

Leveraging chemistry data to improve exposure analyses using the EPA's CompTox Dashboard

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AGRO 29

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Exposure Science in the 21st Century: What are Key Drivers?

Understanding causes of disease

"...70-90% of disease risks are probably due to differences in environments"

EPIDEMIOLOGY

Environment and Disease Risks

Stephen M. Rappaport and Martyn T. Smith

lthough the risks of developing chronic diseases are attributed to both genetic and environmental factors, 70 to 90% of disease risks are probably due to differences in environments (1-3). Yet, epidemiologists increasingly use genomewide association studies (GWAS) to investigate diseases, while relying on questionnaires to characterize "environmental exposures." This is because GWAS represent the only approach for exploring the totality of any risk factor (genes, in this case) associated with disease prevalence. Moreover, the value of costly genetic information is diminished when inaccurate and imprecise environmental data lead to biased inferences regarding gene-environment interactions (4). A more comprehensive and quantitative view of environmental expo-

School of Public Health, University of California, Berkeley, CA 94720-7356, USA. E-mail: srappaport@berkeley.edu sure is needed if epidemiologists are to discover the major causes of chronic diseases.

An obstacle to identifying the most important environmental exposures is the fragmentation of epidemiological research along lines defined by different factors. When epidemiologists investigate environmental risks, they tend to concentrate on a particular category of exposures involving air and water pollution, occupation, diet and obesity, stress and behavior, or types of infection. This slicing of the disease pie along parochial lines leads to scientific separation and confuses the definition of "environmental exposures." In fact, all of these exposure categories can contribute to chronic diseases and should be investigated collectively rather than separately.

To develop a more cohesive view of environmental exposure, it is important to recognize that toxic effects are mediated through

A new paradigm is needed to assess how a lifetime of exposure to environmental factors affects the risk of developing chronic diseases.

chemicals that alter critical molecules, cells, and physiological processes inside the body. Thus, it would be reasonable to consider the "environment" as the body's internal chemical environment and "exposures" as the amounts of biologically active chemicals in this internal environment. Under this view, exposures are not restricted to chemicals (toxicants) entering the body from air, water, or food, for example, but also include chemicals produced by inflammation, oxidative stress, lipid peroxidation, infections, gut flora, and other natural processes (5, 6) (see the figure). This internal chemical environment continually fluctuates during life due to changes in external and internal sources, aging, infections, life-style, stress, psychosocial factors, and preexisting diseases.

The term "exposome" refers to the totality of environmental exposures from conception onwards, and has been proposed to be a

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2) Ensuring chemical safety and human/eco health





Comparing Analysis Approaches

Targeted Analysis:

- We know exactly what we're looking for
- 10s 100s of chemicals
- <<1% of the exposome

Suspect Screening Analysis (SSA):

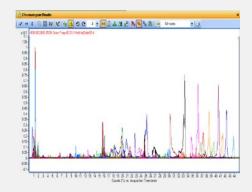
- We have chemicals of interest
- 100s 1,000s of chemicals
- ~5-10% of the exposome

Non-Targeted Analysis (NTA):

- We have no preconceived lists
- 1,000s 10,000s of chemicals
- 90-95% of the exposome
- In dust, soil, food, air, water, products- potential exposure source for plants, animals, and humans

