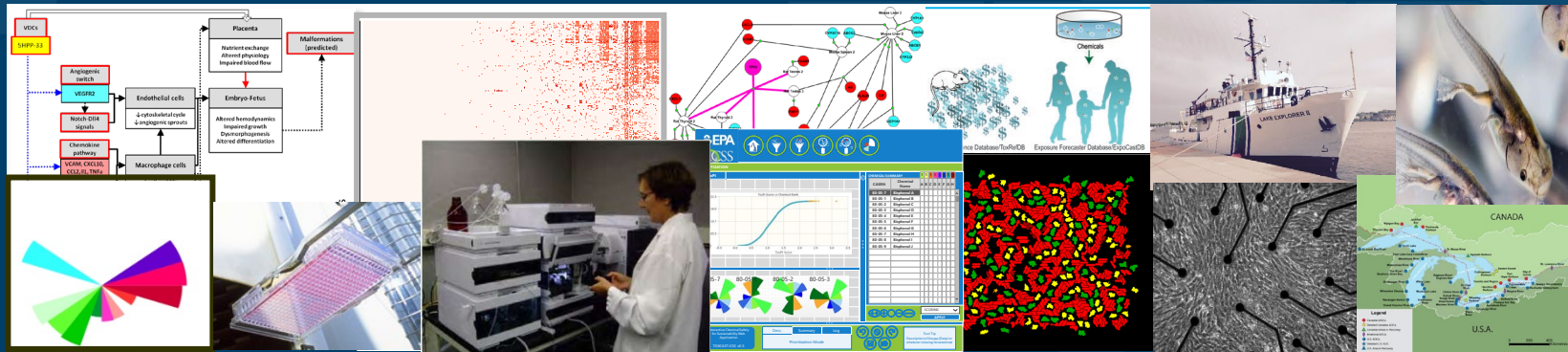


NAMs Paradox

When an Unstoppable Force Meets an Immovable Object



Alliance for Risk Assessment Workshop XIII

February 15, 2022

Rusty Thomas
Director
Center for Computational Toxicology and Exposure

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

The Original Paradox...



Teumessian fox and the hunting dog Laelaps

The NAM Paradox...



The unstoppable NAM force and the unmovable regulatory systems and processes

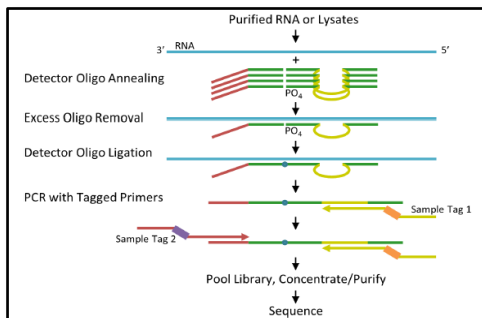
A 'Zeus-like' Seven Step Plan to Address This Paradox



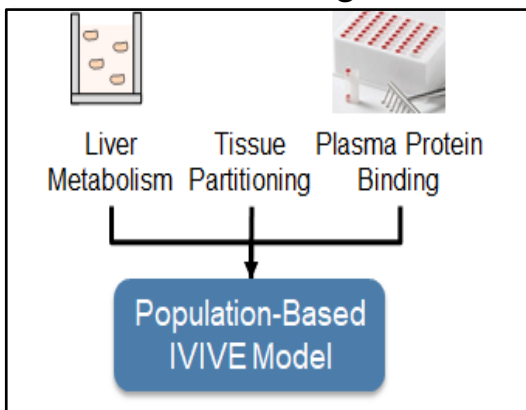
1. Continue to innovate with NAMs while systematically address the limitations (a couple examples...)
2. Accept that there is likely not a primary mechanism/mode of action for most environmental/industrial chemicals
3. Work through how to assemble NAMs in a coherent, practical, fit for purpose testing framework
4. Understand how to benchmark new approaches
5. Grapple with the issue of protection vs. prediction in our current and future approaches
6. Evaluate regulatory flexibilities and develop a fit for purpose validation/confidence framework to evaluating new approaches
7. Quantify public health and economic trade-offs of uncertainty, cost, and time in toxicity testing methods

