

Building a Non-Targeted Analysis Research Program at the U.S. EPA

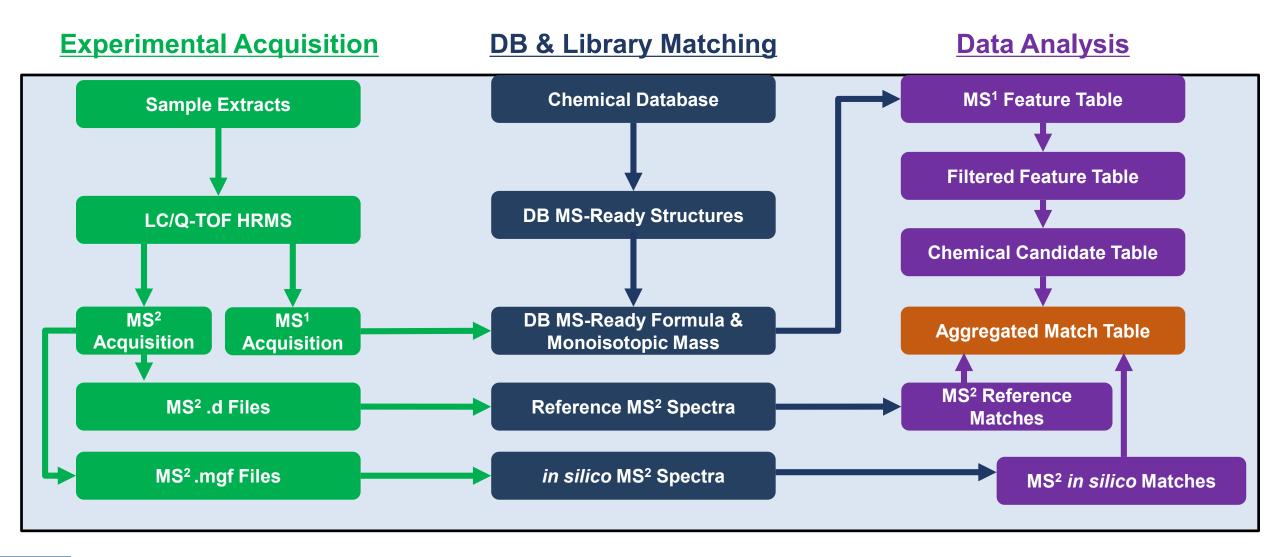
Jon R. Sobus, Ph.D. & the EPA/ORD NTA Team