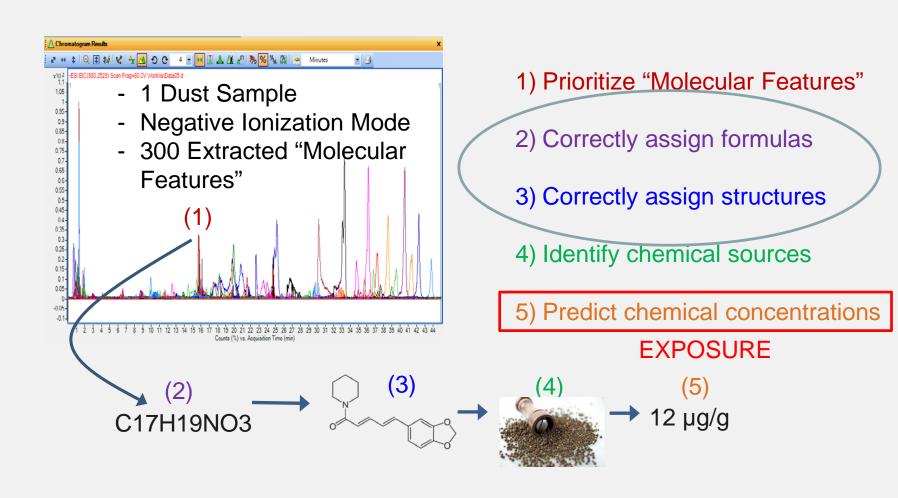






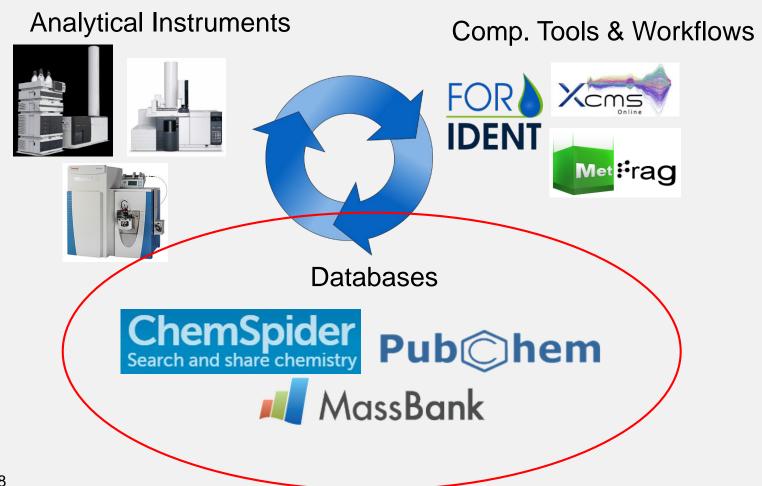


General Goals of SSA/NTA





The General Approach



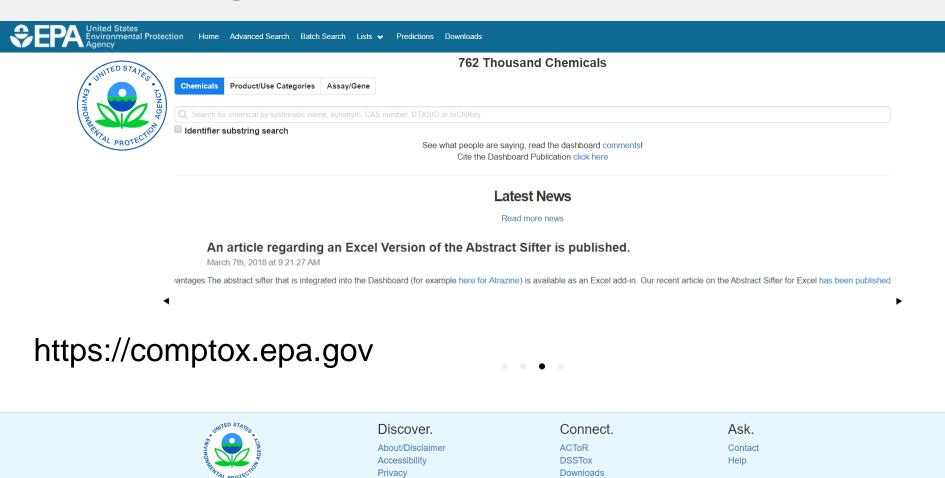


The General Approach

Analytical Instruments Comp. Tools & Workflows **IDENT** $C_8H_{10}N_4O_2$ $C_{17}H_{19}NO_3$ C17H18F3NO 285.136 194.080 309.134 **Databases** ChemSpider Search and share chemistry CompTox Chemistry Dashboard MassBo



CompTox Dashboard

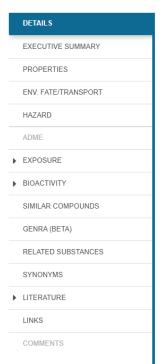


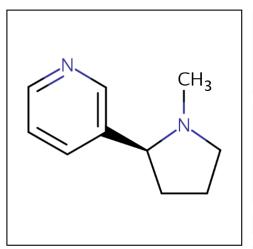


Nicotine

54-11-5 | DTXSID1020930

Searched by Approved Name.