Step 1: Continue to Innovate and Address Limitations in NAMs

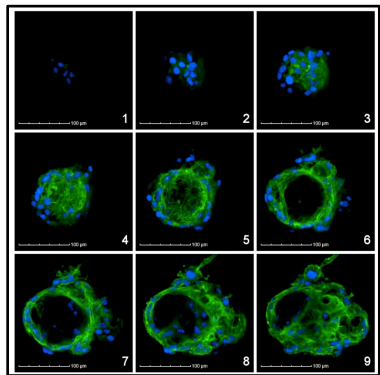
Whole Genome Transcriptomics



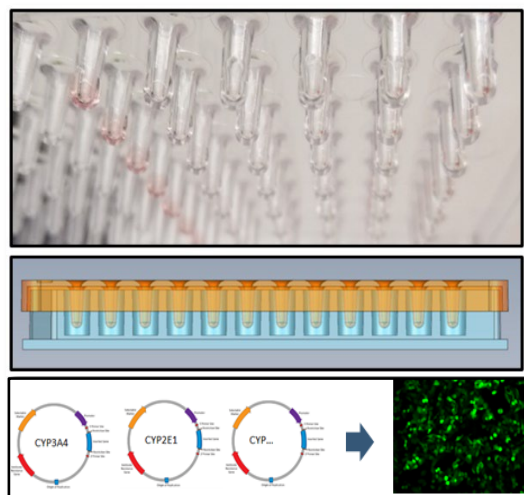
Toxicokinetic Measurements and Modeling



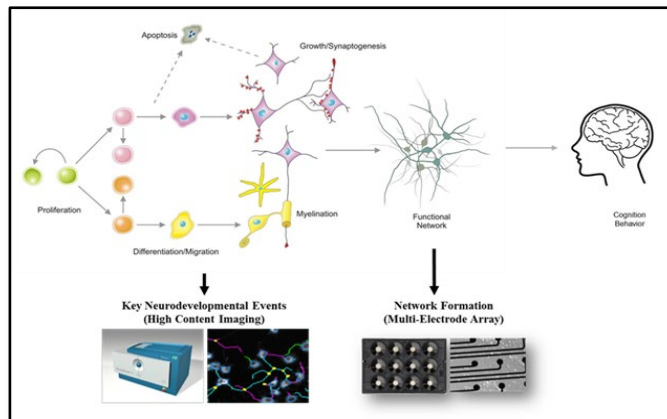
Organotypic Culture Models



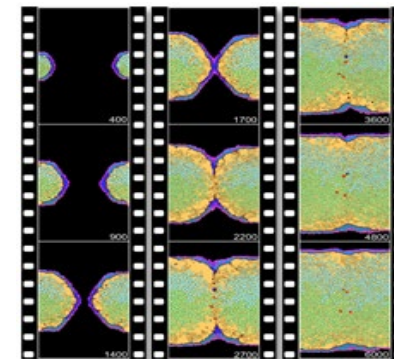
Metabolic Retrofitting



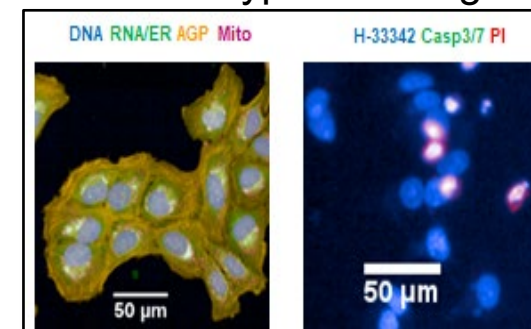
Integrated Approach to Testing and Assessment for DNT



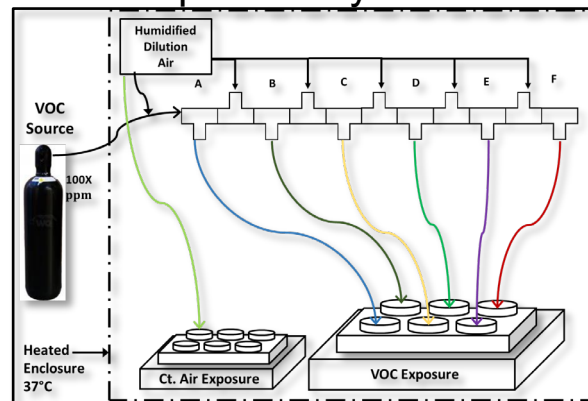
Virtual Tissue Models



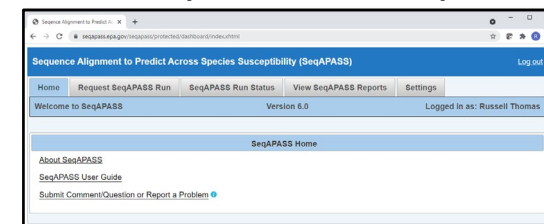
Multi-Parameter Cellular Phenotypic Profiling



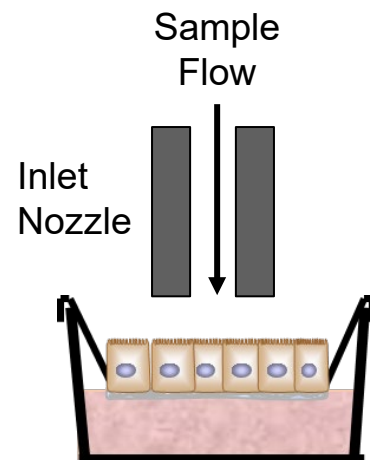
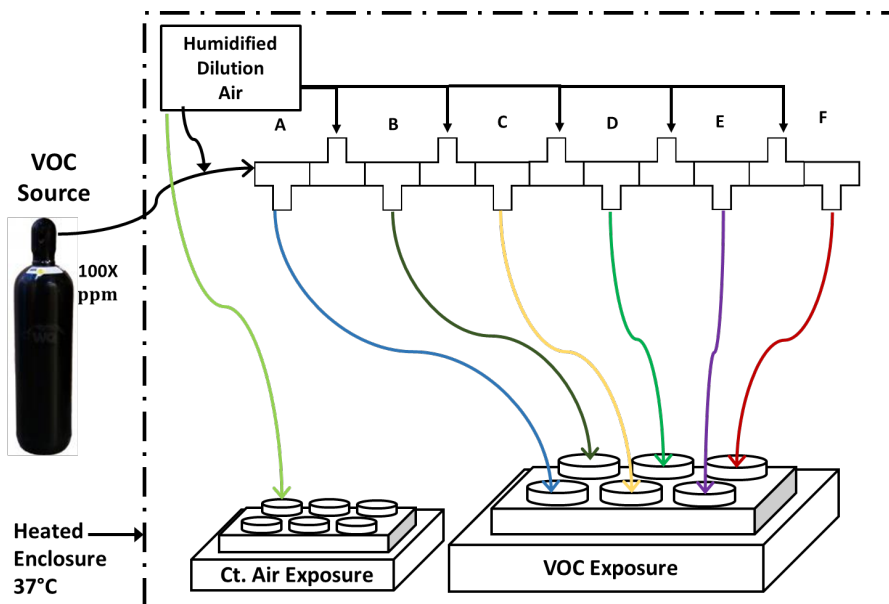
Volatile/Aerosol In Vitro Exposure Systems



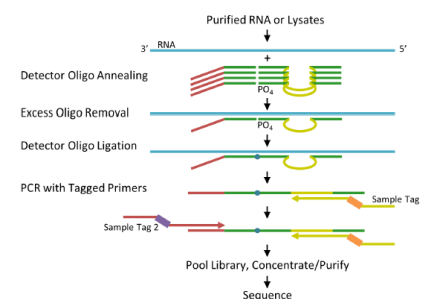
Sequence Alignment to Predict Across Species Susceptibility



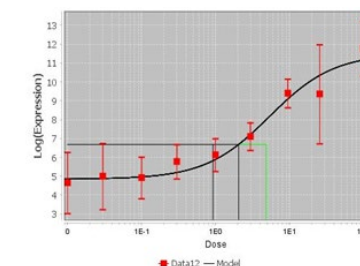
Developing *In Vitro* Exposure Systems for Volatile Chemicals



Whole Genome Transcriptomics (HTTr)



Concentration Response Modeling



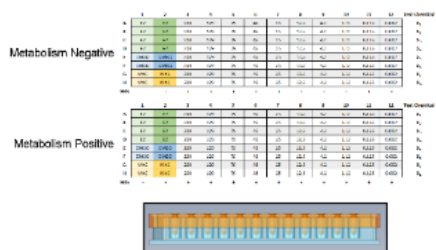
	ACGIH TLV-TWA (ppm)	BEAS-2B HTTr POD (ppm)	HBEC HTTr POD (ppm)
Acrolein	0.1	0.58	--
Formaldehyde	0.3	NA	--
1,3-Butadiene	10	13.98	--
Acetaldehyde	25	NA	--
1-Bromopropane	0.1 *	2.25	NA
Carbon Tetrachloride	10	9.56	NA
Trichloroethylene	50	44.8	28.1
Dichloromethane	100	142.13	266.7

* The ACGIH TLV TWA for 1-bromopropane was updated to 0.1 ppm in 2012. Prior to that the TLV-TWA for 1-bromopropane was 10 ppm.