Center for Computational Toxicology and Exposure Research Triangle Park, NC

Office of Research and Development

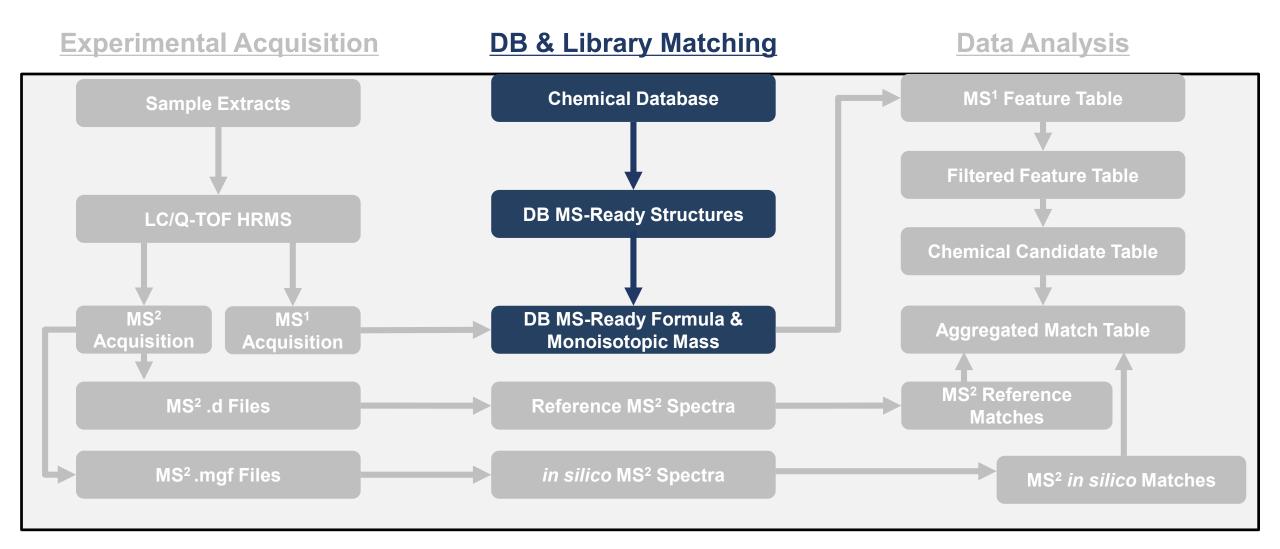


Agilent LC/Q-TOF Simplified Workflow



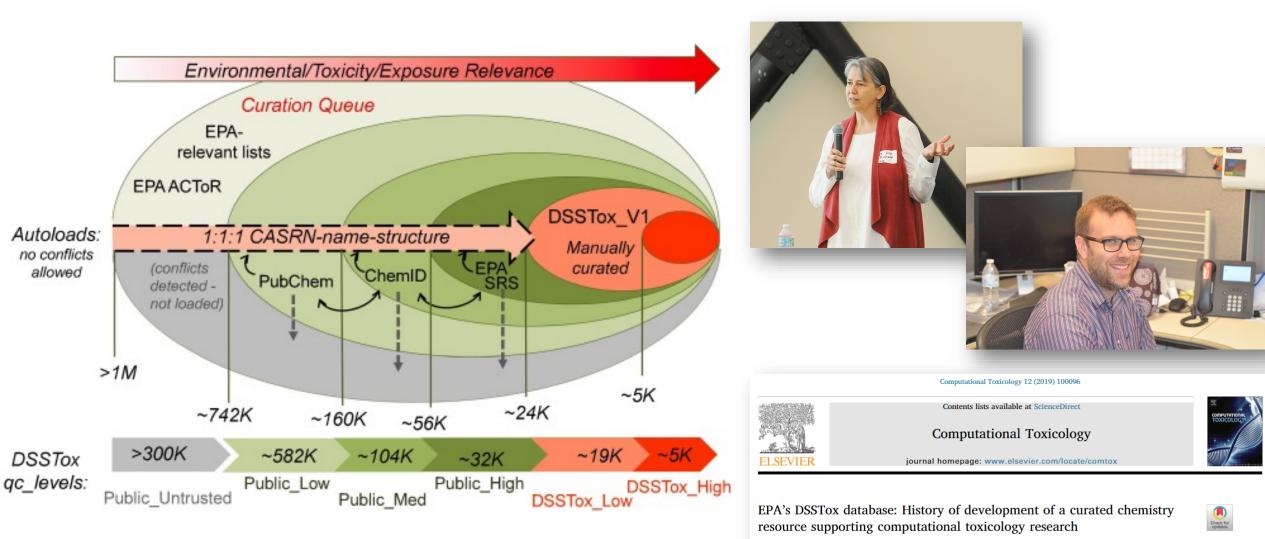


Agilent LC/Q-TOF Simplified Workflow





Chemical Database = DSSTox



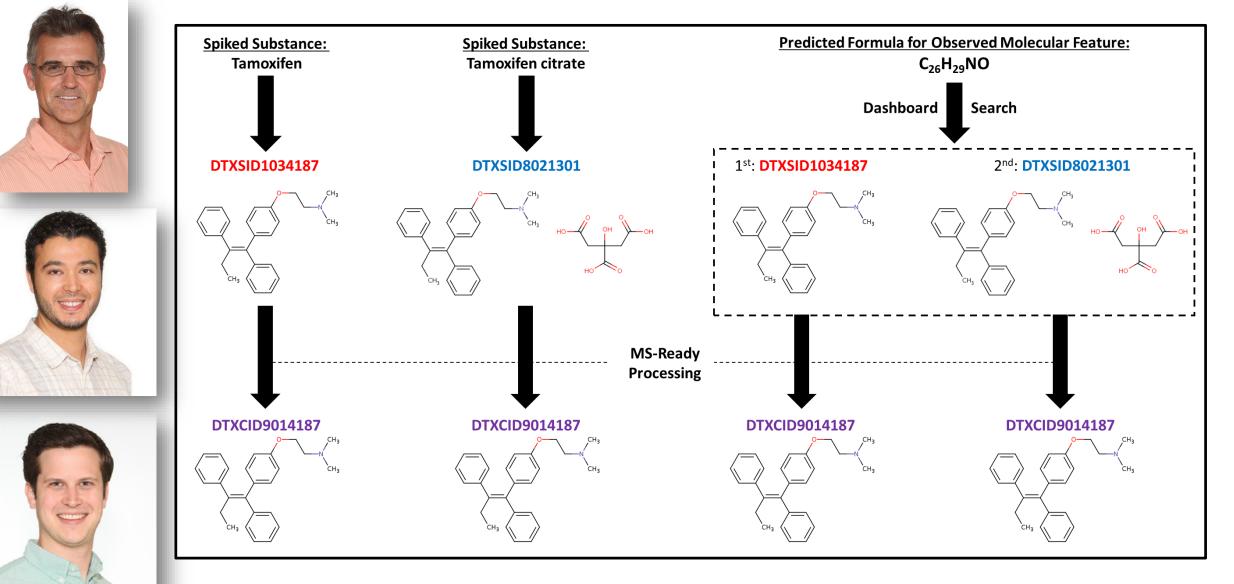
Christopher M. Grulke^a, Antony J. Williams^a, Inthirany Thillanadarajah^b, Ann M. Richard^{a,*}

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^b Senior Environmental Employment Program, US Environmental Protection Agency, Research Triangle Park, NC 27711, USA



MS-Ready Structures



Dashboard Access

CH₂

DTXCID50209864 C₆H₈N₂O

DTXCID8023761

 $C_{13}H_{10}N_4O_5$

Agency						
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The states of th	emicals Product/Use Categories Assay/Gene					
BONNE CONTRACTOR	Search for chemical by systematic name, synonym, CAS number, DTXSID or InChIKey					
WIAL PROTECTION II Id	entifier substring search					
	See what people are saying, read the dashboard comments! Cite the Dashboard Publication click here					
multiple components. However, mass spectrometry det	ss spectrometrists for the purpose of structure identification. A normal formula search would search the exact formula associated with any chemical, whether it include sol ects ionized chemical structures and molecular formulae searches should be based on desalted, and desolvated structures with stereochemistry removed. We refer to thes eadsheets containing the Preferred Name, CAS-RN. DTXSID, Formula, Formula of the MS-ready structure and associated masses, SMILES and InChI Strings/Keys. (UPDATED	e as "N	of hyd 1S rea	dy stru	salts o	or
McEachran <i>et al. J Cheminform (2018) 10:45</i> https://doi.org/10.1186/s13321-018-0299-2	Journal of Cheminformatics Nicarbazin DTXSID6034762 $C_{19}H_{18}N_6O_6$ $C_{13}H_{10}N_4O_5/C_6H_8N_2O$	H ₃		ï		
METHODOLOGY	Open Access 11 The					
"MS-Ready" str	uctures for non-targeted	l		L		
	mass spectrometry screening	H N	CH3			

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

studies

PA

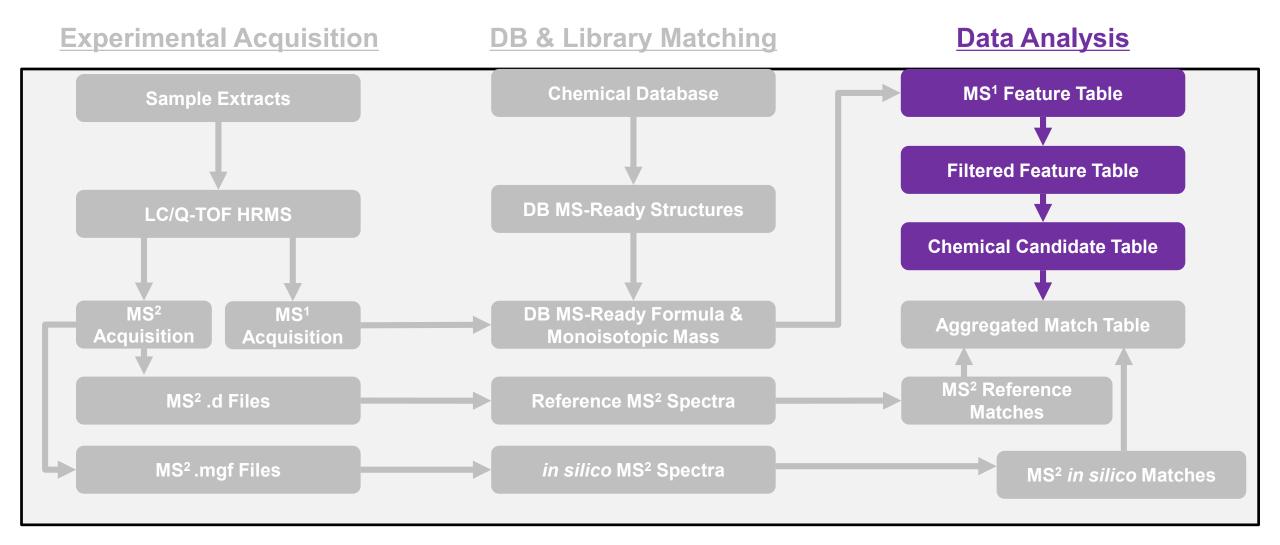
Environmental Protection

United States

Agonov



Agilent LC/Q-TOF Simplified Workflow





EPA NTA WebApp









Feature Removal:

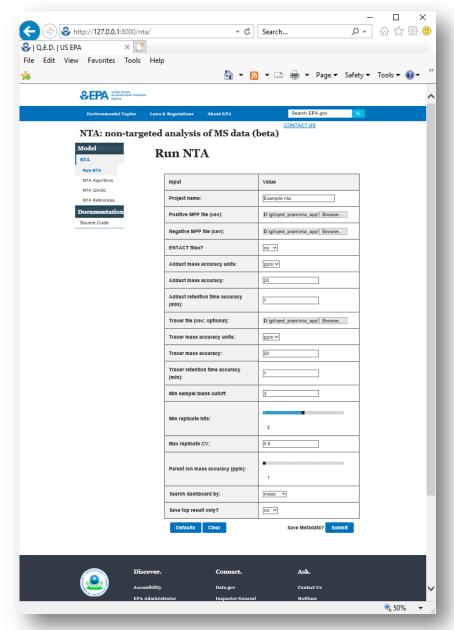
- 1) Duplicate features
- 2) Non-reproducible features
- 3) Blank features (sample:blank)
- 4) Non-responsive features (dilutions)