Home Advanced Search Batch Search Lists ♥ Predictions Downloads

Property

Summary

▲ Download ▼

Columns V



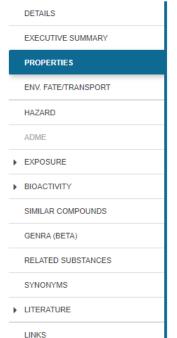
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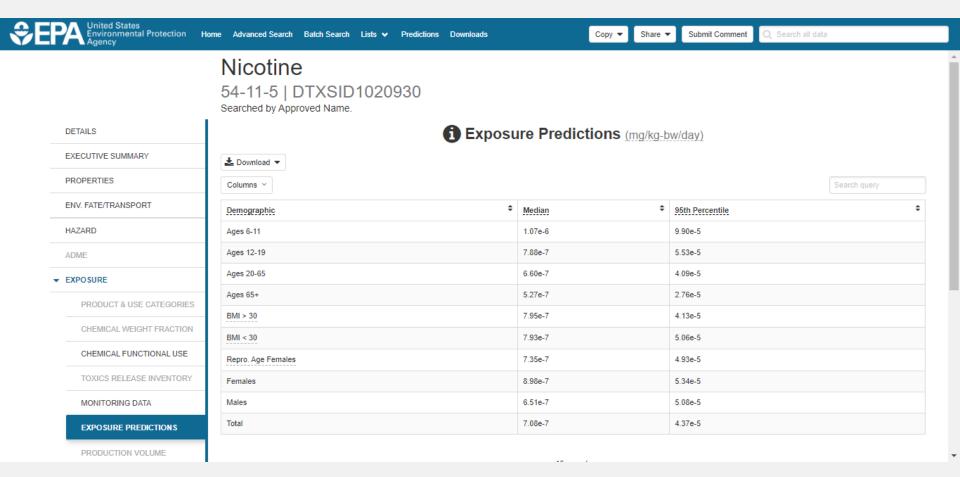
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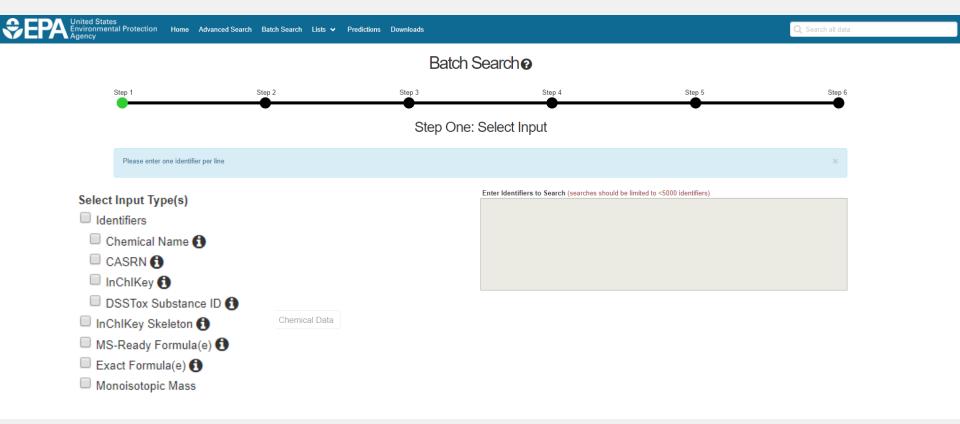
Summary

Property \$	Experimental average	Predicted average \$	Experimental median	Predicted median	Experimental range \$	Predicted range \$	<u>Unit</u>
LogP: Octanol-Water	1.17 (1)	0.751		0.821	1.17	3.85e-2 to 1.18	
Melting Point	-79.0 (3)	12.4	-79.0	13.4	-79.0	-34.4 to 57.3	°C
Boiling Point	247 (2)	249	247	248	247	244 to 254	°C
Vapor Pressure	3.80e-2 (1)	1.70e-2		1.76e-2	3.80e-2	2.39e-3 to 3.03e-2	mmHg
Water Solubility	6.16 (1)	3.74		4.51	6.16	8.00e-2 to 6.63	mol/L
Flash Point	-	99.8		99.8	-	97.9 to 102	°C
Surface Tension	-	38.6		38.6	-	37.7 to 39.6	dyn/cm
Index of Refraction	-	1.54			-	1.54	
Molar Refractivity	-	49.3			-	49.3	cm^3
Polarizability	-	19.5			-	19.5	Å^3