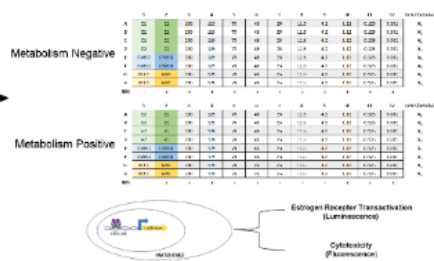
A.Speen (CPHEA), M. Higuchi
(CPHEA), and J. Harrill,
Unpublished

Retrofitting NAMs for Metabolic Competence

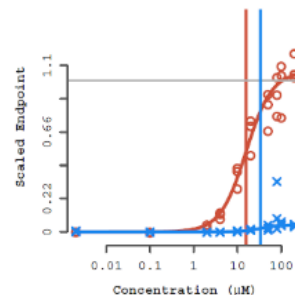
AIME Metabolism Assay



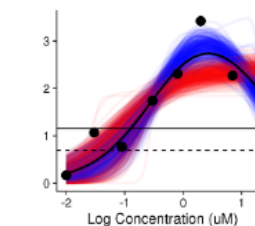
Estrogen Receptor Transactivation Assay



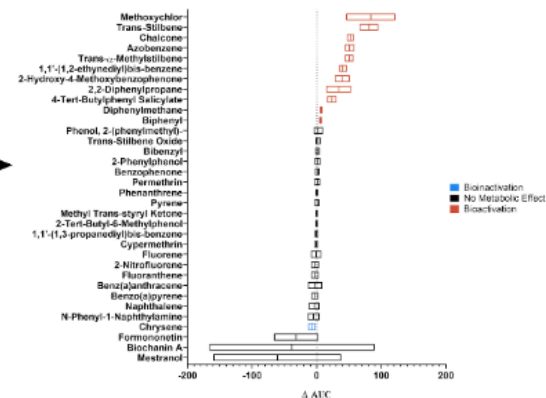
ToxCast Pipeline



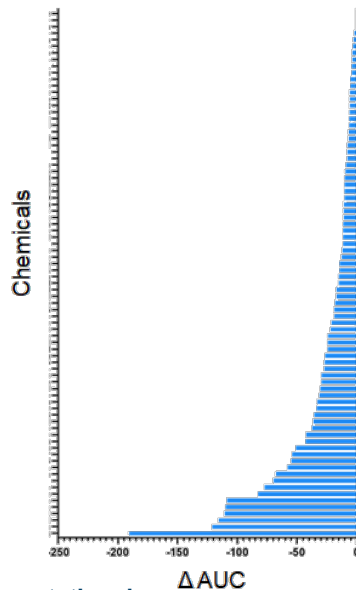
Toxboot Uncertainty Quantification



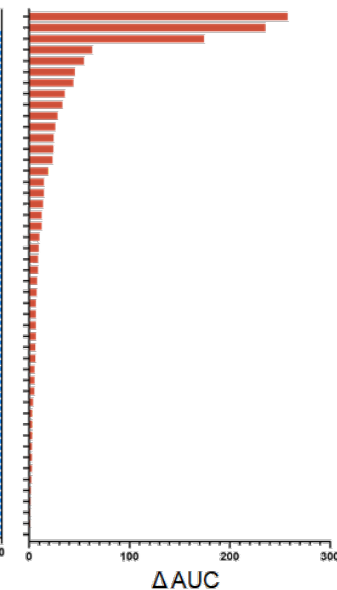
Rank Metabolism- dependent Bioactivity



Bioactivation



Bioactivation

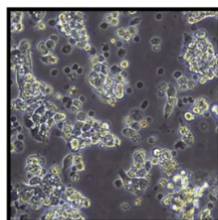


Preliminary Analysis of 768 ToxCast Chemical Screen

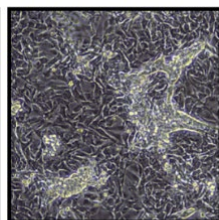
- Application of Deisenroth et al., *Toxicol Sci.*, 2020
- 11% of chemicals exhibit metabolism-dependent changes in ER bioactivity. Most are estrogenic \pm metabolism.
- False positive and false negative chemicals represent 3.6% of total chemicals screened.
- Profiles of predicted routes of biotransformation and potential metabolites.

