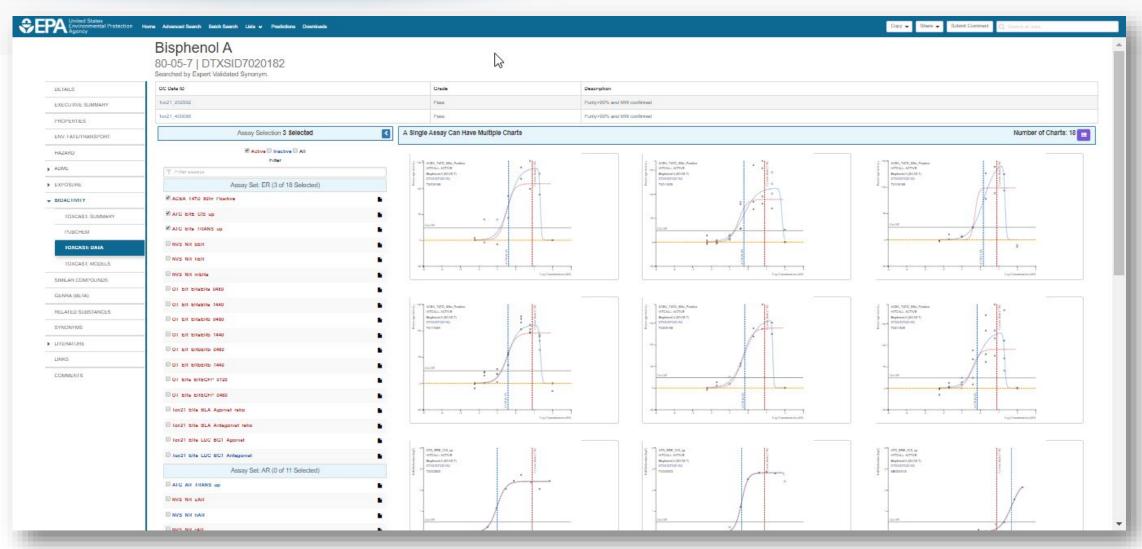
### Rich data tables – full transparency





# Bioactivity Data (ToxCast/Tox21) Data below for Bisphenol A





#### Use Models Derived from the Data



#### Screening Chemicals for Estrogen Receptor Bioactivity Using a Computational Model

Patience Browne\*†, Richard S. Judson‡, Warren M. Casey§, Nicole C. Kleinstreuer<sup>||</sup>, and Russell S. Thomas‡

View Author Information >

© Cite this: Environ. Sci. Technol. 2015, 49, 14, 8804–8814

Publication Date: June 12, 2015 V https://doi.org/10.1021/acs.est.5b02641 Article Views 3796

Altmetric

Citations 157

\_\_\_\_

I FADN AROUT THESE METDICS

<u>Vol. 124, No. 7</u> | Research

## CERAPP: Collaborative Estrogen Receptor Activity Prediction Project

Kamel Mansouri, Ahmed Abdelaziz, Aleksandra Rybacka, Alessandra Roncaglioni, Alexander Tropsha, Alexandre Varnek, Alexey Zakharov, Andrew Worth, Ann M. Richard, Christopher M. Grulke, Daniela Trisciuzzi, Denis Fourches, Dragos Horvath, Emilio Benfenati, Eugene Muratov, Eva Bay Wedebye, Francesca Grisoni, Giuseppe F. Mangiatordi, ... See all authors

Published: 1 July 2016 https://doi.org/10.1289/ehp.1510267 Cited by: 76

#### Development and Validation of a Computational Model for Androgen Receptor Activity

Nicole C. Kleinstreuer<sup>\*†</sup> [i], Patricia Ceger<sup>‡</sup>, Eric D. Watt<sup>§</sup> [i], Matthew Martin<sup>§</sup>, Keith Houck<sup>§</sup>, Patience Browne<sup>||</sup>, Russell S. Thomas<sup>§</sup>, Warren M. Casey<sup>†</sup>, David J. Dix<sup>⊥</sup>, David Allen<sup>‡</sup>, Srilatha Sakamuru<sup>#</sup>, Menghang Xia<sup>#</sup>, Ruili Huang<sup>#</sup>, and Richard Judson<sup>§</sup>

View Author Information ∨

© Cite this: Chem. Res. Toxicol. 2017, 30, 4, 946–964 Publication Date: November 18, 2016 × https://doi.org/10.1021/acs.chemrestov.6b00347

https://doi.org/10.1021/acs.chemrestox.6b00347
Copyright © 2016 American Chemical Society

Article Views 4338

Altmetric **77** 

LEARN AROUT THESE METRICS

Citations 94









Vol. 128, No. 2 | Research

## CoMPARA: Collaborative Modeling Project for Androgen Receptor Activity

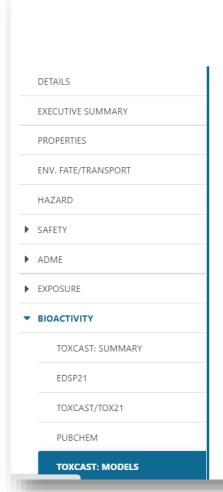
Kamel Mansouri 🖂, Nicole Kleinstreuer, Ahmed M. Abdelaziz, Domenico Alberga, Vinicius M. Alves, Patrik L. Andersson, Carolina H. Andrade, Fang Bai, Ilya Balabin, Davide Ballabio, Emilio Benfenati, Barun Bhhatarai, Scott Boyer, Jingwen Chen, Viviana Consonni, Sherif Farag, Denis Fourches, Alfonso T. García-Sosa, Paola Gramatica, Francesca Grisoni, ... See all authors

Published: 7 February 2020 | CID: 027002 | https://doi.org/10.1289/EHP5580 | Cited by: 2

# For Endocrine (AR and ER) better to use summary models

Download ToxCast Model Predictions







Positive ToxCast ER pathway agonist and ToxCast AR antagonist scores.

ToxCast: Models
ToxCast Model Predictions

Model	Receptor	Agonist	Antagonist	Binding
<b>1</b> ToxCast Pathway Model (AUC)	Androgen	0.00	0.345	-
<b>1</b> ToxCast Pathway Model (AUC)	Estrogen	0.450	0.00	-
COMPARA (Consensus)	Androgen	Inactive	Active	Active
CERAPP Potency Level (From Literature)	Estrogen	Active (Weak)	-	Active (Weak)
CERAPP Potency Level (Consensus)	Estrogen	Active (Weak)	Active (Strong)	Active (Weak)

CERAPP = consensus ER QSAR (from 17 groups)

COMPARA = consensus AR QSAR

ToxCast Pathway Model AUC ER = full ER model (18 assays)

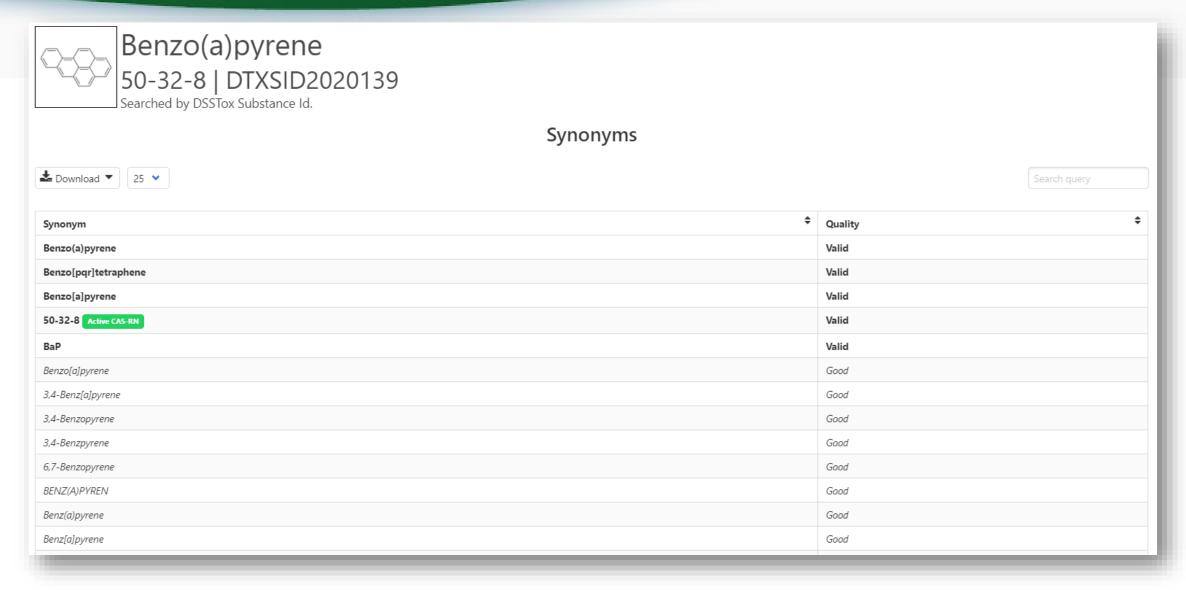
ToxCast Pathway Model AUC AR = full AR model (11 assays)