Feature Flagging:

- 1) Multi-mode hits (+ and -)
- 2) Meas. precision (CV threshold)
- 3) Formula match (score \geq threshold)
- 4) Negative mass defect
- 5) Halogenation
- 6) Has/is adduct
- 7) Has/is neutral loss
- 8) Has/is multimer

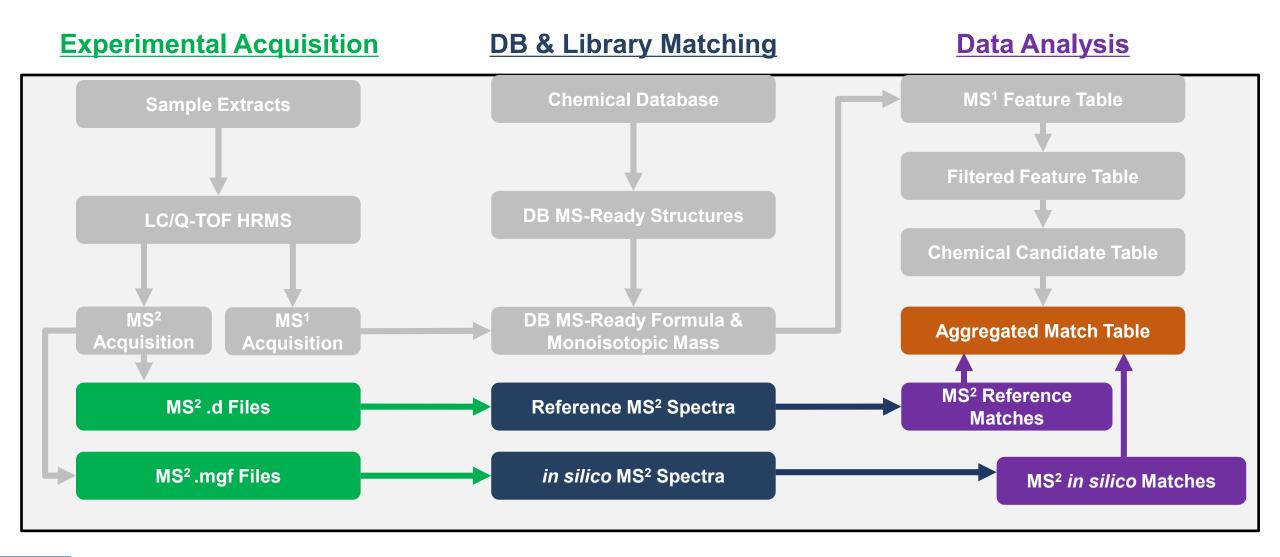
Dashboard Integration:

- 1) Data source & pub counts
- 2) Bioactivity & exposure levels
- 3) Presence on lists
- 4) Product & use categories





Agilent LC/Q-TOF Simplified Workflow





Generation of in silico Spectra

Linking *in silico* MS/MS spectra with

chemistry data to improve identification

CFM-ID v2.0

Competitive fragmentation modeling of ESI-MS/MS spectra for putative metabolite identification

Authors Authors and affi Felicity Allen , Russ Greiner, David Wishart			OWNS Iran [™] , Ilya Balabin, Tommy Cathey, Thomas R. Tr us & Antony J. Williams [™]	ransue, Hussein Al-Ghoul, Chris
	Machine Learning	Fragmentation Prediction Model	- ~~~	
<u>Training Set</u> : Metlin MS2 spectra and structures			DSSTox MS-Ready Structures (~765,000)	DSSTox MS2 spectra (10, 20, 40v)

McEachran, Andrew D., et al. *Scientific data* 6.1 (2019): 1-9 Allen, Felicity, et al. *Metabolomics* 11.1 (2015): 98-110.

CFM-ID Database Matching

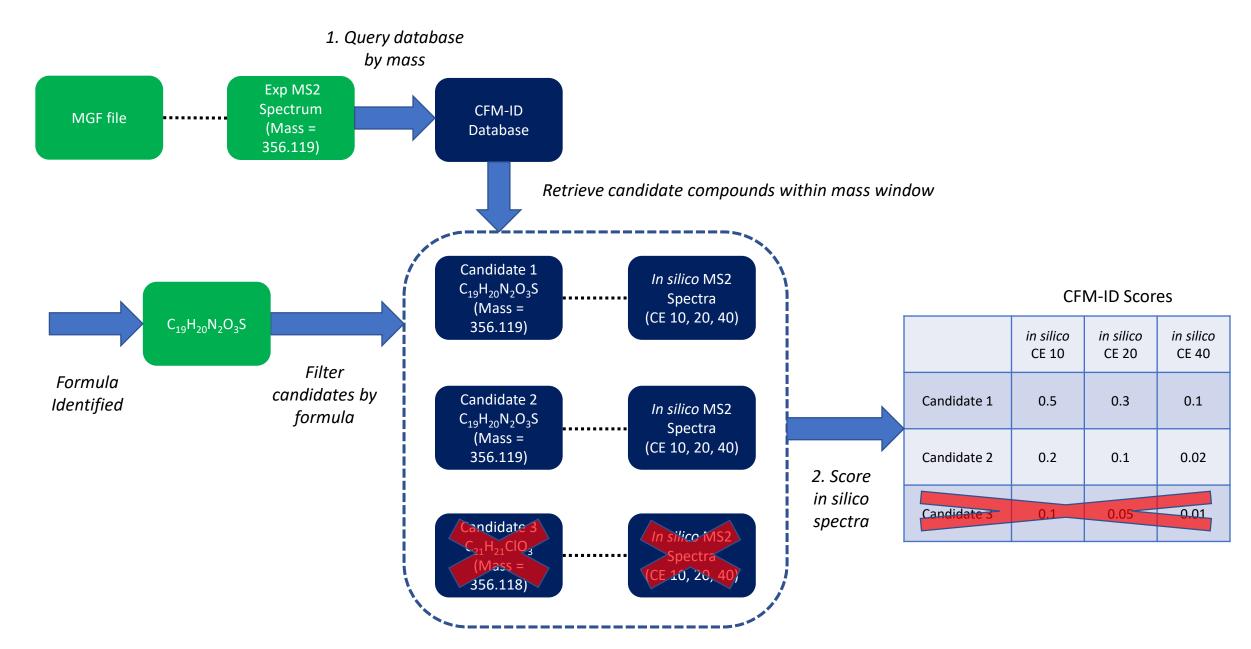
1. Query database Check for updates In silico MS/MS spectra for identifying unknowns: a critical by mass examination using CFM-ID algorithms and ENTACT mixture samples Exp MS2 Alex Chao^{1,2} • Hussein Al-Ghoul^{1,2} • Andrew D. McEachran^{1,3} • Ilya Balabin⁴ • Tom Transue⁴ • Tommy Cathey⁴ • Jarod N. Grossman^{2,3} • Randolph R. Singh^{1,5} • Elin M. Ulrich⁶ • Antony J. Williams⁷ • Jon R. Sobus⁶ Spectrum CFM-ID MGF file (Mass = Database 356.119) Retrieve candidate compounds within mass window Candidate 1 In silico MS2 $C_{19}H_{20}N_2O_3S$ **CFM-ID Scores** Spectra (Mass = (CE 10, 20, 40) 356.119) in silico in silico in silico CE 10 CE 20 CE 40 Candidate 2 Candidate 1 0.5 0.3 0.1 In silico MS2 C₁₉H₂₀N₂O₃S Spectra (Mass = (CE 10, 20, 40) 356.119) 0.2 0.1 0.02 Candidate 2 2. Score in silico Candidate 3 0.1 0.05 0.01 spectra Candidate 3 *In silico* MS2 $C_{21}H_{21}CIO_3$ Spectra (Mass = (CE 10, 20, 40) 356.118)