Developing Complex Organotypic Culture Models to Evaluate Tissue/Organ Effects

Normal Human
Thyroid Gland



Harvest Follicle
Fragments



Attachment and
Outgrowth of Cells

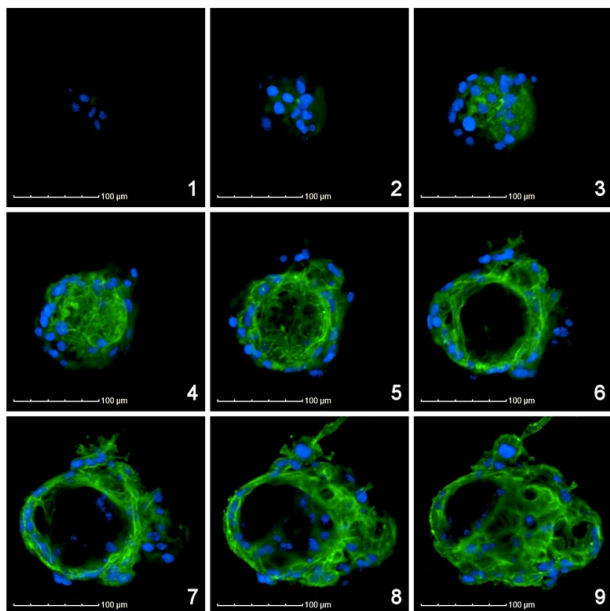
2D Cell Expansion



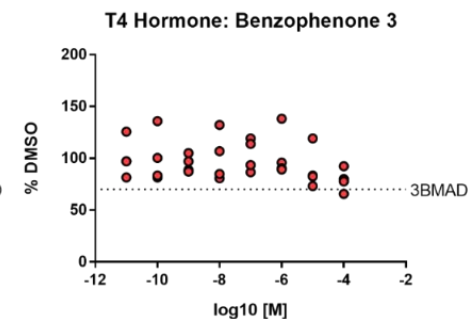
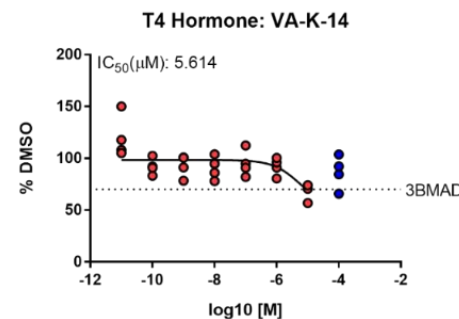
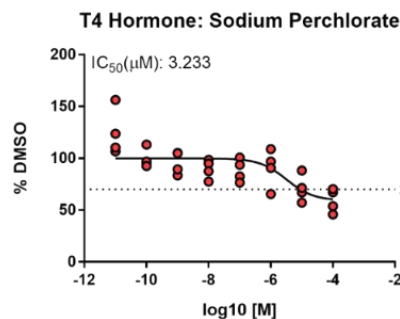
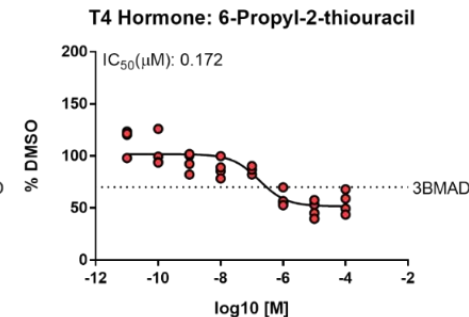
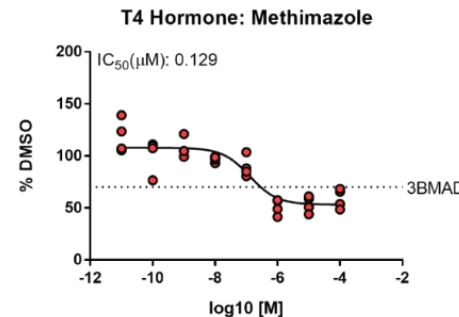
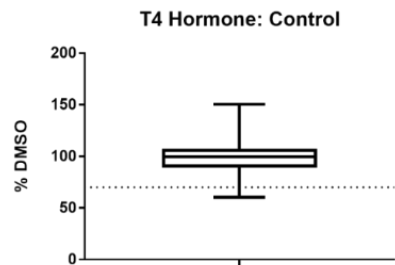
2D Monolayer
Culture

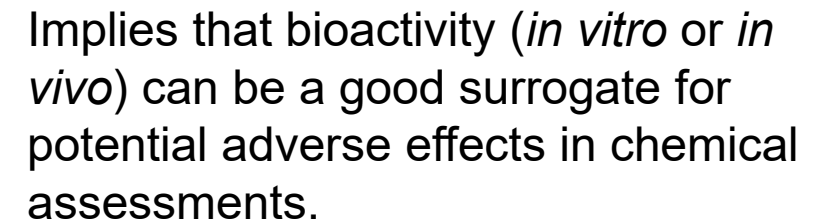


3D Sandwich
Culture



Blue, Hoechst 33342 /DNA
Green, Phalloidin/Actin





**Center for Computational
Toxicology & Exposure**

Step 3: Assemble NAMs into a Practical Testing Framework

TOXICOLOGICAL SCIENCES, 169(2), 2019, 317-332
doi: 10.1093/toxsci/kfz058
Advance Access Publication Date: March 5, 2019
Forum

SOT | Society of Toxicology
www.toxsci.oxfordjournals.org

FORUM
The Next Generation Blueprint of Computational Toxicology at the U.S. Environmental Protection Agency

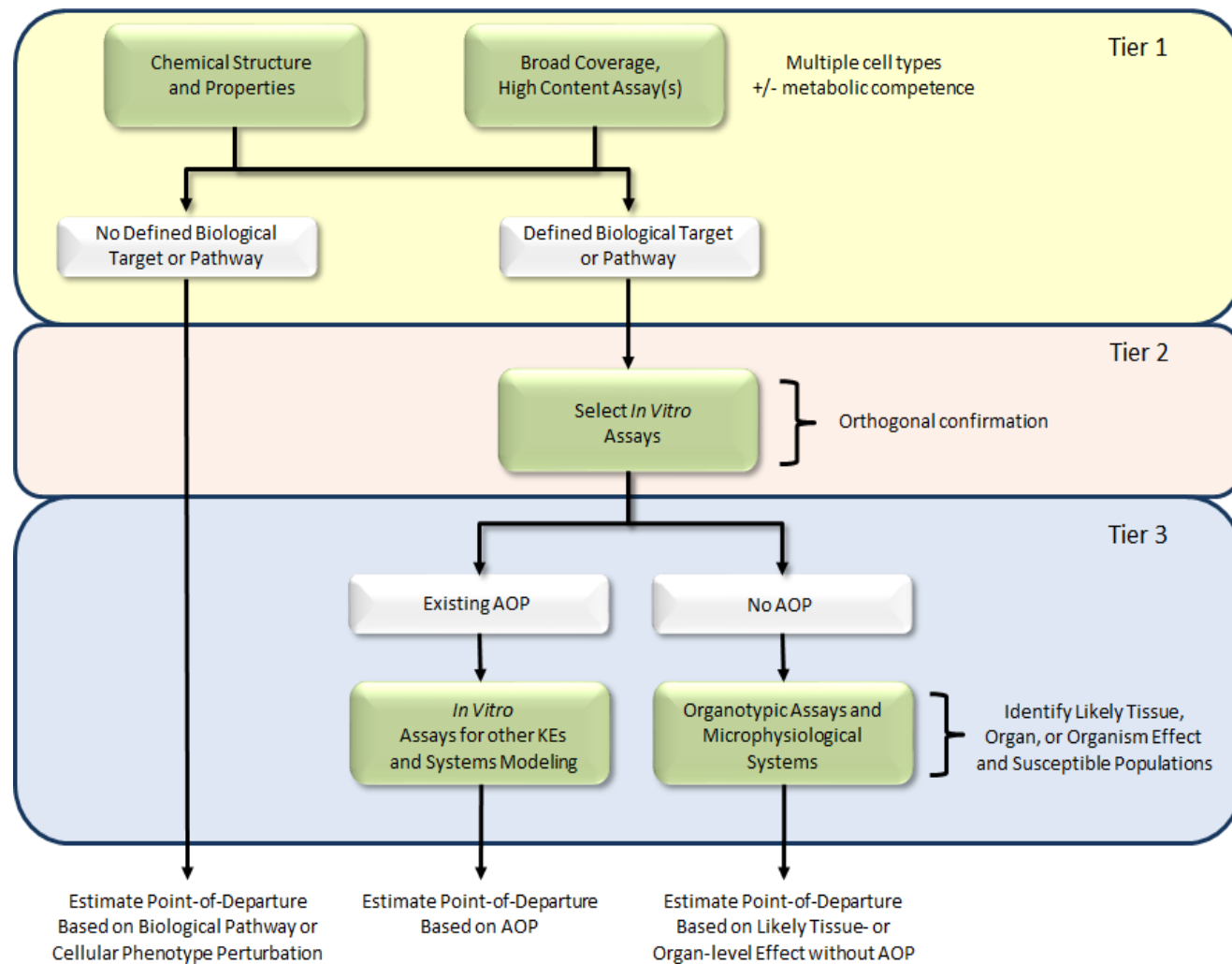
Russell S. Thomas,^{*,1} Tina Bahadori,[†] Timothy J. Buckley,[‡] John Cowden,^{*} Chad Deisenroth,^{*} Kathie L. Dionisio,[‡] Jeffrey B. Frithsen,[§] Christopher M. Grulke,^{*} Maureen R. Gwinn,^{*} Joshua A. Harrill,^{*} Mark Higuchi,[¶] Keith A. Houck,^{*} Michael F. Hughes,[¶] E. Sidney Hunter, III,[¶] Kristin K. Isaacs,[‡] Richard S. Judson,^{*} Thomas B. Knudsen,^{*} Jason C. Lambert,[¶] Monica Linnenbrink,^{*} Todd M. Martin,^{||} Seth R. Newton,[‡] Stephanie Padilla,[¶] Grace Patlewicz,^{*} Katie Paul-Friedman,^{*} Katherine A. Phillips,^{*} Ann M. Richard,^{*} Reeder Sams,^{*} Timothy J. Shafer,[¶] R. Woodrow Setzer,^{*} Imran Shah,^{*} Jane E. Simmons,[¶] Steven O. Simmons,^{*} Amar Singh,^{*} Jon R. Sobus,[‡] Mark Strynar,[‡] Adam Swank,[‡] Rogelio Tornero-Valez,[‡] Elin M. Ulrich,[‡] Daniel L. Villeneuve,^{||} John F. Wambaugh,^{*} Barbara A. Wetmore,[‡] and Antony J. Williams^{*}