# Searching Literature and the Internet

# Identifiers Support Searches in other systems

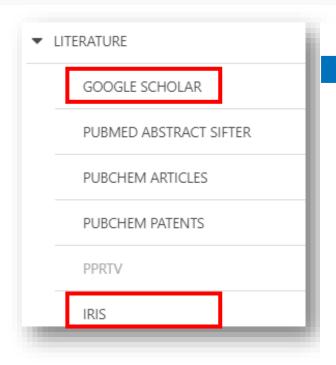


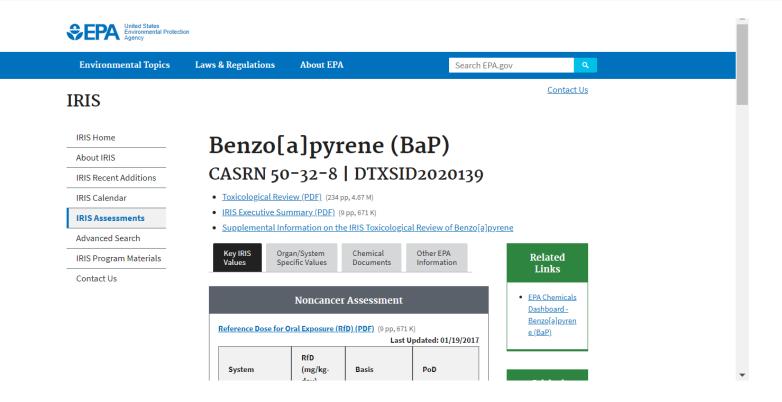


#### Identifiers are used in the app



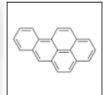
Identifiers are used to feed and link into "Literature"





#### Literature Searching

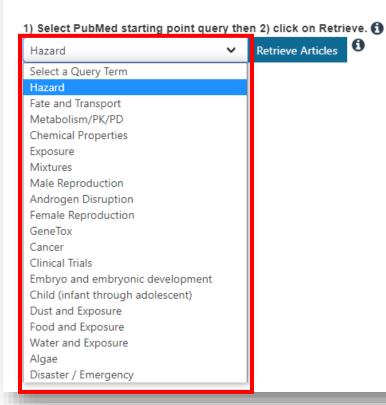




#### Benzo(a)pyrene 50-32-8 | DTXSID2020139

Searched by DSSTox Substance Id.

#### **Abstract Sifter**



Optionally, edit the query before retrieving.	
("50-32-8" OR "Benzo(a)pyrene") AND (NOAEL OR NOEL OR LOEL OR Rfd OR "reference dose" OR "reference concentration" OR "adverse effect level"[tiab] OR "cancer slope factor"[tiab])	

- Real-time retrieval of data from PubMed ~30 million abstracts and growing)
- Choose from set of pre-defined queries
- Adjust and fine tune queries based on interests

## Literature Searching



- "Sifting" of results using multiple terms
- Frequency counting terms
- Color highlighting of terms
- Download list to Excel
- Send list to PubMed for downloading ref. file
- Direct link via PubMed ID

der	lermal cancer pyrene Clear Terms			Download / Send to   Download Sifter for Excel							
	dermal	cancer ↓	pyrene	Total	PMID	Year	Title	Aut	hors	Journal	Rev
וכ	0	7	1	8	23922326	2013	Using immunotoxicity information to improve cance	r risk a	Zaccaria; McClure	International journal of toxicology	√
5	8	7	2	17	16632147	2006	Development of a dermal cancer slope factor for be	nzo[a]	Knafla; Phillipps; Brecher; Petrovic; Richardson	Regulatory toxicology and pharmacology : RTP	✓
וכ	4	6	2	12	33359623	2020	Testing the validity of a proposed dermal cancer slo	pe fac	Magee; Forsberg	Regulatory toxicology and pharmacology : RTP	√
וכ	0	5	1	6	28477805	2017	Pollution characteristics, sources and lung cancer r	isk of	Wang; Xia; Wu; Zhang; Sun; Yin; Zhou; Yang	Journal of environmental sciences (China)	
	4	4	2	10	20888881	2010	Development and application of a skin cancer slope	factor	Knafla; Petrovic; Richardson; Campbell; Rowat	Regulatory toxicology and pharmacology : RTP	
	4	4	1	9	16307791	2005	Health risk assessment on human exposed to envir	Health risk assessment on human exposed to environme		The Science of the total environment	
	2	4	1	7	11807932	2002	Cancer risk assessment for oral exposure to PAH mixtures.		Schneider; Roller; Kalberlah; Schuhmacher-Wolz	Journal of applied toxicology : JAT	
	2	3	1	6	32460055	2020	PAHs in Chinese atmosphere Part II: Health risk assessm		Ma; Zhu; Liu; Jia; Yang; Li	Ecotoxicology and environmental safety	
5	0	3	1	4	23379661	2013	Parent and halogenated polycyclic aromatic hydroc	Parent and halogenated polycyclic aromatic hydrocarbon		Journal of agricultural and food chemistry	
5	0	3	1	4	20800879	2010	Health risk assessment on dietary exposure to poly	cyclic	Xia; Duan; Qiu; Liu; Wang; Tao; Jiang; Lu; Song; H	Hu The Science of the total environment	
	2	3	1	6	16293284	2005	Probabilistic risk assessment for personal exposure	to car	Liao; Chiang	Chemosphere	
	0	2	1	3	17544483	2007	Health risk assessment for traffic policemen expose	ed to p	Hu; Bai; Zhang; Wang; Zhang; Yu; Zhu	The Science of the total environment	
	0	1	1	2	28795279	2017	Human health risk assessment and PAHs in a stret	Human health risk assessment and PAHs in a stretch of ri		Environmental monitoring and assessment	
	0	1	1	2	12634119	2003	Deviation from additivity in mixture toxicity: relevance of n		Lutz; Vamvakas; Kopp-Schneider; Schlatter; Stopp	per Environmental health perspectives	
	0	1	2	3	3709501	1986	The adsorption of polyaromatic hydrocarbons on na	atural a	Menard; Noel; Khorami; Jouve; Dunnigan	Environmental research	
7	0	0	1	1	33136306	2020	Effects on Apical Outcomes of Regulatory Relevan-	e of F	Crump: Boulanger: Farhat: Williams: Basu: Hecke	r: Environmental toxicology and chemistry	

#### Development of a dermal cancer slope factor for benzo[a] pyrene.

Polycyclic aromatic hydrocarbons (PAHs) are commonly found at environmentally impacted sites in both Canada and the United States, and also occur naturally. Typically, benzo[a] pyrene (B[a]P) is selected as a standard to which the cancer potencies of other carcinogenic PAHs are compared. Cancer potency estimates for B[a]P have been published for the oral and inhalation routes of exposure, however, no such estimate has been established by a regulatory agency for dermal exposure. The main objectives of the current investigation were to: evaluate approaches used to examine the relative carcinogenicity of PAHs; to conduct a review of mammalian dermal carcinogenicity studies for B[a]P and derive a cancer slope factor for dermal exposure to PAHs using B[a]P as a surrogate for other PAHs. The toxicological database of dermal B[a]P studies was examined for relevant animal bloassays. Seven relevant studies were identified. A cancer slope factor for B[a]P was developed using the benchmark dose approach and the linearized multistage model. The upper 95th C1 at the 5% effect level above background incidence was used as the point of departure for low-dose linear extrapolation. An average slope factor of 0.55 (microg/animal day)(-1) was calculated for mice, which was converted to a dose-equivalent slope factor of 25 (mg/kg day)(-1). This latter slope factor is proposed for application to human health risk assessment with no scaling adjustment. Dermal potency equivalency factor values were identified which may be used with other carcinogenic PAH in the calculation of total B[a]P equivalent dermal cancer risk estimates. An identified area for further investigation is the consideration of scaling in extrapolating the calculated dermal cancer slope factor from mice to humans.