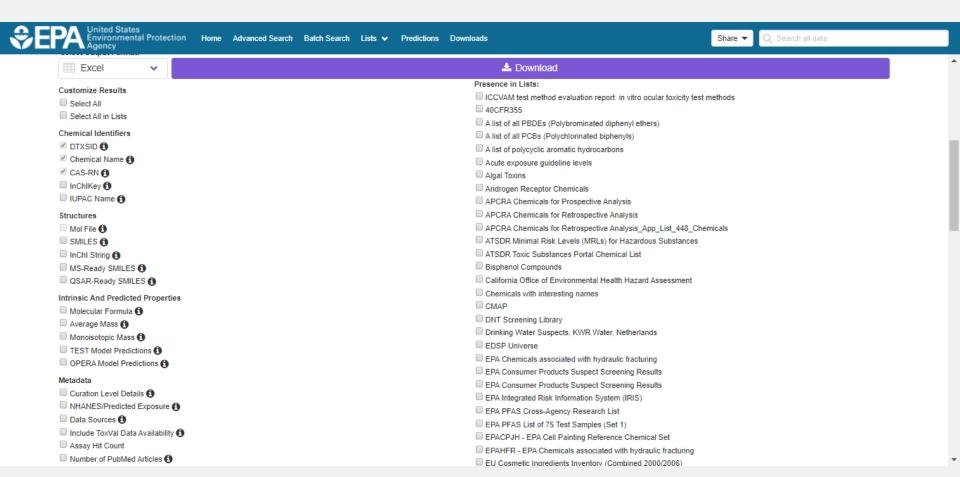






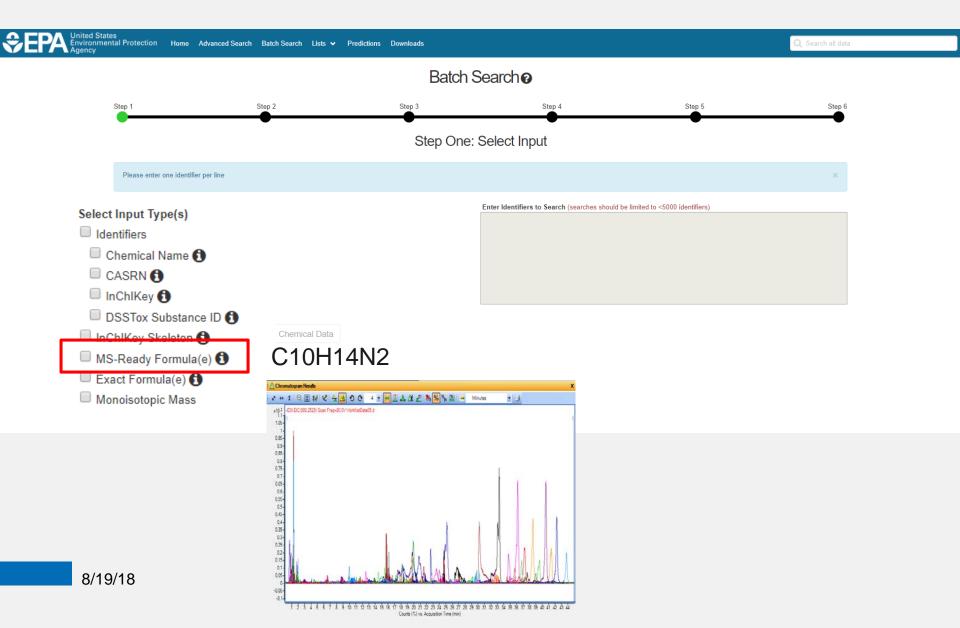






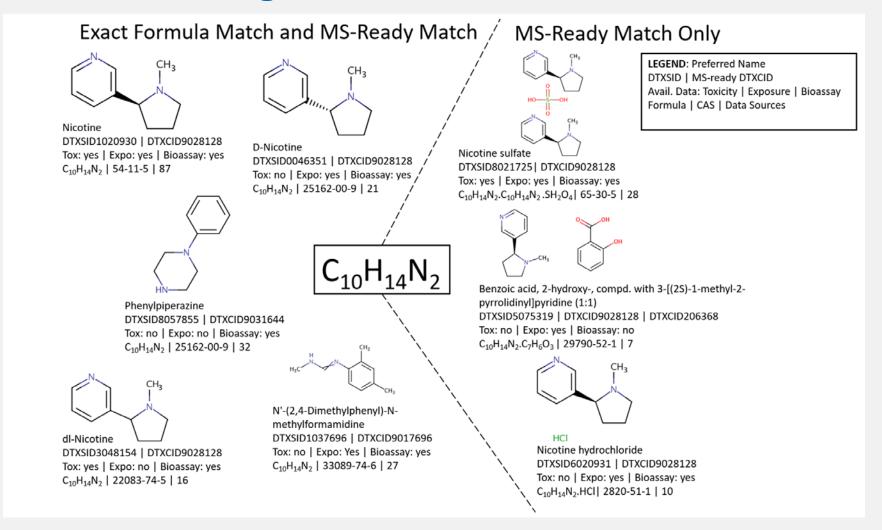


Batch Search for SSA/NTA





MS-Ready Structures improve database searching





Data Source Ranking for Identification in SSA/NTA

- Mass and/or formula unknown to a researcher, contained within a reference database
- Most likely candidate chemicals have the most references/sources

C14H22N2O3 266.16304



Chemical Reference Database



Sorted candidate structures





RAPID COMMUNICATION

Identifying known unknowns using the US EPA's CompTox Chemistry Dashboard

Andrew D. McEachran 1 · Jon R. Sobus 2 · Antony J. Williams 3

 On same 162 chemicals, Dashboard outperforms ChemSpider