^{*}National Center for Computational Toxicology, U.S. Environmental Protection Agency, Research Triangle Park, NC 27711, [†]National Center for Environmental Assessment, U.S. Environmental Protection Agency, Washington, D.C. 20004, [‡]National Exposure Research Laboratory, U.S. Environmental Protection Agency, Research Triangle Park, NC 27711, [§]Chemical Safety for Sustainability National Research Program, U.S. Environmental Protection Agency, Washington, D.C. 20004, [¶]National Health and Environmental Effects Research Laboratory, U.S. Environmental Protection Agency, Research Triangle Park, NC 27711, ^{||}National Center for Environmental Assessment, U.S. Environmental Protection Agency, Cincinnati, OH 45220, [¶]National Risk Management Research Laboratory, U.S. Environmental Protection Agency, Cincinnati, OH 45220, and ^{||}National Health and Environmental Effects Research Laboratory, U.S. Environmental Protection Agency, Duluth, MN 55804

[†]To whom correspondence should be addressed at National Center for Computational Toxicology, Office of Research and Development, U.S. Environmental Protection Agency, 209 T.W. Alexander Drive, Room D110-D, Mail Code: D145-02, Research Triangle Park, NC 27711. Fax: (919) 541-1594. E-mail: thomas.russell@epa.gov

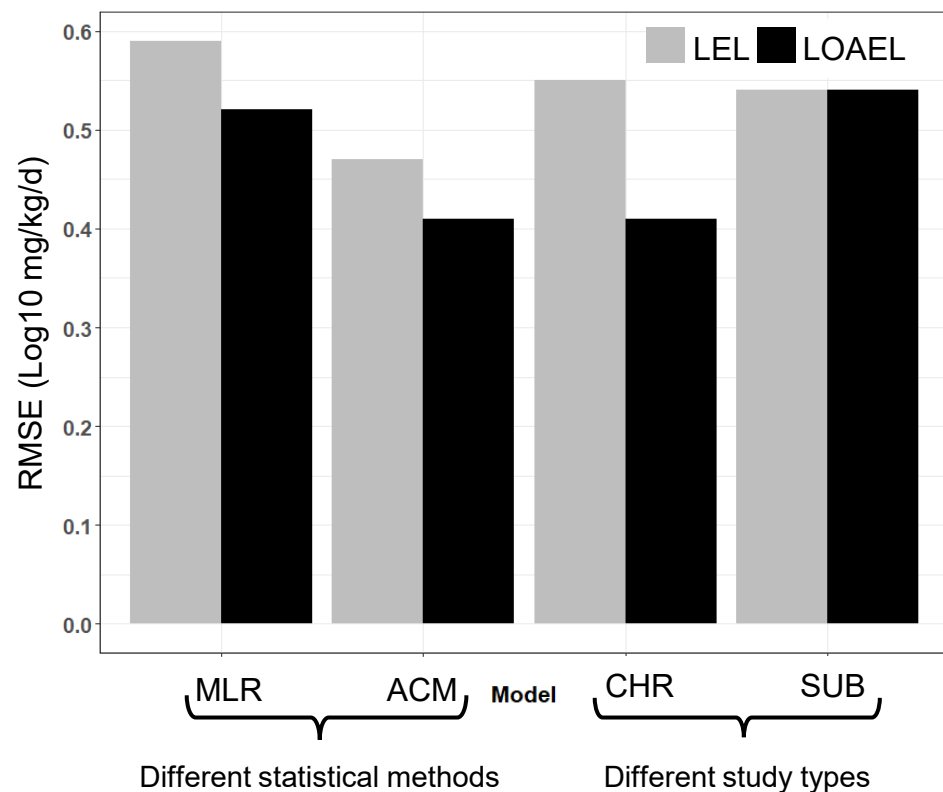
ABSTRACT
The U.S. Environmental Protection Agency (EPA) is faced with the challenge of efficiently and credibly evaluating chemical safety often with limited or no available toxicity data. The expanding number of chemicals found in commerce and the environment, coupled with time and resource requirements for traditional toxicity testing and exposure characterization, necessitates the development of new approaches to chemical safety evaluation. This work is written by US Government employees and is in the public domain in the US.