# What's the best way to search the internet for chemical data?



- We know how complex chemicals identifiers are...
  - CASRN(s)
  - Hundreds of names (maybe)
  - SMILES
  - InChIs
  - EINECS, EC numbers
- What can WE do to help you navigate the internet?

# External Links – Also use Identifiers Names, CASRN, PubChem IDs, InChls.





#### Benzo(a)pyrene 50-32-8 | DTXSID2020139

Searched by DSSTox Substance Id.

#### General

- **EPA Substance Registry Service**
- PubChem
- Chemspider
- CPCat
- DrugBank
- W Wikipedia
- Q MSDS Lookup
- ChEMBL
- ToxPlanet
- ACS Reagent Chemicals
- ₩ Wolfram Alpha
- ECHA Infocard
- ChemAgora
- Consumer Product Information Database
- ChEBI
- NIST Chemistry Webbook
- **WEBWISER**
- PubChem Safety Sheet
- Consumer Product Information Database
- PubChem: Chemical Vendors

#### Toxicology

- ACToR
- он<sub>е</sub> DrugPortal
- CCRIS
- ChemView
- **©** CTD
- eChemPortal
- Gene-Tox
- HSDB
- ACToR PDF Report
- CREST
- National Air Toxics Assessment
- ECOTOX
- ChemView
- Chemical Checker
- ☑ BindingDB
- CalEPA OEHHA
- MIOSH IDLH Values
- LactMed
- ECOTOX

#### **Publications**

- Toxline
- PPRTVWFB
- PubMed
- IRIS Assessments
- EPA HERO
- NIOSH Skin Notation Profiles
- NIOSH Pocket Guide
- RSC Publications
- BioCaddie DataMed
- Springer Materials
- Bielefeld Academic Search Engine
- CORE Literature Search
- Google Books (Text Search)
- Google Patents (Text search)
- G Google Scholar (Text search)
- Google Patents (Structure search)
- Google Books (Structure Search)
- Google Scholar (Structure search)
- Federal Register

#### Analytical

- RSC Analytical Abstracts
- → Tox21 Analytical Data
- MONA: MassBank North America
- mzCloud
- NIST IR Spectrum
- NIST MS Spectrum
- MassBank
- NIST Antoine Constants
- IR Spectra on PubChem
- NIST Kovats Index values
- ☑ Protein DataBank
- National Environmental Methods Index

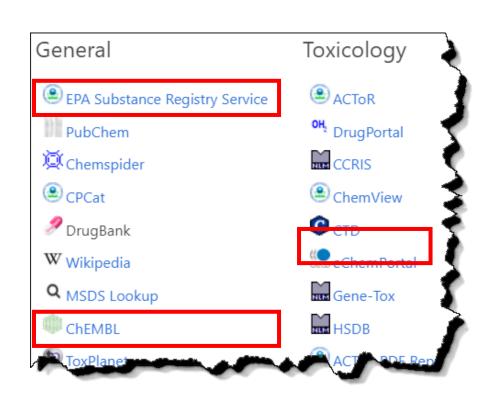
#### Prediction

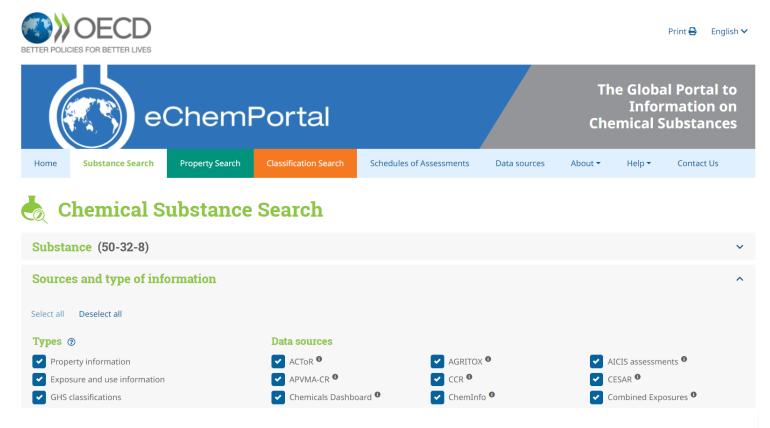
- 2D NMR HSQC/HMBC Prediction
- Carbon-13 NMR Prediction
- Proton NMR Prediction
- ChemRTP Predictor
- **€** LSFRD

#### **External Links**



 Links to ~90 websites providing access to additional data on the chemical of interest