Analytical and Bioanalytical Chemistry (2020) 412:1303-1315

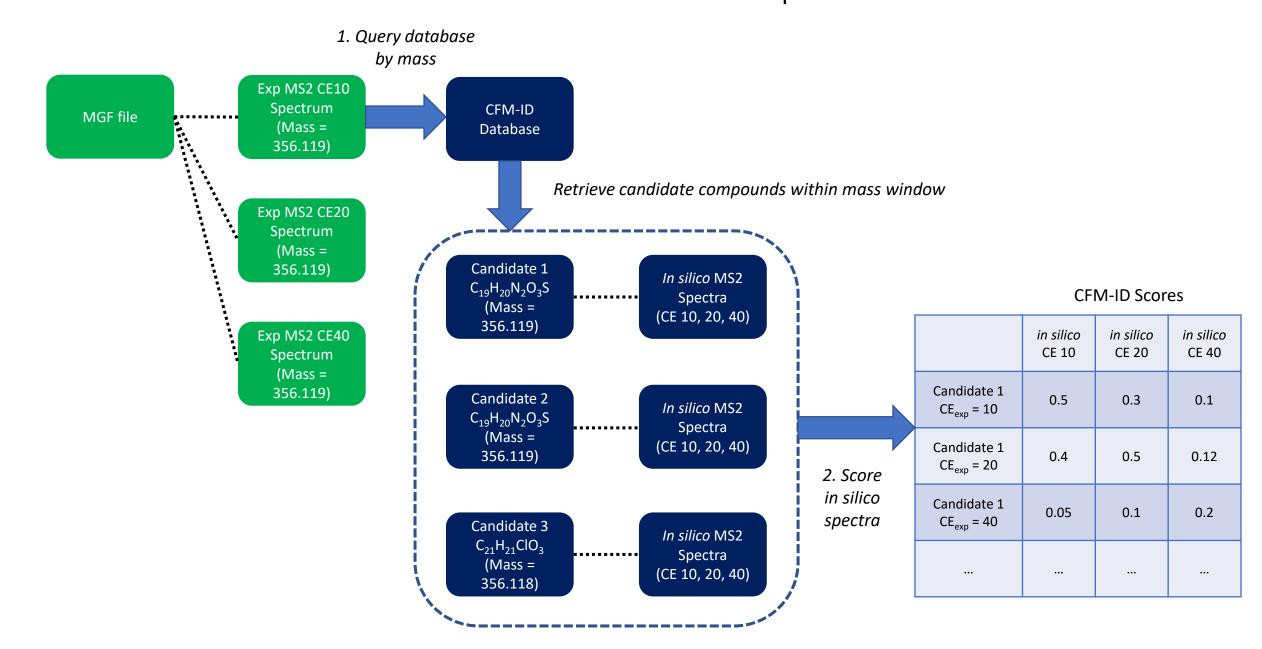
https://doi.org/10.1007/s00216-019-02351-7

RESEARCH PAPER

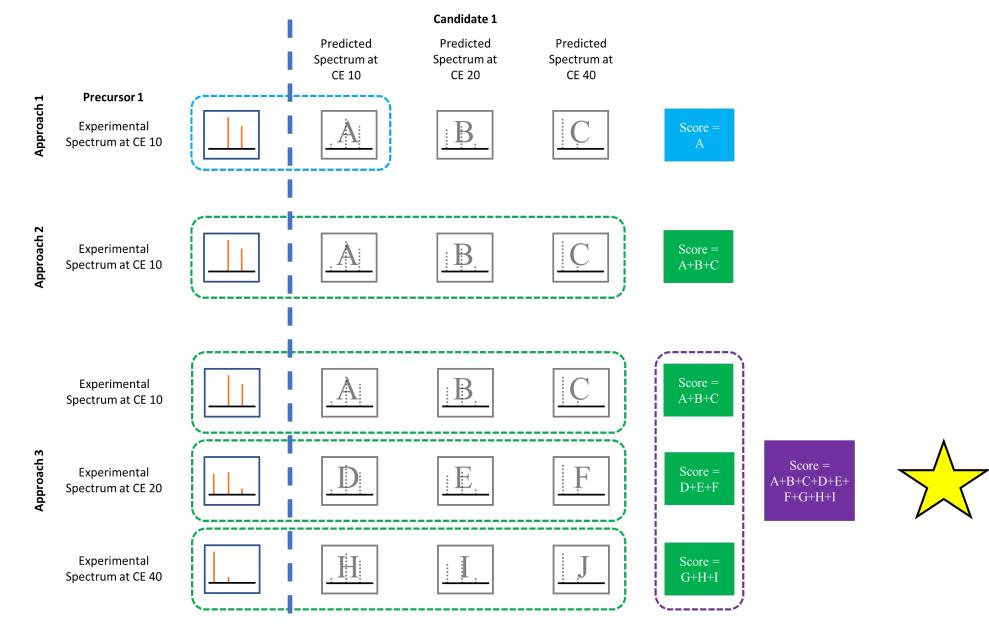
CFM-ID Database Matching (w/ Formula Information)



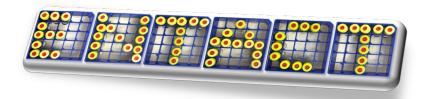
CFM-ID Database Matching (w/ Multiple CE_{experimental})



CFM-ID Scoring Approaches



EPA'S Non-Targeted Analysis Collaborative Trial



The Trial Mixtures:



10 Mixtures ranging from 95 to 365 compounds (Total: 1,269 unique compounds)

"Pass" compounds = 377 with MS2 data

EPA Setup:

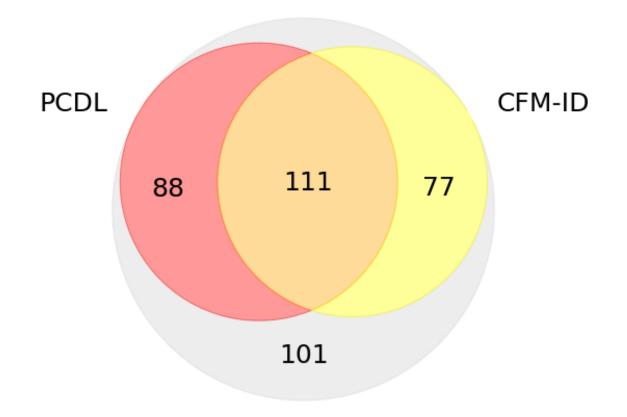


Agilent 1290 UPLC Agilent 6530B Q-TOF with ESI source

Ulrich, Elin M., et al. *Analytical and bioanalytical chemistry* 411.4 (2019): 853-866. Sobus, Jon R., et al. *Analytical and bioanalytical chemistry* 411.4 (2019): 835-851.