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Step 4: Understand How to Benchmark Approaches

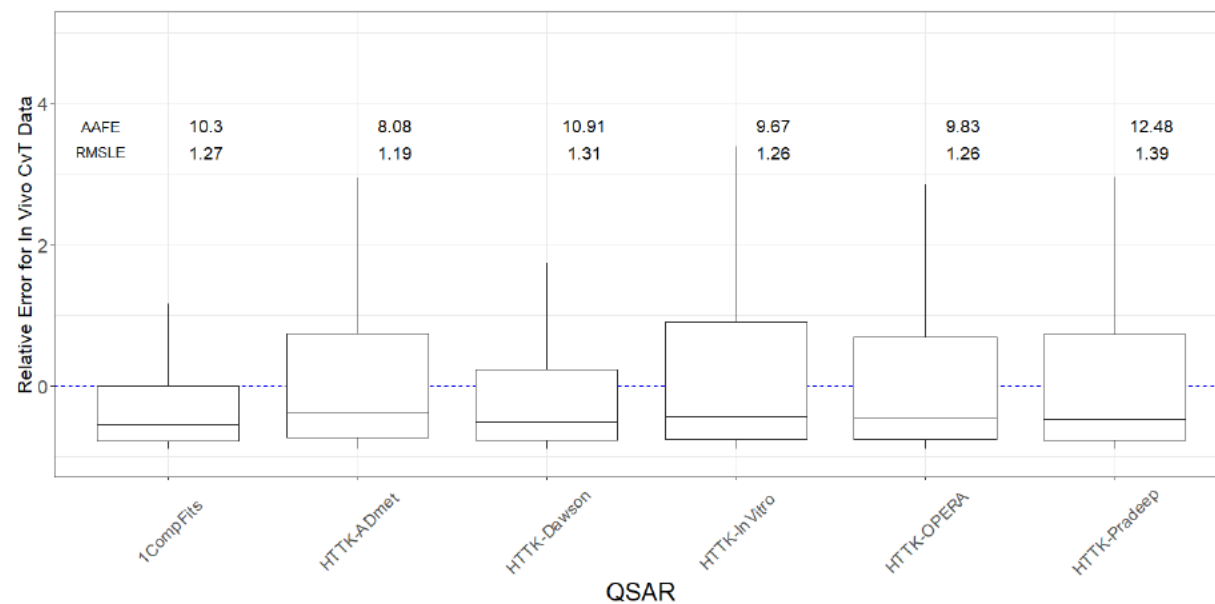
Evaluating LEL/LOAEL Variability in Traditional Toxicity Studies to Set Expectations for NAMs



Using an RMSE=0.59, the 95% Prediction Interval of an LEL/LOAEL is +/- 10-fold (e.g., 1 mg/kg/day, 0.07 – 14)

Pham et al., Comp Toxicol., 2020

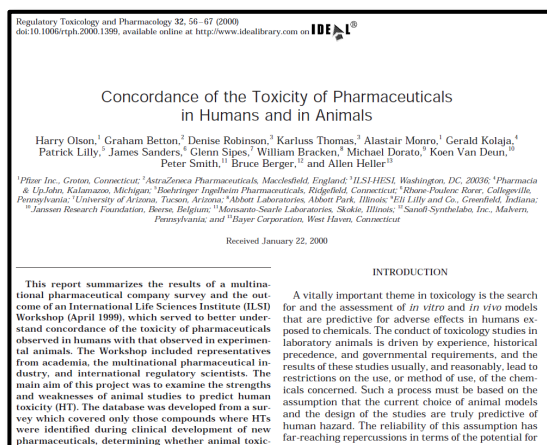
Comparing *In Silico*, *In Vitro*, and *In Vivo* Data for Toxicokinetic Modeling



Wambaugh et al., QSAR2021 meeting poster

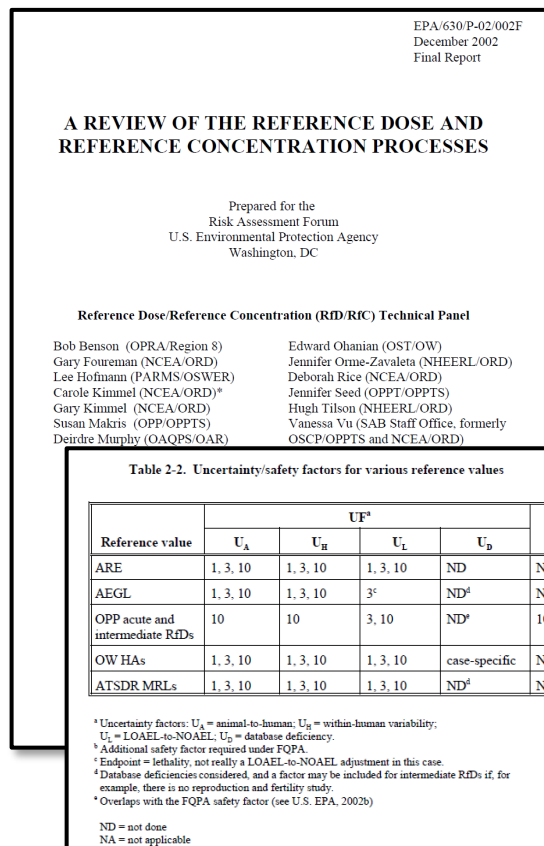
Step 5: Grapple With the Issue of Protection vs Prediction with Current Models and NAMs

Limited Qualitative Concordance of Rodent and Human Toxicological Responses

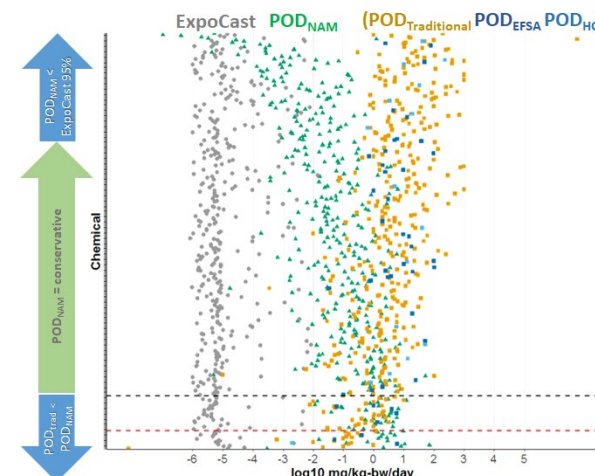


...data compiled from 150 compounds with 221 human toxicity events reported. The results showed the true positive human toxicity concordance rate of 71% for rodent and non-rodent species, with non-rodents alone being predictive for 63% of human toxicity and **rodents alone for 43%.**

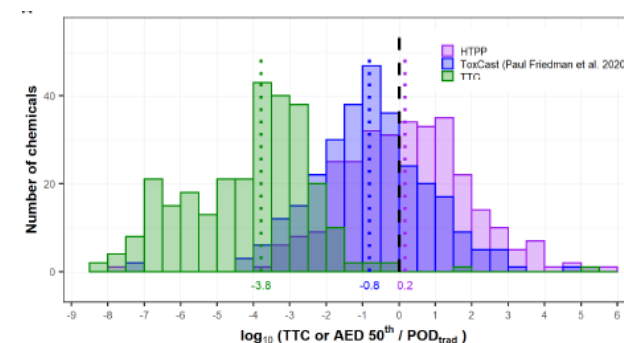
Current Risk Assessment Practices Geared Towards Protection Not Prediction