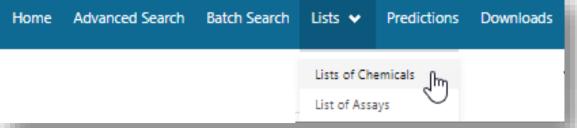


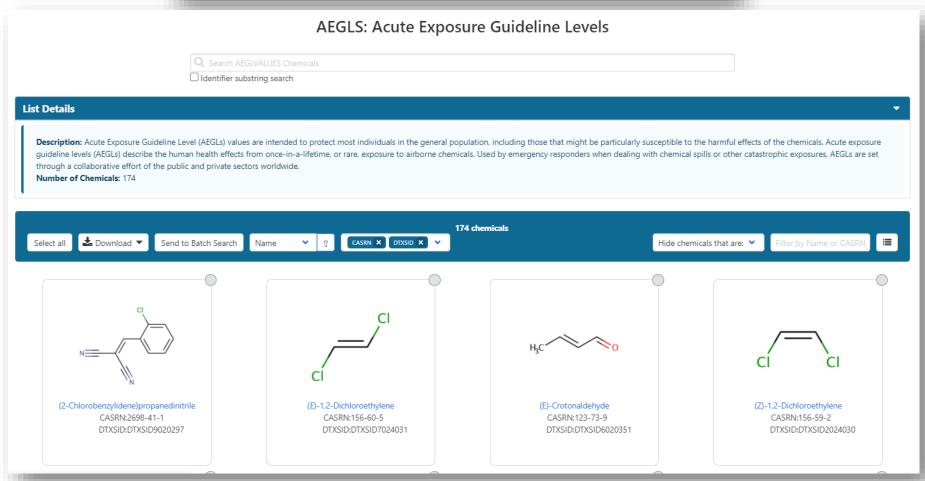


# Chemical Lists and Categories

## Example: AEGLs list







## PFAS lists of Chemicals



Copy Filtered Lists URL

#### **Select List**

PFAS



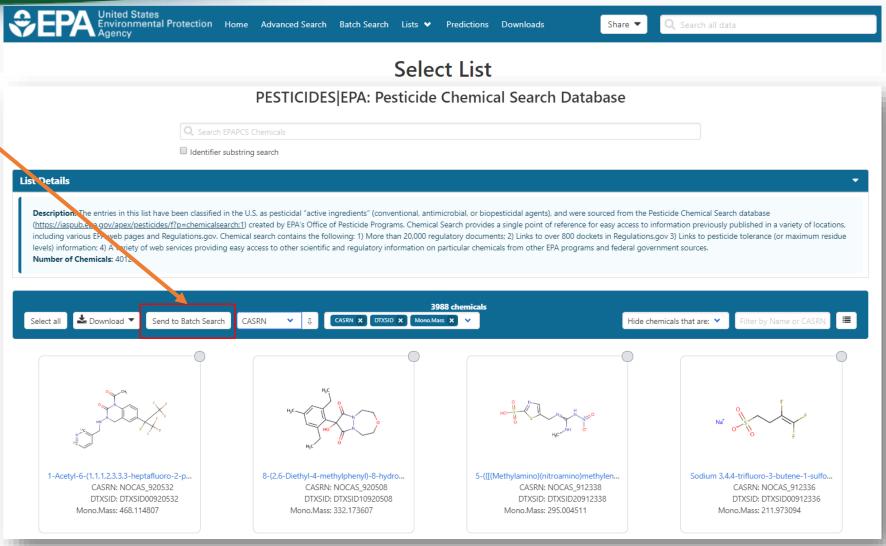
List Acronym	List Name \$	Last Updated 🕏	Number of Chemicals 🕏	List Description
EPAPFAS75S1	PFAS EPA: List of 75 Test Samples (Set 1)	2018-06-29	74	PFAS list corresponds to 75 samples (Set 1) submitted for initial testing screens conducted by EPA researchers in collaboration with researchers at the National Toxicology Program.
EPAPFAS75S2	PFAS EPA: List of 75 Test Samples (Set 2)	2019-02-21	75	PFAS list corresponds to a second set of 75 samples (Set 2) submitted for testing screens conducted by EPA researchers in collaboration with researchers at the National Toxicology Program.
EPAPFASCAT	PFAS EPA Structure- based Categories	2018-06-29	64	List of registered DSSTox "category substances" representing PFAS categories created using ChemAxon's Markush structure-based query representations.
EPAPFASINSOL	PFAS EPA: Chemical Inventory Insoluble in DMSO	2018-06-29	43	PFAS chemicals included in EPA's expanded ToxCast chemical inventory found to be insoluble in DMSO above 5mM.
EPAPFASINV	PFAS EPA: ToxCast Chemical Inventory	2018-06-29	430	PFAS chemicals included in EPA's expanded ToxCast chemical inventory and available for testing.
EPAPFASRL	PFAS EPA: Cross-Agency Research List	2017-11-16	199	EPAPFASRL is a manually curated listing of mainly straight-chain and branched PFAS (Per- & Poly-fluorinated alkyl substances) compiled from various internal, literature and public sources by EPA researchers and program office representatives.
PFASKEMI	PFAS: List from the Swedish Chemicals Agency (KEMI) Report	2017-02-09	2416	Perfluorinated substances from a Swedish Chemicals Agency (KEMI) Report on the occurrence and use of highly fluorinated substances.
PFASMASTER	PFAS Master List of PFAS Substances	2018-07-26	5061	PFASMASTER is a consolidated list of PFAS substances spanning and bounded by the below lists of current interest to researchers and regulators worldwide.
PFASOECD	PFAS: Listed in OECD Global Database	2018-05-16	4729	OECD released a New Comprehensive Global Database of Per- and Polyfluoroalkyl Substances, (PFASs) listing more than 4700 new PFAS
PFASTRIER	PFAS Community- Compiled List (Trier et al., 2015)	2017-07-16	597	PFASTRIER community-compiled public listing of PFAS (Trier et al, 2015)

#### **Curated List of Pesticides**



Find list of interest

 Select list and send to batch





# Batch Searching

#### **Batch Searching**

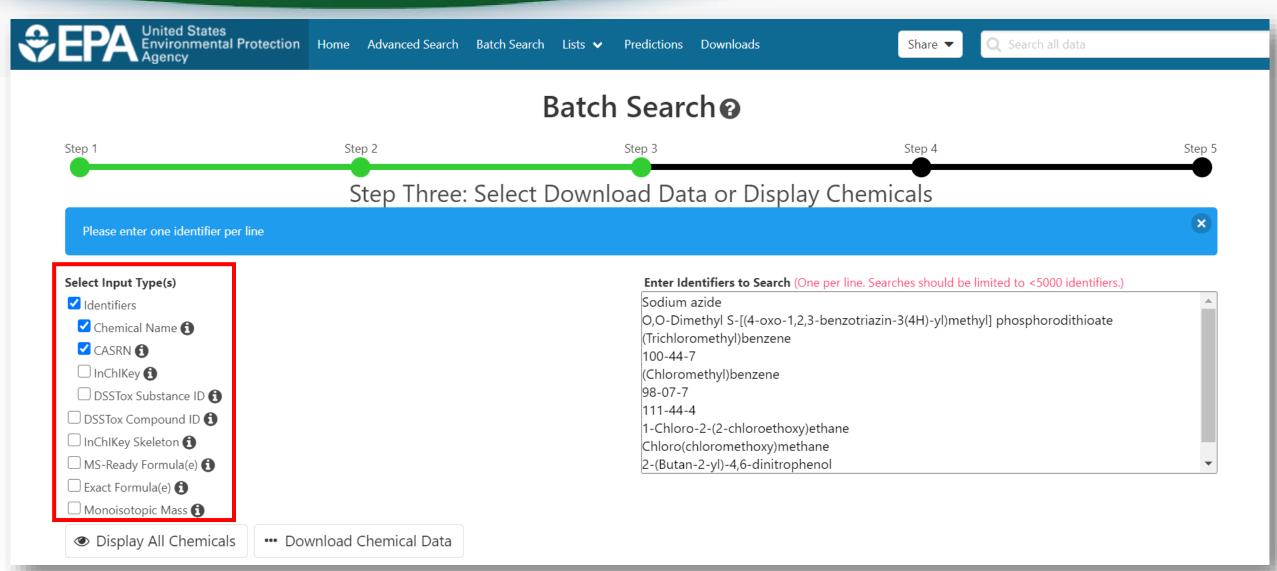


- Singleton searches are great but...
- ...we generally want data on LOTS of chemicals!

- Typical questions
  - What are the structures for a set of chemical names? Set of CASRNs?
  - Can I get chemical lists in Excel files? As a list of SMILES strings?
     Can I get an SDF file?
  - Can I include predicted properties? OPERA? TEST?
  - Are "these chemicals" screened in Toxcast?
  - I need masses and formulae for a list of chemicals

#### Access data en masse for thousands of chemicals...





#### Select Output Format and Content



#### Step Four: Select Data Output Format and Choose Data Fields to Download

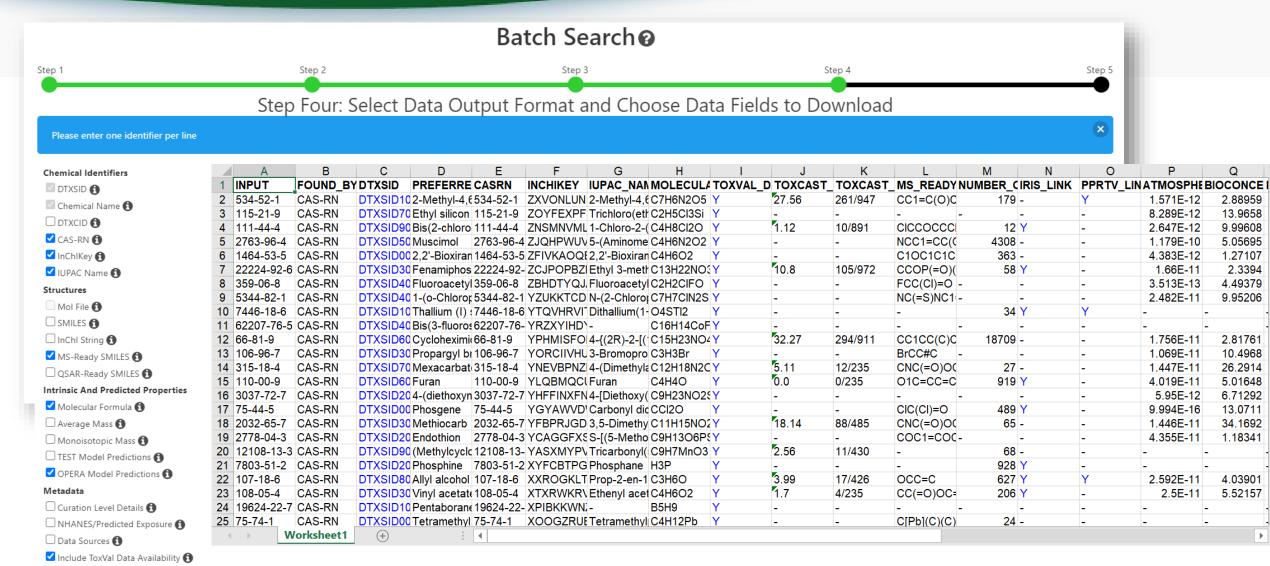
■ Excel		▲ Download
Customize Results	Intrinsic And Predicted Properties	Presence in Lists:
Select All	☐ Molecular Formula <b>1</b> ☐ Average Mass <b>1</b>	40CFR116.4 Designation of Hazardous Substances (Above Ground Storage Tanks)
Chemical Identifiers  DTXSID ① Chemical Name ①	☐ Monoisotopic Mass <b>1</b> ☐ TEST Model Predictions <b>1</b> ☐ OPERA Model Predictions <b>1</b>	<ul> <li>40CFR355 Extremely Hazardous Substance List and Threshold Planning Quantities</li> <li>AEGLS: Acute Exposure Guideline Levels</li> <li>△ ANDROGEN: Androgen Receptor Chemicals</li> <li>△ ARTICLE; Bench-Mark Dose Human Health Assessment List (Wignall et al., 2014)</li> </ul>
DTXCID (1) CAS-RN (1) InChIKey (1) IUPAC Name (1) Structures Mol File (1) SMILES (1) InChI String (1) MS-Ready SMILES (1) QSAR-Ready SMILES (1)	Metadata  Curation Level Details (1)  NHANES/Predicted Exposure (1)  Data Sources (1)  Include ToxVal Data Availability (1)  Assay Hit Count (1)  Number of PubMed Articles (1)  PubChem Data Sources (1)  CPDat Product Occurrence Count (1)  IRIS (1)  PPRTV (1)  Wikipedia Article  QC Notes (1)  Include links to ACTOR reports - SLOW! (BETA) (1)	□ ARTICLE: Collaborative Estrogen Receptor Activity Prediction Project (CERAPP)   □ ARTICLE: Collaborative Estrogen Receptor Activity Prediction Project (COMPARA)   □ ATSDR Toxicological Profiles   □ ATSDR: Minimal Risk Levels (MRLs) for Hazardous Substances   □ ATSDR: Toxic Substances Portal Chemical List   □ California Office of Environmental Health Hazard Assessment   □ Canadian Domestic Substances List 2019   □ CATEGORY: Amino acids   □ CATEGORY: Color Index dyes   □ CATEGORY: Flame Retardants