	Mass-based sear	Mass-based searching		Formula-based searching	
	Dashboard	ChemSpider	Dashboard	ChemSpider	
Average rank position Percent in #1 position	1.3 85%	2.2 ^a 70%	1.2 88%	1.4 80%	

^a Average rank in ChemSpider shown here does not include an outlier where the rank was 201, when added the average rank position is 3.5



Increasing Data Streams to Improve Identifications



- US EPA CompTox Dashboard Data Sources (DS)
- PubChem Data Source Count
- PubMed Reference Count



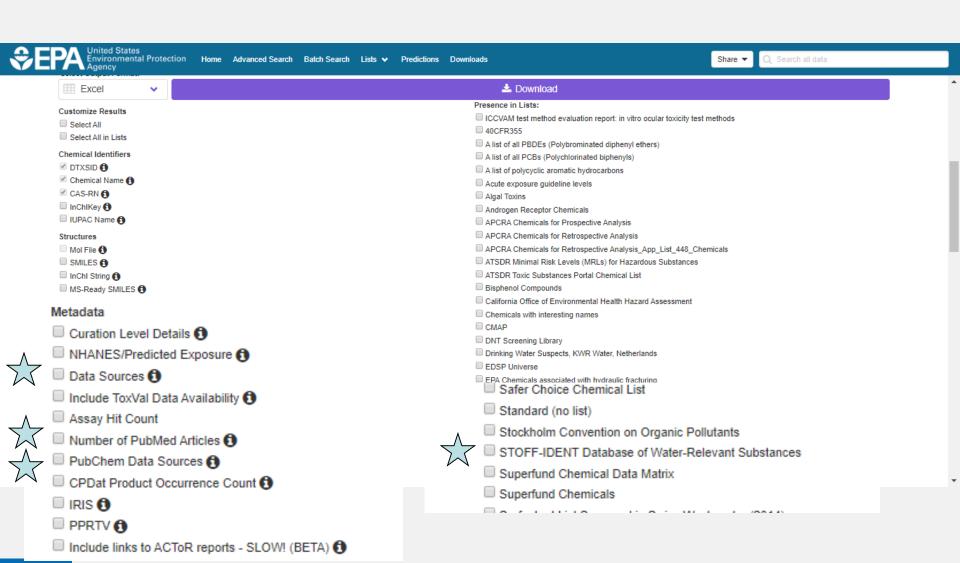
- Presence in STOFF-IDENT Database
- Predicted Environmental Media Occurrence
- CPDat Product Occurrence Count
- OPERA PhysChem Properties
- NORMAN Network Priority List







All available via Batch Search:

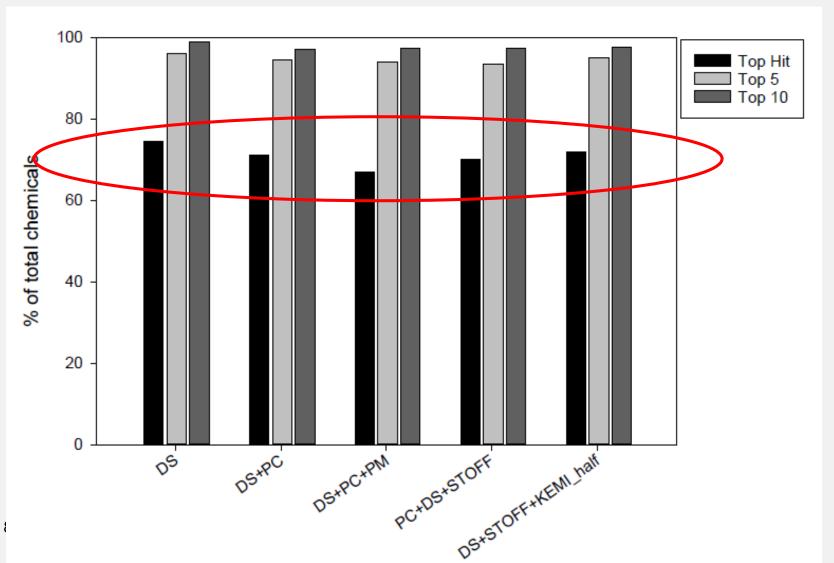


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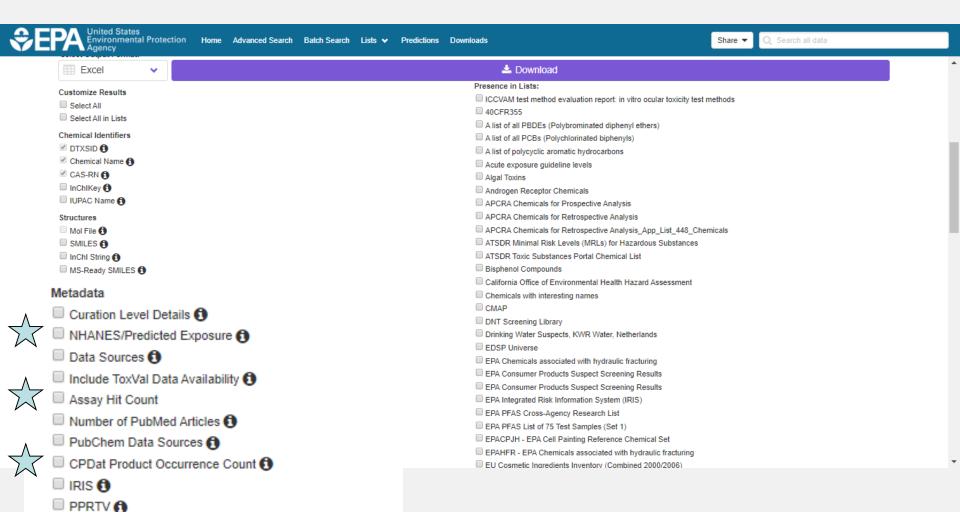
Identification ranks for 1783 chemicals using multiple data streams

$$SC_{TOTAL} = SC_{DS} + SC_{PM} + SC_{RT} + SC_{MO} + \cdots$$





Linking exposure potential and bioactivity for prioritization



Include links to ACToR reports - SLOW! (BETA)





Environmental Pollution

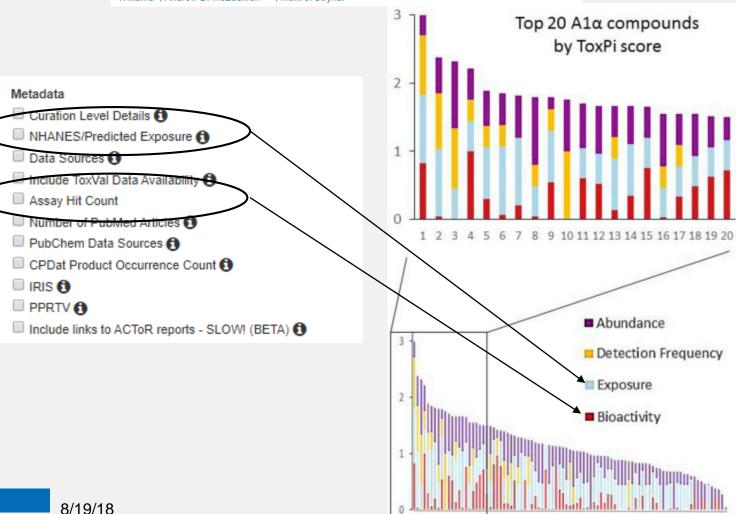
Volume 234, March 2018, Pages 297-306



All A1α compounds

Suspect screening and non-targeted analysis of drinking water using point-of-use filters \$\price \tag{\tag{7}}