Reference vs. in silico Library Coverage



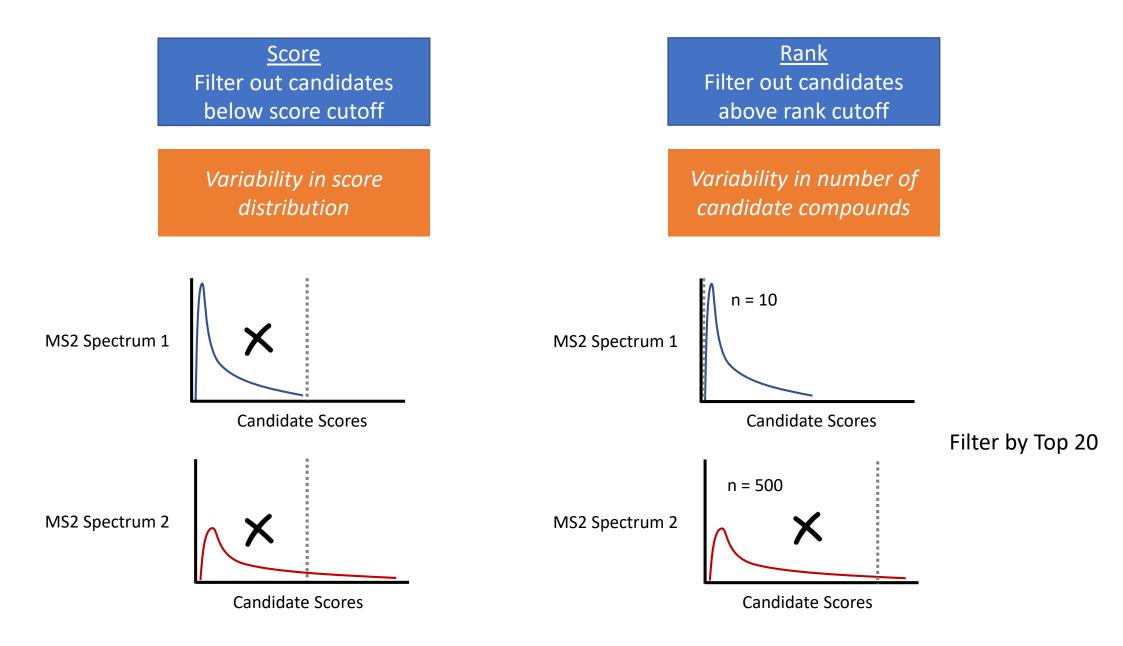
MS2 Library	% of "Pass" Compounds Identified
Agilent PCDL	53%
CFM-ID Top Hit	50%
PCDL and/or CFM-ID Top Hit	73%

"Pass" Compounds

PCDL \rightarrow Agilent reference MS² library

"Pass" compounds (n=377) \rightarrow ENTACT chemicals observed with MS² data

NTA Workflows: Using CFM-ID Results as Filters



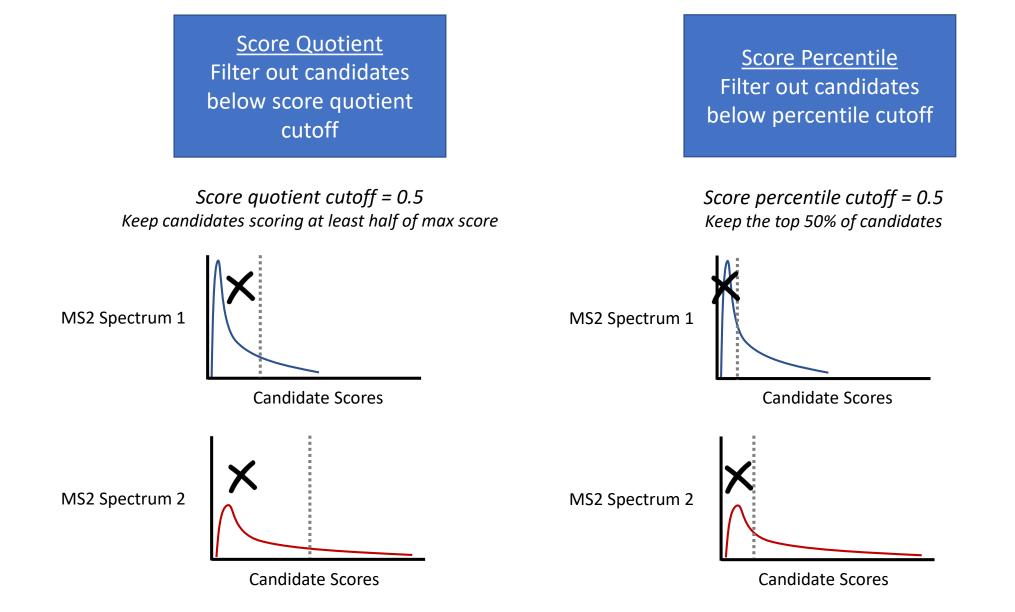
Normalizing CFM-ID Results Values

Score Quotient Normalize score to the highest candidate compound score Score Percentile Normalize rank to the number of candidate compounds

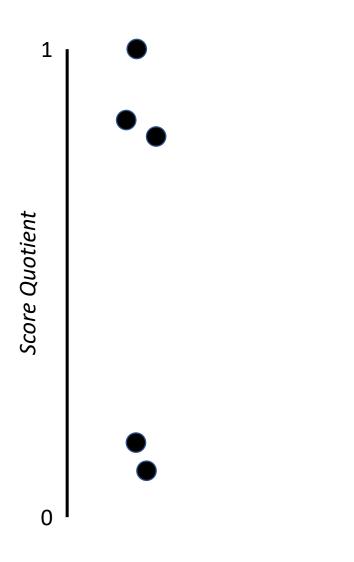
	Rank	CFM-ID Score	Maximum Score	Score Quotient	Score Percentile
Candidate Compound 1	1	0.5	0.5	1	100
Candidate Compound 2	2	0.4	0.5	0.8	80
Candidate Compound 3	3	0.39	0.5	0.78	60
Candidate Compound 4	4	0.1	0.5	0.2	40
Candidate Compound 5	5	0.05	0.5	0.1	20

Score Quotient = Score / Maximum Score

NTA Workflows: Using CFM-ID Normalized Results as Filters



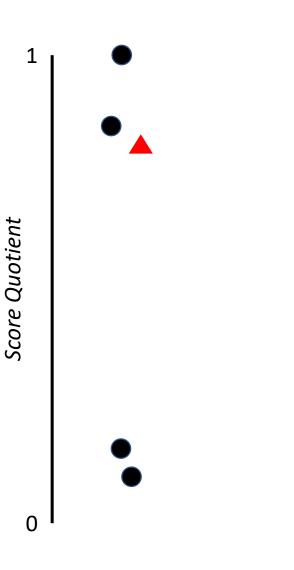
	CFM-ID Score	Maximum Score	Score Quotient
Candidate Compound 1	0.5	0.5	1
Candidate Compound 2	0.4	0.5	0.8
Candidate Compound 3	0.39	0.5	0.78
Candidate Compound 4	0.1	0.5	0.2
Candidate Compound 5	0.05	0.5	0.1