#### Batch Search CASRNs

✓ Assay Hit Count <a>6</a>

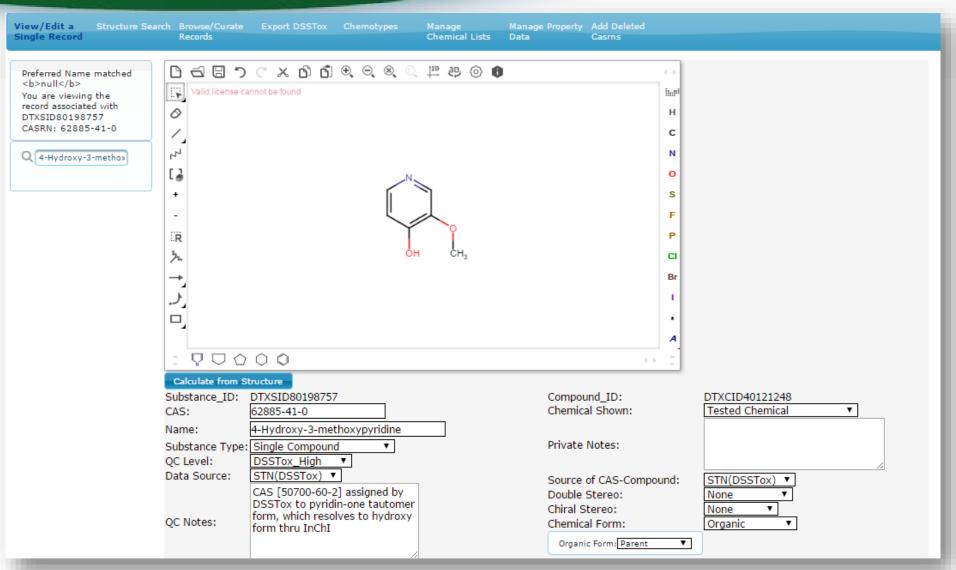
✓ Number of PubMed Articles ♠





#### Underneath the Dashboard





#### A little more about our data quality



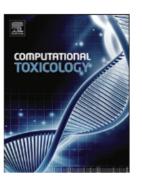
Computational Toxicology 12 (2019) 100096



Contents lists available at ScienceDirect

#### Computational Toxicology





EPA's DSSTox database: History of development of a curated chemistry resource supporting computational toxicology research



Christopher M. Grulke<sup>a</sup>, Antony J. Williams<sup>a</sup>, Inthirany Thillanadarajah<sup>b</sup>, Ann M. Richard<sup>a,\*</sup>

<sup>&</sup>lt;sup>a</sup> National Center for Computational Toxicology, Office of Research & Development, US Environmental Protection Agency, Mail Drop D143-02, Research Triangle Park, NC 27711, USA

<sup>&</sup>lt;sup>b</sup> Senior Environmental Employment Program, US Environmental Protection Agency, Research Triangle Park, NC 27711, USA



# Generalized Read-Across

#### GenRA (Generalised Read-Across)

Predicting toxicity as a similarity-weighted

chemistry and bioactivity descriptors (Shah

performance baseline for read-across and

quantify the uncertainty in the predictions

activity of nearest neighbours based on

·Goal: To establish an objective

et al, 2016)

made

- United States
  Environmental Protection
  Agency
- Regulatory Toxicology and Pharmacology 79 (2016) 12-24

Contents lists available at ScienceDirect

## ELSEVIER

#### Regulatory Toxicology and Pharmacology

journal homepage: www.elsevier.com/locate/yrtph



Systematically evaluating read-across prediction and performance using a local validity approach characterized by chemical structure and bioactivity information



Imran Shah <sup>a, \*</sup>, Jie Liu <sup>b, c</sup>, Richard S. Judson <sup>a</sup>, Russell S. Thomas <sup>a</sup>, Grace Patlewicz <sup>a</sup>

- <sup>a</sup> National Center for Computational Toxicology, Office of Research and Development, U.S. Environmental Protection Agency, Research Triangle Park, NC 27211 USA
- b Department of Information Science, University of Arkansas at Little Rock, AR 72204, USA
- Ouk Ridge Institute for Science Education Fellow, National Center for Computational Toxicology, Office of Research and Development, U.S. Environmen Protection Agency, Research Triangle Park, NC 27711, USA

#### ARTICLE INFO

Article history: Received 25 September 2015 Received in revised form 20 April 2016

Accepted 3 May 2016 Available online 9 May 2016

Keywords: Read-across Nearest neighbors Local validity domains ABSTRACT

Read-scross is a popular data gap filling technique within category and analogue approaches for regulatory purposes. Acceptance of read-across remains an ongoing challenge with several efforts underwoy for identifying and addressing uncertainties. Here we demonstrate an algorithmic, automated approach to evaluate the utility of using in virio bioactivity data ("bioactivity descriptors," from EPNs Toxacats program) in conjunction with chemical descriptor information to derive local validity domains (specific sets of nearest neighbors) to facilitate read-across for up to ten in vive repeated done toxicity study types Over 2299 different chemical structure descriptors were generated for a set of 1778 chemicals and supplemented with the outcomes from 821 in virous assays. The read-across prediction of toxicity for 600 chemicals with in vivo data was based on the similarity weighted endpoint outcomes of its nearest neighbors. The approach enabled a performance baseline for read-across predictions of specific study outcomes to be established. Bioactivity descriptors were often found to be more predictive of in vivo toxicity outcomes than chemical descriptors or combination of both. This generator read-across receing level bazard assessment for new untested chemicals.

© 2016 Published by Elsevier In

$$y_i^{\beta,\alpha} = \frac{\sum_{j=1}^k S_{ij}^{\alpha} x_j^{\beta}}{\sum_{j=1}^k S_{ij}^{\alpha}}$$

Jaccard similarity:

$$s_{ij} = \frac{\sum_{l} (x_{il} \wedge x_{jl})}{\sum_{l} (x_{il} \vee x_{jl})}$$

 $a \square \{chm, bio, bc\}$ 

 $\beta \square \{bio, tox\}$ 

 $y_i$ = predicted activity of chemical( $c_i$ )

 $x_i^{\beta}$  = activity of  $c_i$  in  $\beta$ 

 $s_{ij}^{\alpha}$  = Jacccard similarity between  $x_i^{\alpha}$ ,  $x_j^{\alpha}$ 

k = up to k nearest neighbours

#### Read-across workflow in GenRA v1.0



#### **Decision Context**

Screening level assessment of hazard based on toxicity effects from ToxRefDB v1



# Analogue identification

Similarity context is based on structural characteristics



Data gap analysis for target and source analogues



# **Uncertainty** assessment

Assess prediction and uncertainty using AUC and p value metrics



#### Read-across

Similarity weighted average - many to one read-across



#### Analogue evaluation

Evaluate consistency and concordance of experimental data of source analogues across and between endpoints

# GenRA - Current research



- Consideration of other information to define and refine the analogue selection & evaluation
  - physicochemical similarity (Helman et al 2018)
  - metabolic similarity (in prep),
  - reactivity similarity (Nelms et al 2018)
  - transcriptomics similarity (Tate et al, 2021 in press)
- Transitioning to quantitative predictions of toxicity
  - Using GenRA to predict Lowest Observed Adverse Effect Level (LOAEL), acute oral (median lethal dose) LD50 (Helman et al 2019a,b)
- Developing a compendium of expert driven read-across examples to investigate how data driven read-across with NAM data can mirror expert assessments (in prep)