Seth R. Newton a Rebecca L. McMahen a, b, Jon R. Sobus a, Kamel Mansouri b, c, 1, Antony J. Williams °, Andrew D. McEachran b, °, Mark J. Strynar a



	Compound	
1	1,2-Benzisothiazolin-3-one*	2.99
2	Diethyleneglycol	2.38
3	N-[3-(Dimethylamino)propyl] methacrylamide	2.32
4	Nonylparaben	2.22
5	Dipentyl phthalate	1.89
6	2-[2-(2-Butoxyethoxy) ethoxy]ethanoi*	1.85
7	N,N-Dimethyldodecan- 1-amine*	1.81
8	Sucralose	1.80
9	PFOS*	1.79
10	2-(2-Ethoxyethoxy) ethyl acetate*	1.76
11	TDCPP*	1.71
12	Zearalanol	1.67
13	PFOA*	1.66
14	Butylparaben	1.66
15	Noristerat	1.65
16	p-Synephrine	1.55
17	Alprostadil	1.55
18	Sclareol	1.55
19	PFDA*	1.51
20	Simvastatin	1.50

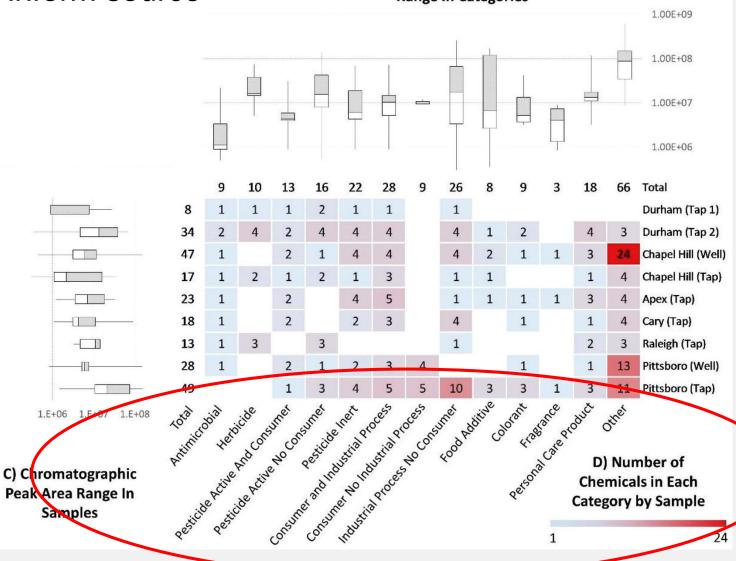
*Confirmed with standard



CPCat terms can inform source

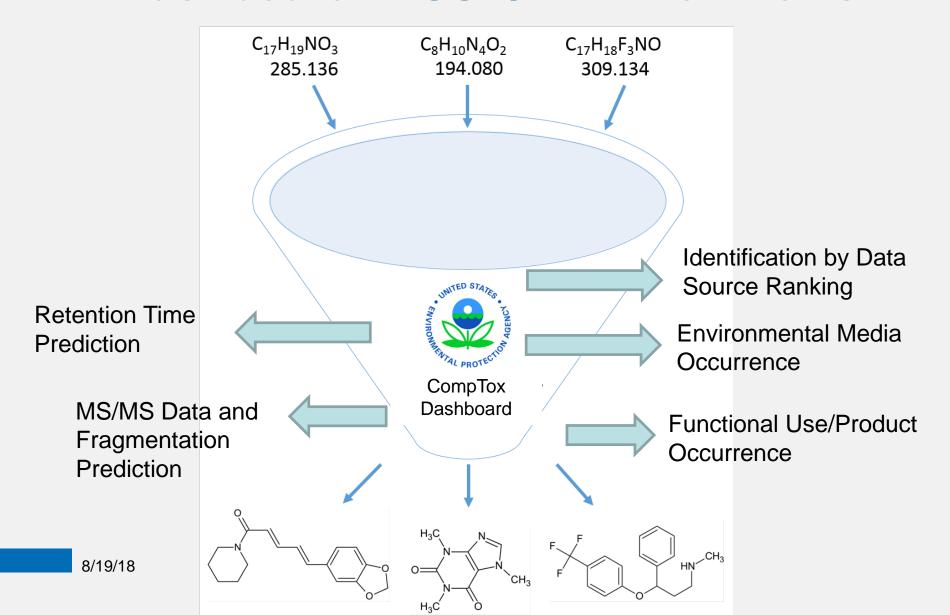
B) Chromatographic Peak Area **Range In Categories**