Case Studies Demonstrating Application of Bioactivity as a Protective POD



Paul-Friedman et al., 2020



Nyffeler and Harrill, ISMB Poster, 2020

Step 6: Evaluate Regulatory Flexibilities and Develop a Fit-for-Purpose Scientific Confidence Framework



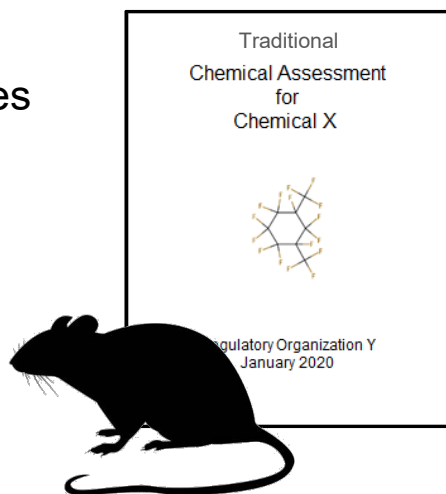
Deliverables:

- EPA review of existing statutes, regulations, policies, and guidance that relate to vertebrate animal testing in 2022
- US National Academies of Sciences report on variability and relevance of existing mammalian toxicity tests in 2023.
- Scientific confidence framework to evaluate the quality, reliability, and relevance of NAMs in 2024.

Step 7: Quantify Trade-Offs of Uncertainty, Cost, and Time in Toxicity Testing Methods

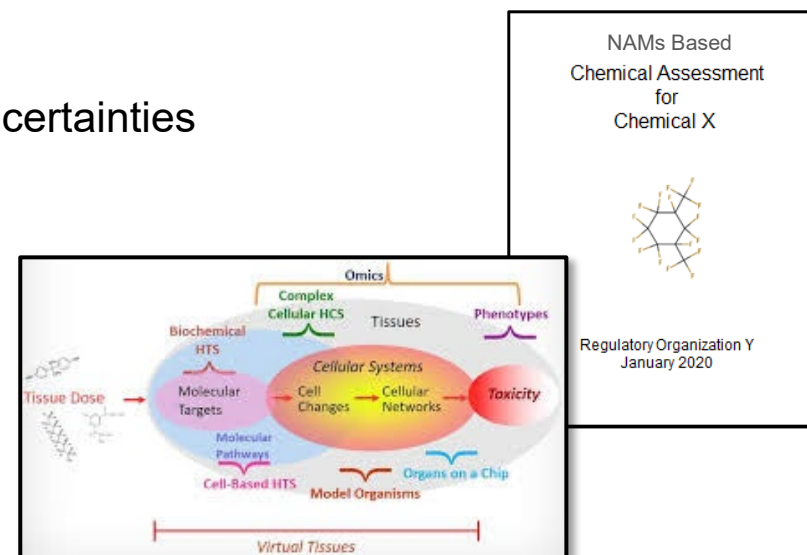
Option 1

- 6 – 20 years
- “Smaller” uncertainties
- \$Ks - \$Ms



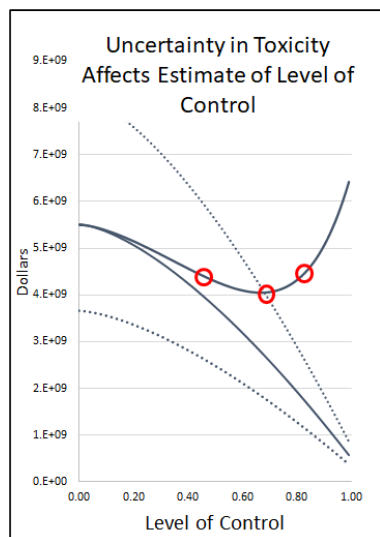
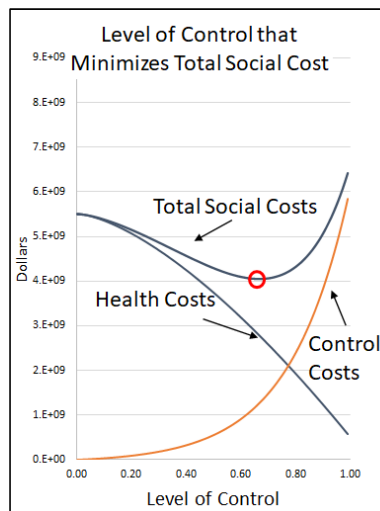
Option 2

- <1 year
- “Bigger” uncertainties
- \$Ks



What choice would you make?

Development of a Value of Information Framework to Evaluate the Trade-Offs in Toxicity Testing



- Value of information (VOI) analysis is a decision analytic method that quantifies the expected value of additional testing/data in reducing decision uncertainty (Tuffaha, 2021).
- VOI requires a method to determine the cost of uncertainty
 - $Total\ Social\ Cost = Total\ Control\ Cost + Total\ Health\ Cost$
- Lots of work in VOI evaluating different tests (e.g., medical tests), but few studies evaluating the impact of time.
- The impact of time can be incorporated by discounting the costs on an annual basis.
- Multiple metrics can be used to compare the value of different toxicity tests adjusted for time and cost of the test
 - **Expected Value of Delayed Sample Information (EVDSI)**
 - Expected Net Benefit of Sampling (ENBS)
 - Return on Investment (ROI)