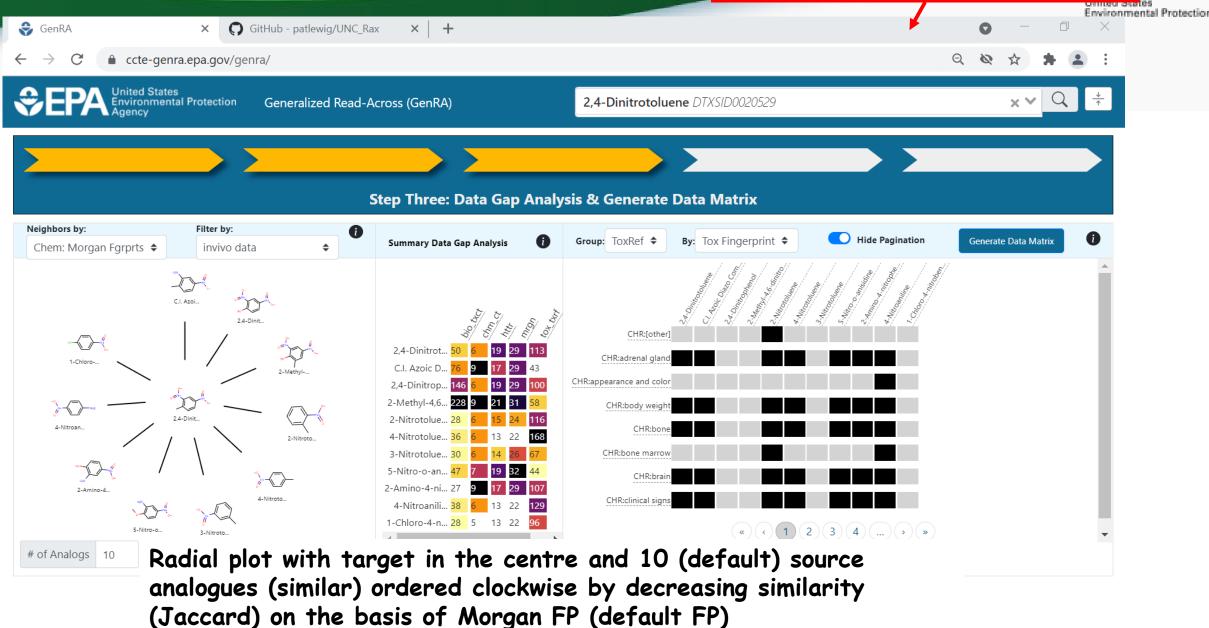
## GenRA Summary



- GenRA is an attempt to move towards an objective read-across approach where uncertainties and performance can be quantified. Provides opportunities for NAM data to be systemically incorporated.
- GenRA v1.0 established a baseline in performance. The approach relies on chemical descriptors to predict binary toxicity values but work continues to characterise other contexts of similarity (e.g. mechanistic, reactivity, metabolism) and quantify their contribution in predicting in vivo toxicity outcomes.
- GenRA v2.0 is anticipated to be released in the coming weeks as a standalone web application linked to the Dashboard. A python package (genra-py) was released (March 2021) to facilitate batch processing using user specific datasets see <a href="https://github.com/i-shah/genra-py">https://github.com/i-shah/genra-py</a>. The ability to try out the package using a sample dataset is available here https://github.com/patlewig/UNC\_Rax