1





Dashboard in SSA/NTA Workflows





Future Directions

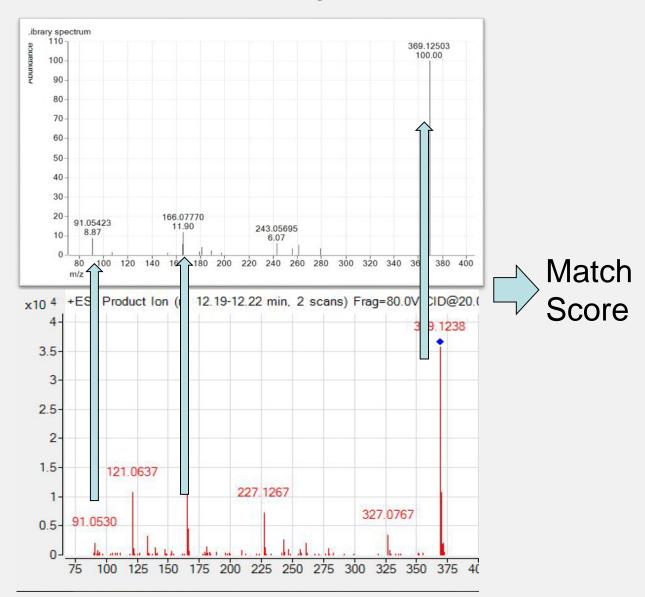
- Media occurrence prediction models
- Combined data visualization
- MS/MS match score calculations via the Dashboard
- Retention time index (RTI) predictions
- Ongoing expansion of the database
- Integration to public MS databases



MS/MS Spectral Matching for Identification

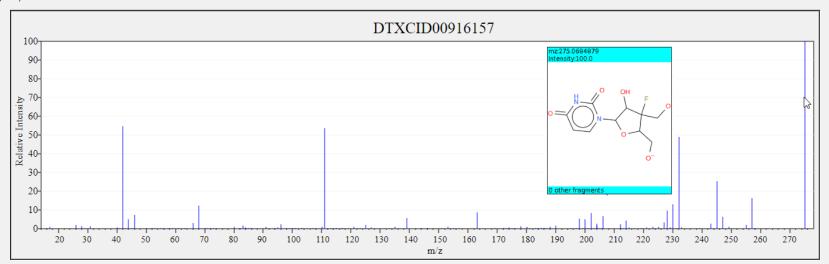
Library Fragmentation Spectra (20eV)

Observed Fragmentation Spectra (20eV)





Predicted MS/MS Spectral Matching for Identification



CASMI Contest Challenge Set (n=208)

CFM-ID only

	# Identified	% of Total
#1 Hits	89	43%
Top 5	154	74%
Top 10	174	84%
Top 20	190	91%

CFM-ID +DSSTox Data Sources

	# Identified	% of Total
#1 Hits	154	74%
Top 5	195	94%
Top 10	198	95%
Top 20	202	97%



Conclusions

- CompTox Dashboard provides access to a wealth of chemistry data that can be leveraged for exposure analyses
- NTA/SSA using the Dashboard results in high performance of identifications
- New data streams and predictions linked within the Dashboard further enhance NTA/SSA capabilities



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Halle

Kamel Mansouri- ILS, Inc

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Questions?

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- Associated presentations:
 - ANYL 100: Developing tools for high resolution mass spectrometry-based screening via the EPA's CompTox Chemistry Dashboard
 - ENVR 152: EPA Comptox Chemistry Dashboard as a data integration hub for environmental chemistry data
 - AGRO 107: Consensus ranking and fragmentation prediction for identification of unknowns in high resolution mass spectrometry