	CFM-ID Score	Maximum Score	Score Quotient
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Candidate Compound 3	0.39	0.5	0.78
Candidate Compound 4	0.1	0.5	0.2
Candidate Compound 5	0.05	0.5	0.1

L True Compound

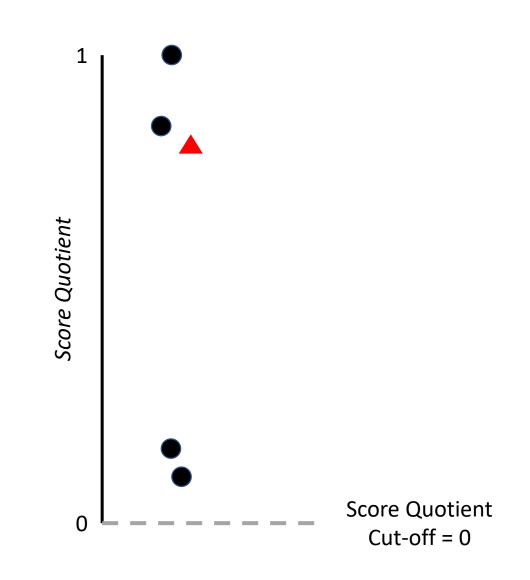
True Positives	
False Negatives	
True Negatives	
False Positives	



	CFM-ID Score	Maximum Score	Score Quotient
Candidate Compound 1	0.5	0.5	1
Candidate Compound 2	0.4	0.5	0.8
Candidate Compound 3	0.39	0.5	0.78
Candidate Compound 4	0.1	0.5	0.2
Candidate Compound 5	0.05	0.5	0.1

L True Compound

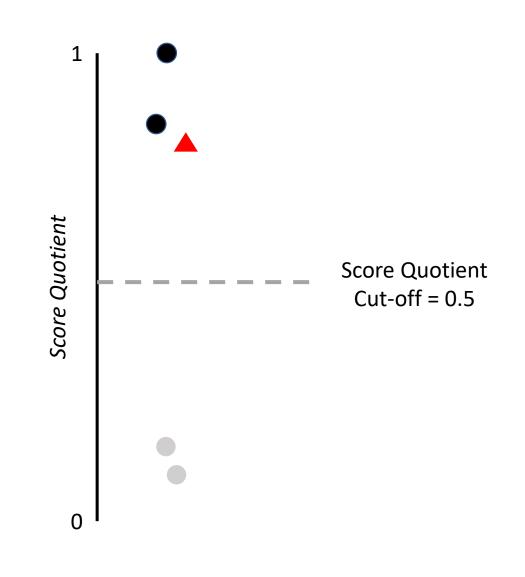
True Positives	1
False Negatives	0
True Negatives	0
False Positives	4



	CFM-ID Score	Maximum Score	Score Quotient
Candidate Compound 1	0.5	0.5	1
Candidate Compound 2	0.4	0.5	0.8
Candidate Compound 3	0.39	0.5	0.78
Candidate Compound 4	0.1	0.5	0.2
Candidate Compound 5	0.05	0.5	0.1

L True Compound

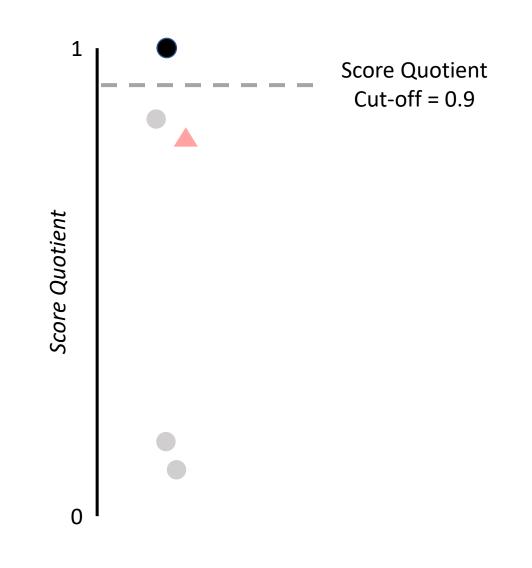
True Positives	1
False Negatives	0
True Negatives	2
False Positives	2



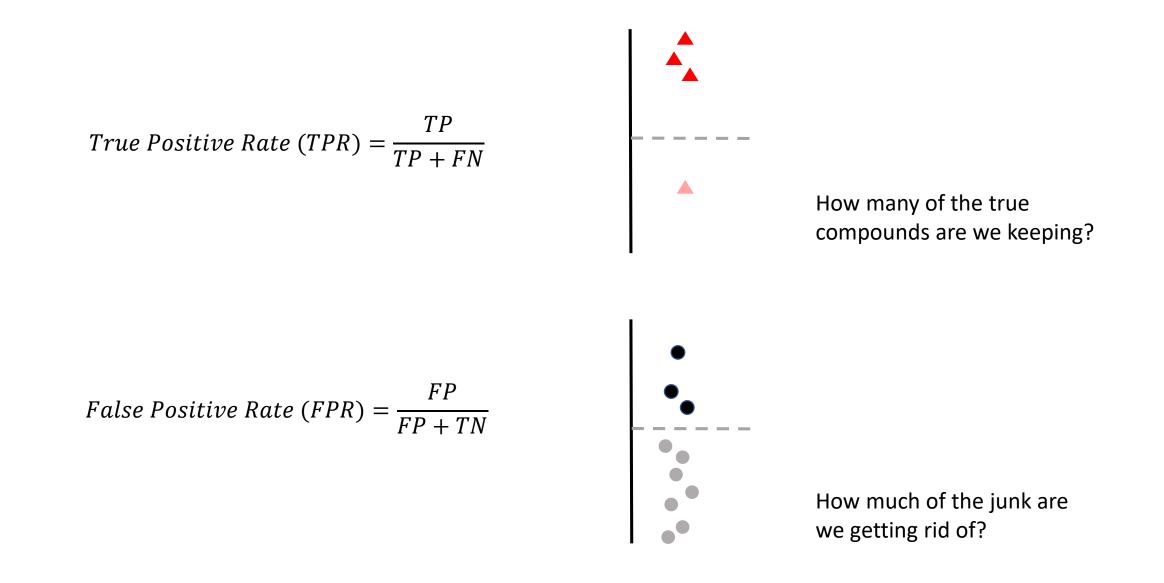
	CFM-ID Score	Maximum Score	Score Quotient
Candidate Compound 1	0.5	0.5	1
Candidate Compound 2	0.4	0.5	0.8
Candidate Compound 3	0.39	0.5	0.78
Candidate Compound 4	0.1	0.5	0.2
Candidate Compound 5	0.05	0.5	0.1