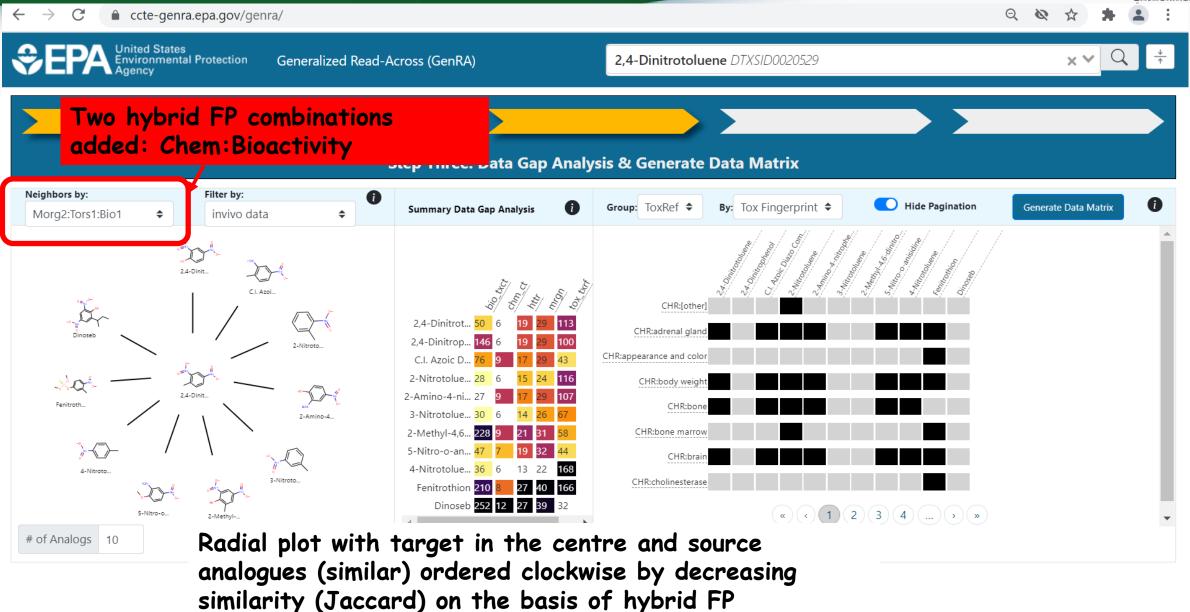
# GenRA v2 in practice

Search for a chemical of interest (target) using the search box

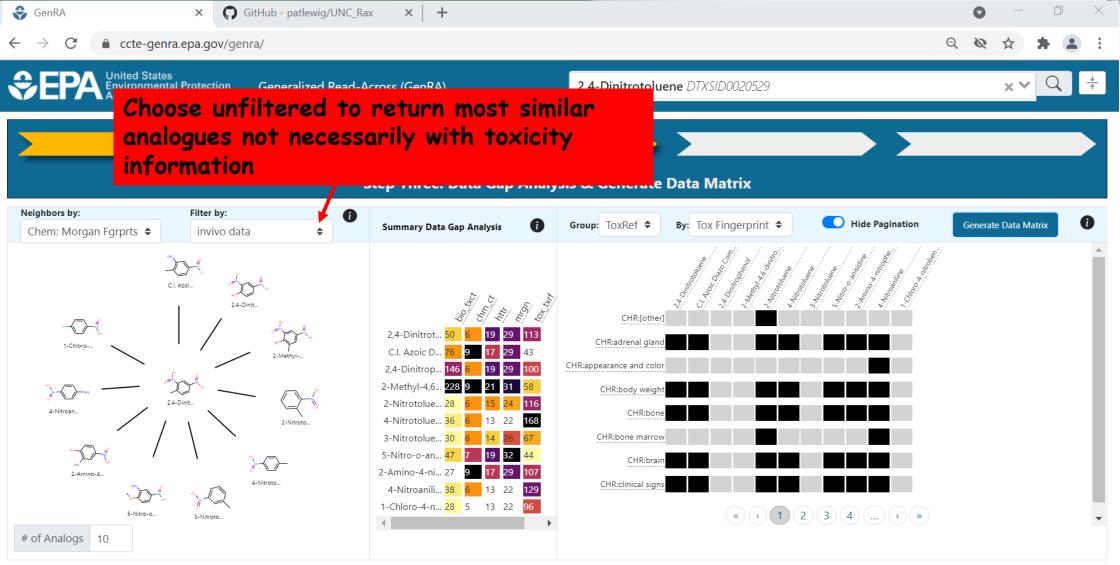


# GenRA v2 in practice

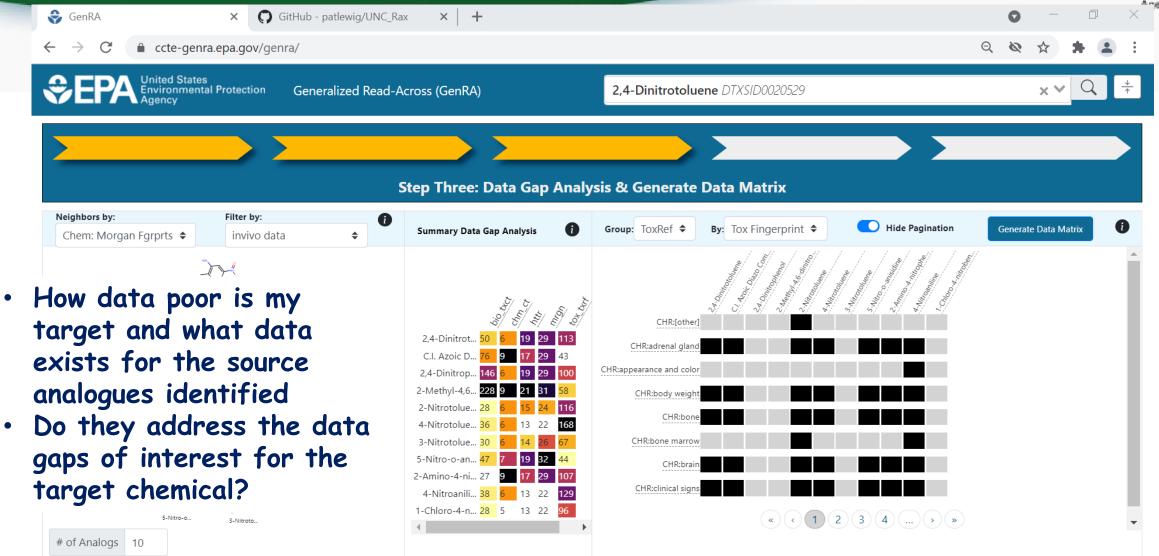








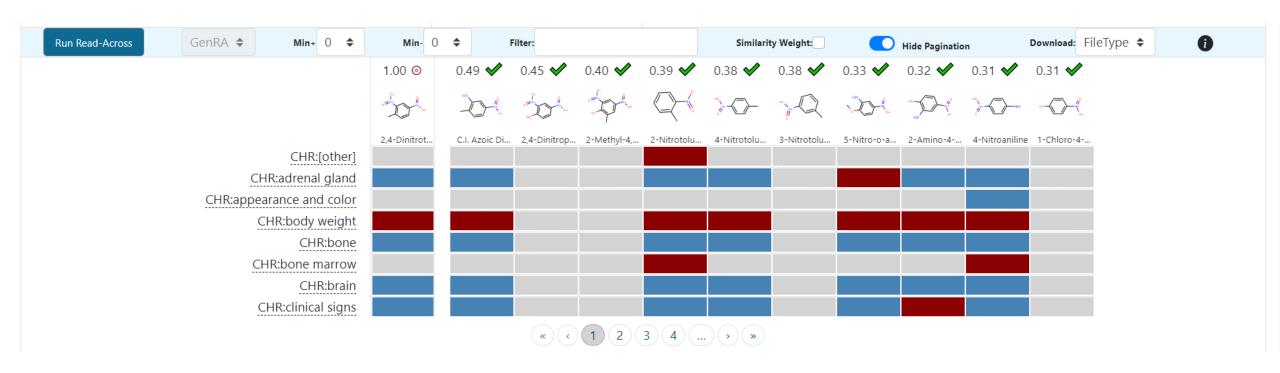






What is the consistency and concordance across my source analogues? Should I deselect analogues from consideration from the entire set of predictions? Should I consider subcategorising the analogues selected?

Toxicity data represented as binary outcomes - red (positive), blue (negative), grey (no data)









- Database underpinning GenRA v2.0: ToxRefDB v2
  - Different study types and effects within them are predicted e.g. chronic\_liver is annotated as CHR\_liver
  - Negative effects are inferred from guideline profiles which define the required tests for each study type. The assumption is that the study required an evaluation but no effects were reported.
  - Positive results min dose at which toxicity effects are observed in a study
- Prediction: Similarity weighted activity
- Performance is categorised by the Area under the Curve (AUC) of the Receiving Operating Characteristic (ROC)
  - The significance was empirically estimated by constructing a null distribution by permuting the toxicity values 100 times and calculating the fraction of times the AUC was more extreme than what would be observed by chance (this is reported as the p-value).



- Ability to export the predictions as an excel file
- Output can be analysed in different ways

	A	В	С	D	
1	role	target	analog	analog	
2	preferred name	2,4-Dinitrotoluene	C.I. Azoic Diazo Component 12	2,4-Dinitrophenol	2-
3	dsstox_sid	DTXSID0020529	DTXSID4020959	DTXSID0020523	
4	molecular weight	182.135	152.153	184.107	
5	similarity	1.00001	0.487179487	0.45	
6	CHR:[other]	GenRA Pos Act=1 (1) AUC=0 p=1	no_data	no_data	
7	CHR:adrenal gland	GenRA TN Act=0 (0.147) AUC=0 p=0.845	no_effect	no_data	
8	CHR:appearance and color	GenRA Neg Act=0 (0) AUC=0 p=1	no_data	no_data	
9	CHR:body weight	GenRA TP Act=1 (1) AUC=0 p=1	2.5mg/kg/day	no_data	
10	CHR:bone	GenRA TN Act=0 (0) AUC=0 p=1	no_effect	no_data	
11	CHR:bone marrow	GenRA Pos Act=1 (1) AUC=0 p=1	no_data	no_data	
12	CHR:brain	GenRA TN Act=0 (0) AUC=0 p=1	no_effect	no_data	
13	CHR:clinical signs	GenRA TN Act=0 (0.144) AUC=0 p=0.785	no_effect	no_data	
14	CHR:clitoral gland	GenRA TN Act=0 (0.326) AUC=0 p=0.96	no_effect	no_data	
15	CHR:ear	GenRA Pos Act=1 (0.562) AUC=0 p=0.51	no_data	no_data	
16	CHR:epididymis	GenRA TN Act=0 (0.178) AUC=0 p=0.865	no_effect	no_data	
17	CHR:erythrocyte (rbc) count differential	GenRA Pos Act=1 (1) AUC=0 p=1	no data	no data	



- Rank order positive results based on AUC and p values
- Look at the distribution of positive vs negatives predictions
- Explore what effects are being identified for the source analogues - consider identifying the underlying data for source analogues (elsewhere on the Dashboard) - is there a critical effect that is driving the toxicity that should be compared with the target chemical predictions?

•

 Depends on the decision context and the level of uncertainty that can be tolerated.



- If for example, body weight effects were important based on the source analogues, we could look at the prediction in GenRA for our target chemical
- GenRA prediction is positive for body weight effects in a chronic CHR (reflected by the True Positive (TP)) as summarised in the Download report
- What if we wanted to predict a point of departure from the source analogues...?
- We can also compute a similarity weighted average (recall the formula in an earlier slide..).



ole	e target		analog		analog		analog	og	Environmental Pro Agency		
referred name	erred name 2,4-Dinitrotoluene		ie	C.I. Azoic Diazo Component 12		2,4-Dinitrophenol 2-I		2-Methyl-4,6-dinitrophenol			
sstox_sid				DTXSID4020959		DTXSID0020523		DTXSID1022053			
nolecular weight         182.135           imilarity         1           HR:[other]         GenRA Pos Act=1 (1) AUC=0 p=1					0.45 0.395 no_data no_		198.134				
							0.395348837				
							no_data				
HR:adrenal gland HR:body weight			GenRA TN Act=0 (0.171) AUC=0 p=0.865 no_e		effect no_data			no_data			
HR:bone	1 A		В		С	D		E	F	G	Н
:HR:bone marr :HR:brain 1	role	preferred nar	ne	dssto	x_sid	molecular weig	ght s	similarity	CHR:body weight		-log10(toxicity/MW)
:HR:clinical sig :HR:clitoral gla									GenRA TP Act=1 (1)	GenRA TP Act=1	
HR:ear 2	target	2,4-Din	itrotoluene	DTX	SID0020529	182.135		1.00001	AUC=0 p=1	(1) AUC=0 p=1	
:HR:esophagu 3	analog	C.I. Azoic Did	zo Component 12	DTX	SID4020959	0959 152.153		0.4871795	1795 2.5mg/kg/day	2.5	1.784340511
4	4 analog 2,4-Dinitrophenol D		DTX	SID0020523	184.107	,	0.45	no_data	no_data		
5 analog 2-Methyl-4,6-dinitrophenol		DTX	SID1022053	198.134		0.3953488	no_data	no_data			
6	analog	2-Nit	rotoluene	DTX	SID4025791	137.138		0.3947368	50mg/kg/day	50	0.438187807
7	analog	g 4-Nitrotoluene		DTX	SID5023792	137.138		0.3783784	60mg/kg/day	60	0.359006561
8	analog	3-Nit	rotoluene	DTX	SID5021831	137.138		0.375	no_data	no_data	
9	analog	5-Nitro	-o-anisidine	DTX	SID0020943	168.152		0.326087	200mg/kg/day	200	-0.075327958
10	1	2 4 .	A	NTV	CTN/0200/2	15/105	l	0.3101010	10E /I. / I	125	0.0000/3077
				S	imilarity w	reighted a	ctivity	/ -log10	) molar equivalent	toxicity in	mg/kg/day
sum(pairwise similarity * toxicity) sum(pairwise similarities)			y)		1.5474	1582	7	0.69947419		36.38473532	
					2.212	2557	9				

Target is predicted to have a NOAEL of 36 mg/kg/day in a CHR study on the basis of body weight effects

# GenRA - Overall goal



- Quantify the contribution that different similarity contexts play in toxicity prediction and how that differs depending on the toxicity endpoint of interest, the chemical of interest and whether it mirrors expert driven read-across
- Quantify level of confidence for prediction made
- => objective, reproducible read-across assessments

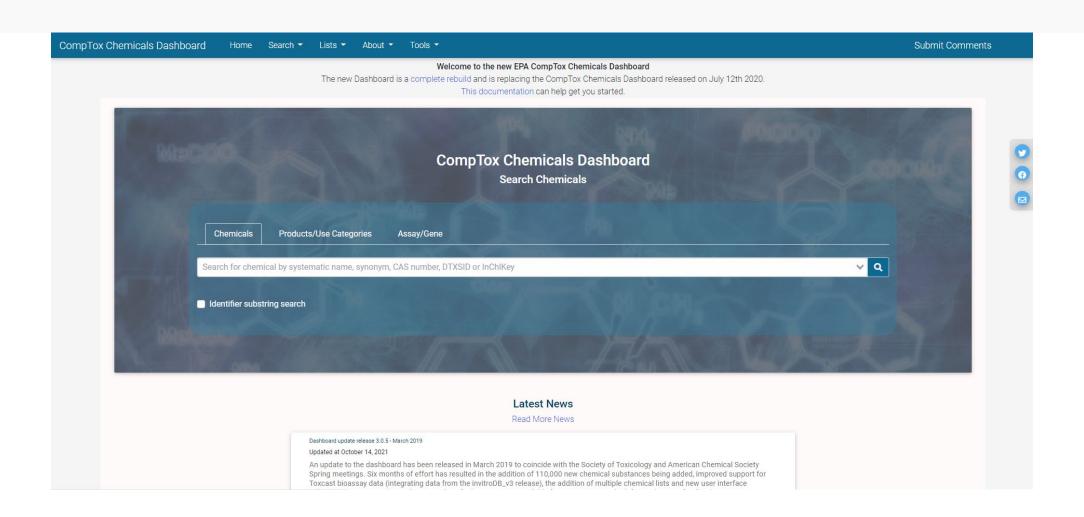
## GenRA v3...new features...



- Introduce structures not already registered in the Dashboard
- Custom fingerprints

# Coming Soon....NEW DASHBOARD https://ccte-ccd.epa.gov/dashboard/





#### Proof-of-Concept Apps

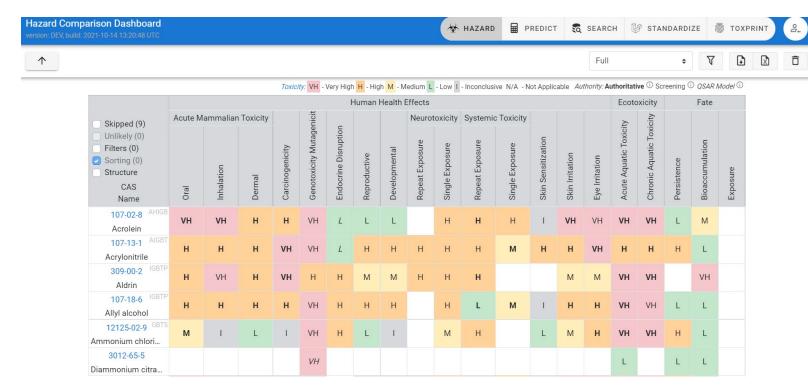


Structure-substructure-similarity searching of Dashboard

Batch prediction of physicochemical properties and toxicity

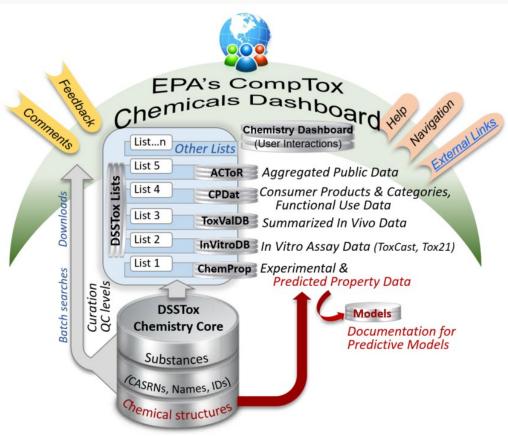
• "Hazard Comparison Dashboard" for comparing hazard profiles

for chemicals

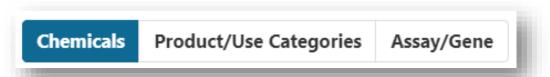


#### **Summary and Conclusion**





- CompTox Chemicals Dashboard a central hub for environmental data
  - ~875k chemical substances
  - Integrating property data, hazard data, exposure data, in vitro bioactivity data
  - Interrogation of bioactivity data -
  - Multiple types of searches

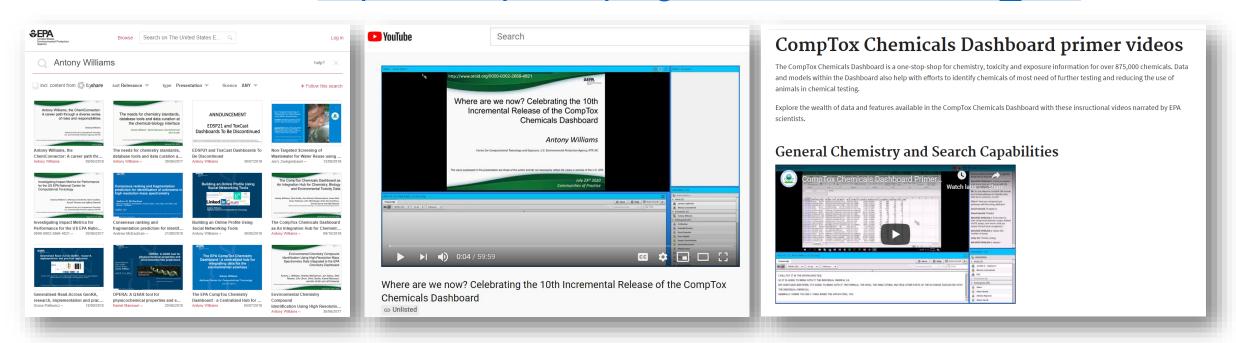


- Batch search for thousands of chemicals
- Real-time property and toxicity predictions
- Downloadable files CSV, TSV and Excel

#### You want to know more...



- Lots of resources available
  - Presentations: <a href="https://tinyurl.com/w5hqs55">https://tinyurl.com/w5hqs55</a>
  - Communities of Practice Videos: <a href="https://rb.gy/qsbno1">https://rb.gy/qsbno1</a>
  - Manual: <a href="https://rb.gy/4fgydc">https://rb.gy/4fgydc</a>
  - Latest News: <a href="https://comptox.epa.gov/dashboard/news">https://comptox.epa.gov/dashboard/news</a> info



#### Acknowledgments



#### Contacts:

Williams.Antony@epa.gov
Patlewicz.grace@epa.gov

- Feedback and follow-up is welcomed! Your questions help
- The dashboard is based on the efforts of many more team members than us. Many collaborators provide data also.



EPA's Center for Computational Toxicology and Exposure