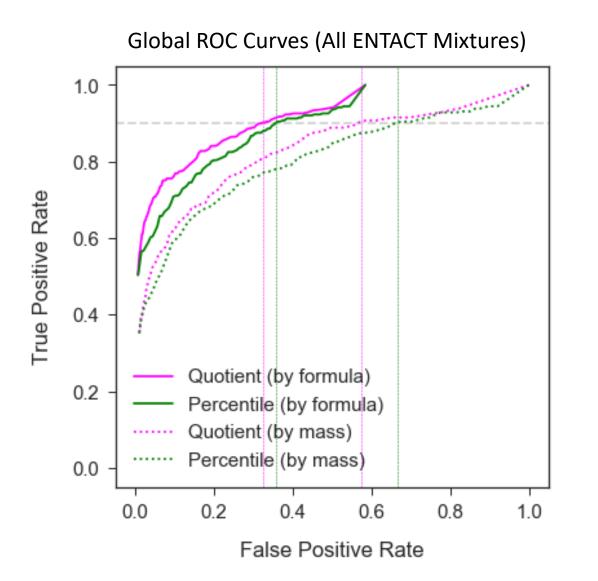
L True Compound

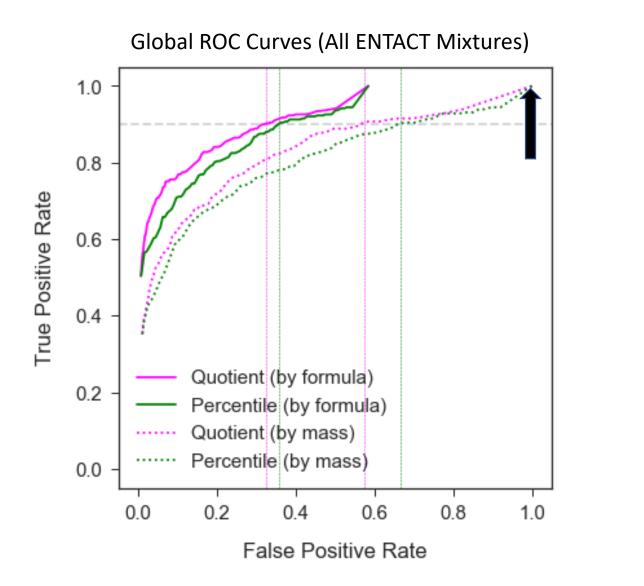
True Positives	0
False Negatives	1
True Negatives	3
False Positives	1

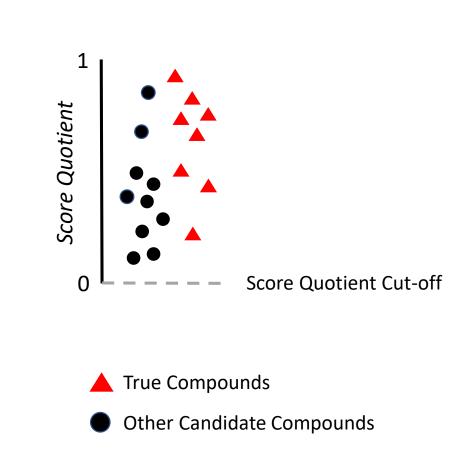


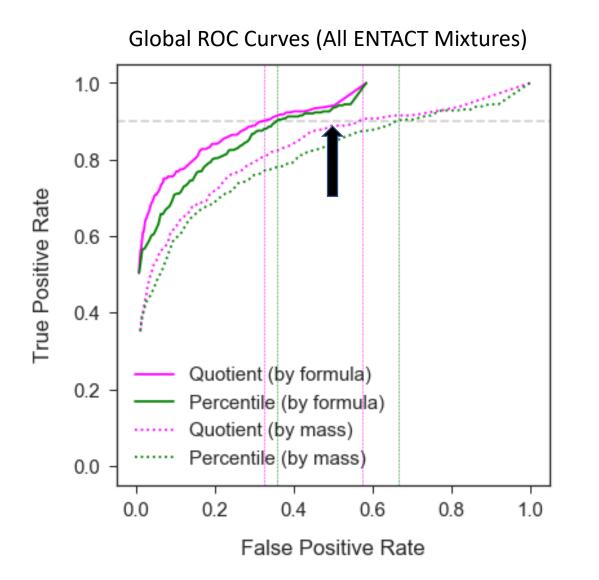
Balancing Cut-offs

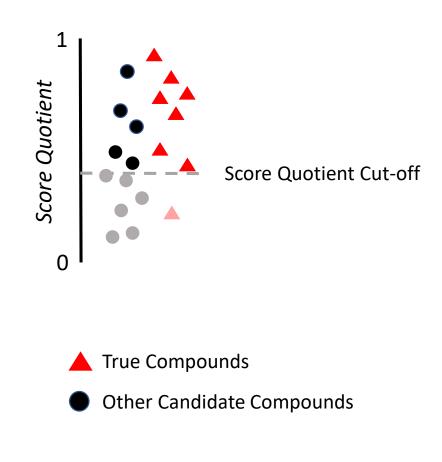


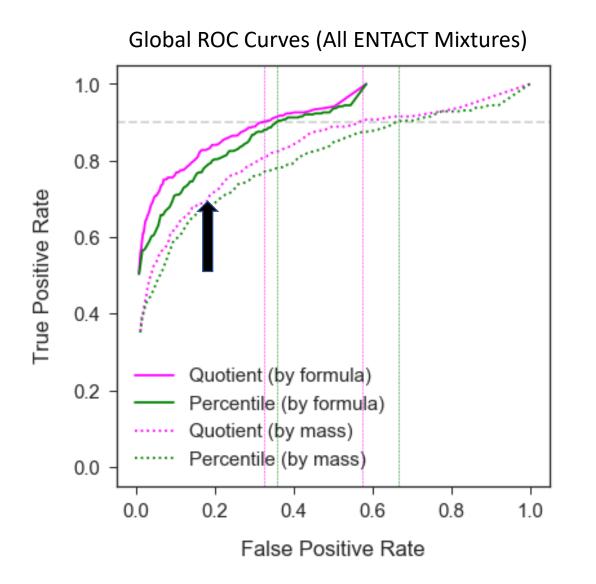


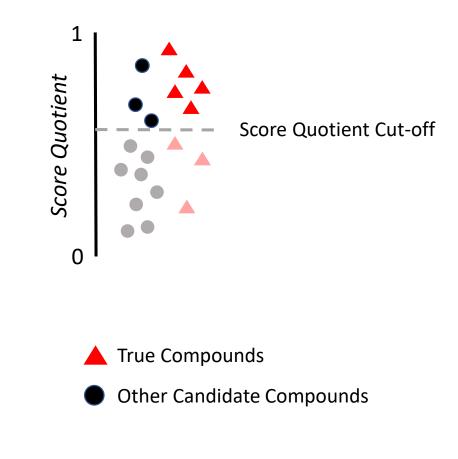


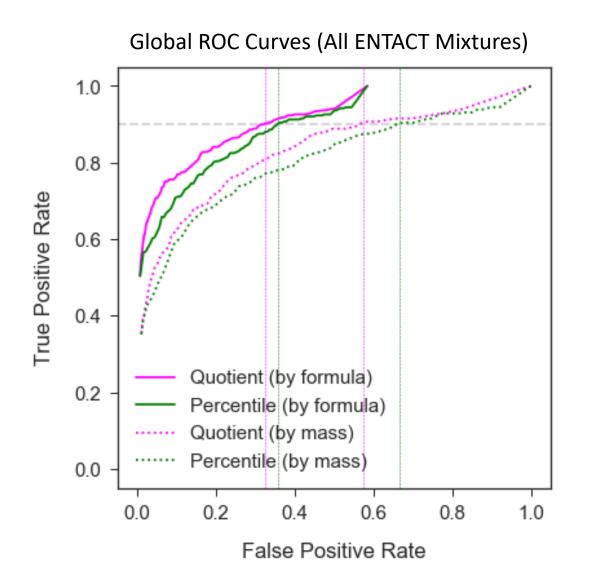


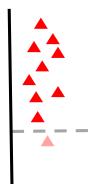






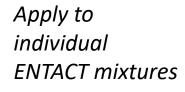




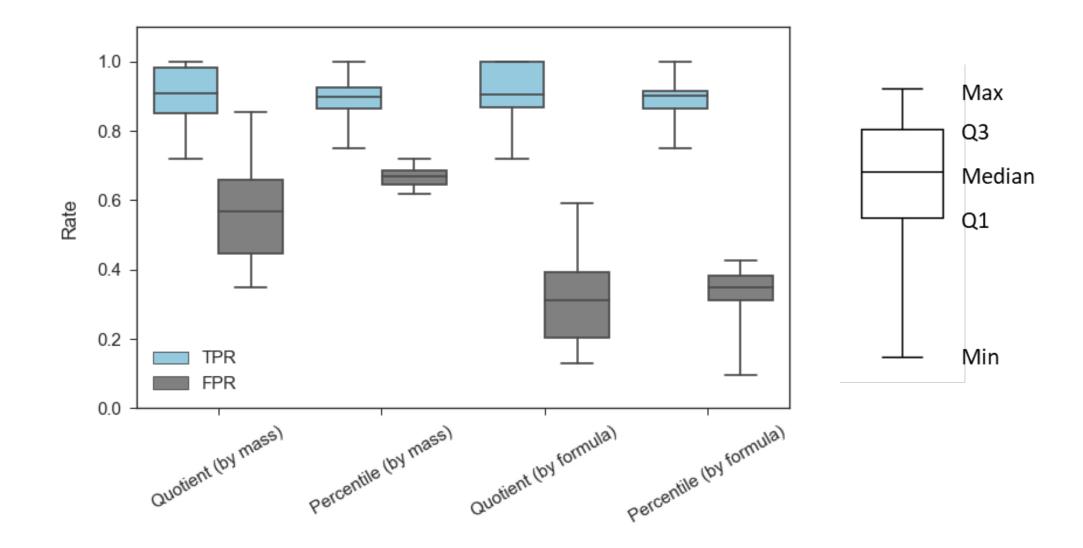


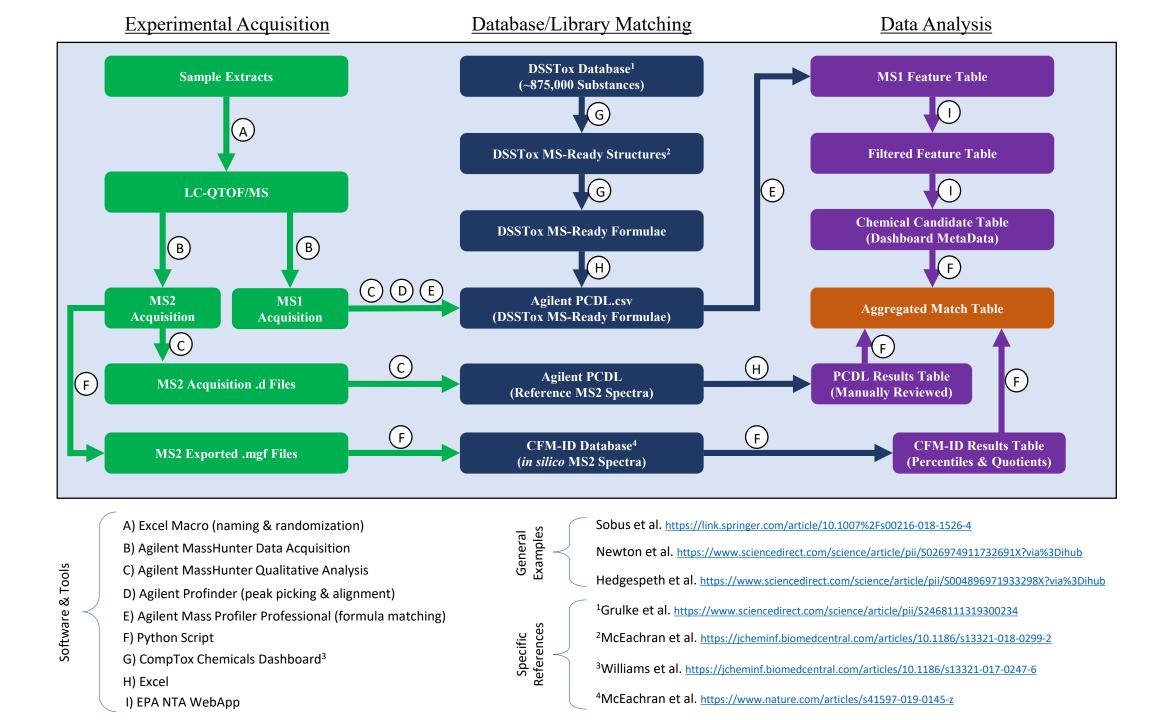
Cut-off Values for Global TPR = 0.9

	Cut-off value
Quotient (by formula)	0.18
Percentile (by formula)	38
Quotient (by mass)	0.13
Percentile (by mass)	32



CFM-ID Cut-off Filtering: Individual ENTACT Mixtures

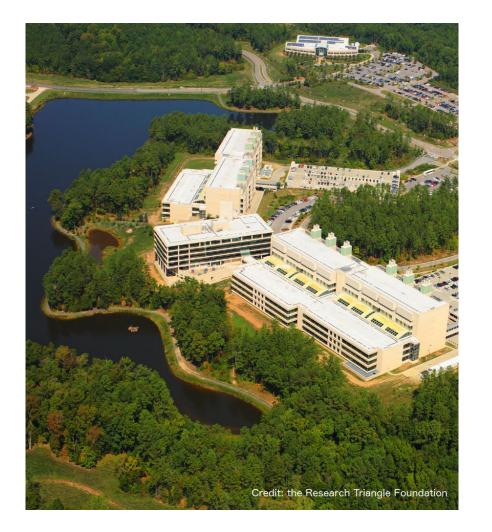




Contributing Researchers



This work was supported, in part, by ORD's Pathfinder Innovation Program (PIP) and an ORD EMVL award



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Chris Grulke Kamel Mansouri* Andrew McEachran* Ann Richard John Wambaugh Antony Williams

Agilent

Jarod Grossman Andrew McEachran

GDIT

Ilya Balabin Tom Transue Tommy Cathey

Questions?



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The views expressed in this presentation are those of the authors and do not necessarily represent the views or policies of the U.S. Environmental Protection Agency.