General Conclusions From the Value of Information Studies

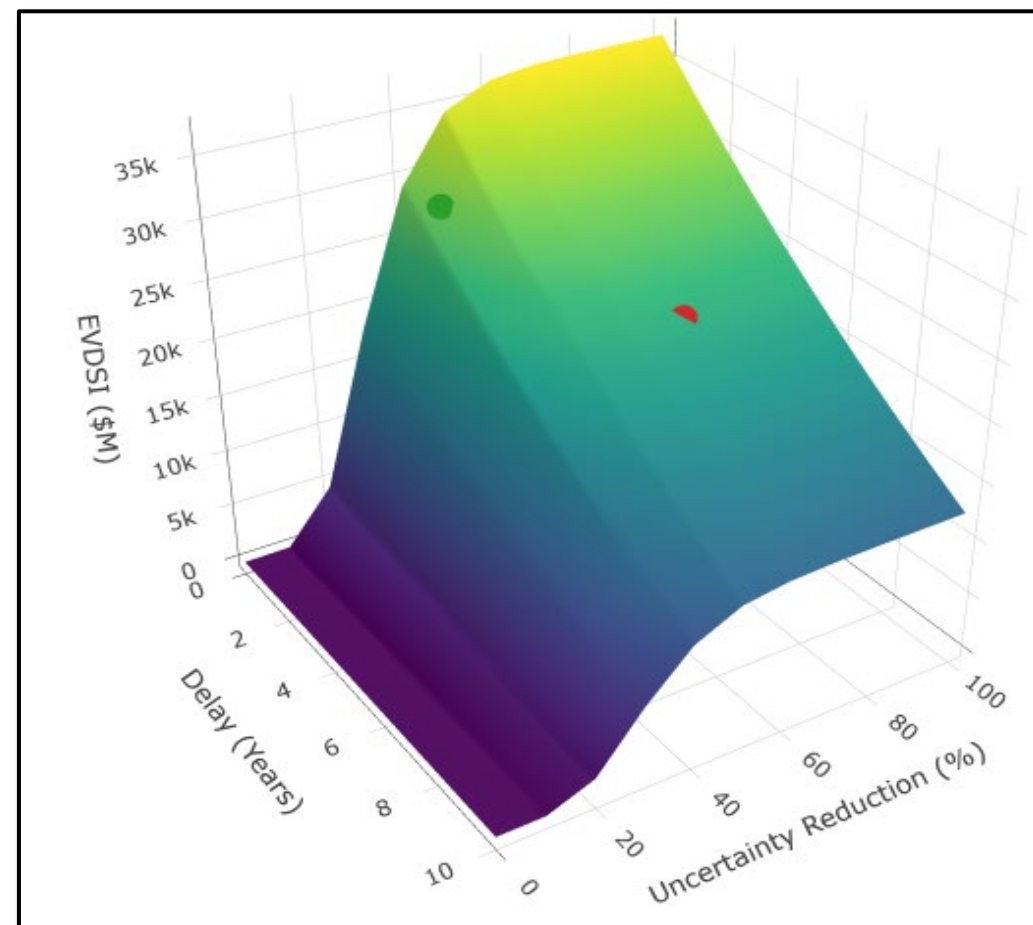
Example Scenarios

- Two hypothetical toxicity tests
 - Test A – lower cost (\$5K), shorter duration (1 yr), higher uncertainty (4 orders of magnitude)
 - Test B – higher cost (\$5M), longer duration (5 yr), lower uncertainty (2 orders of magnitude)
- Different health endpoints and decision types
 - Chronic and acute effects
 - Chemicals regulated based on benefit-cost analysis and target risk levels

Overall Conclusions

- ***Timeliness has a significant positive impact on the VOI of toxicity tests, even in the presence of smaller reductions in uncertainty.***
- The positive impact of the shorter tests may be multiplicatively amplified by the ability to test more chemicals.

Trade-Offs of Uncertainty and Time of Hypothetical Toxicity Testing Methods
(Chronic Effect, Target Risk Decision Maker)

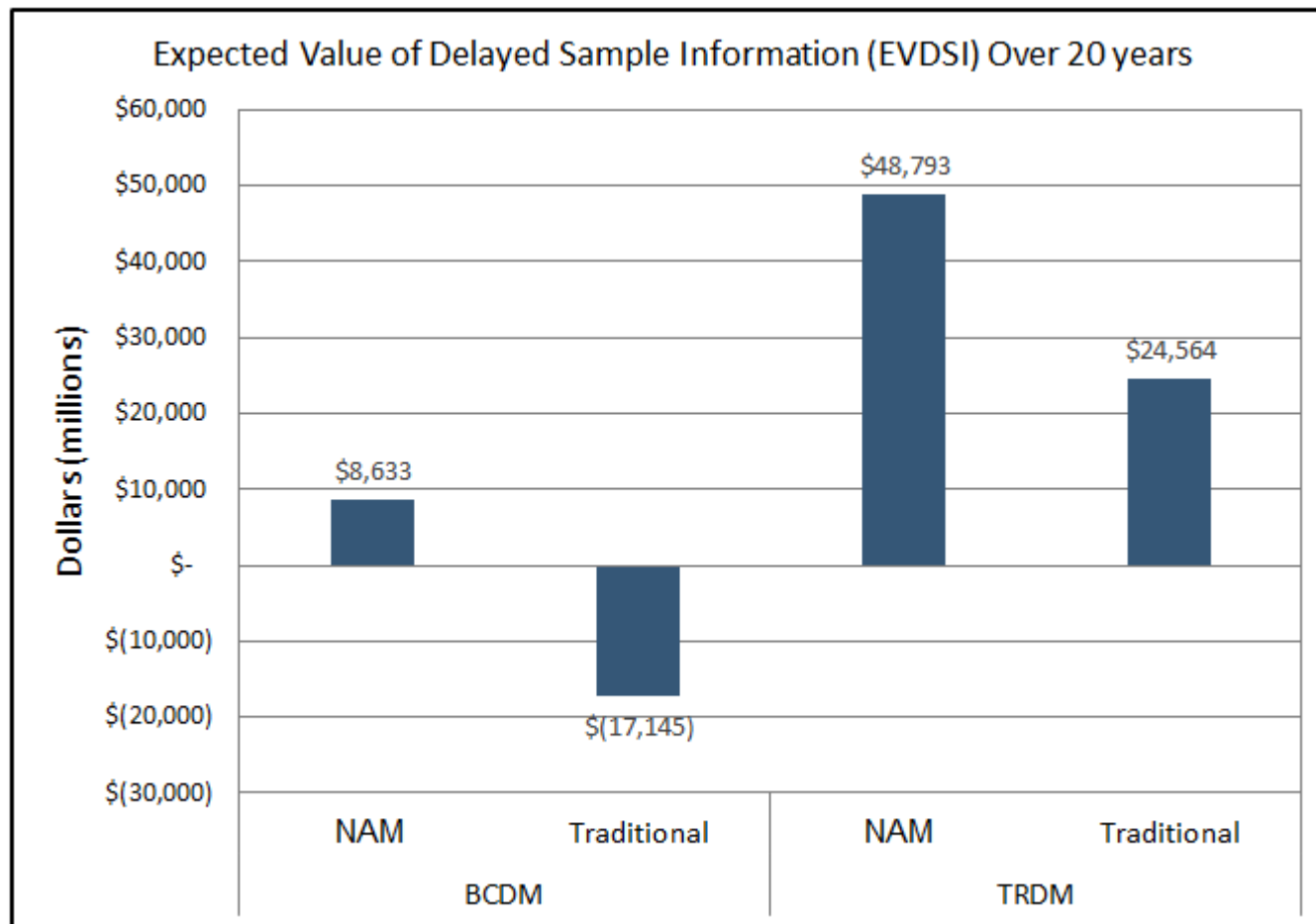


Focused Case Study on Di (2-Ethylhexyl)Phthalate

Parameters

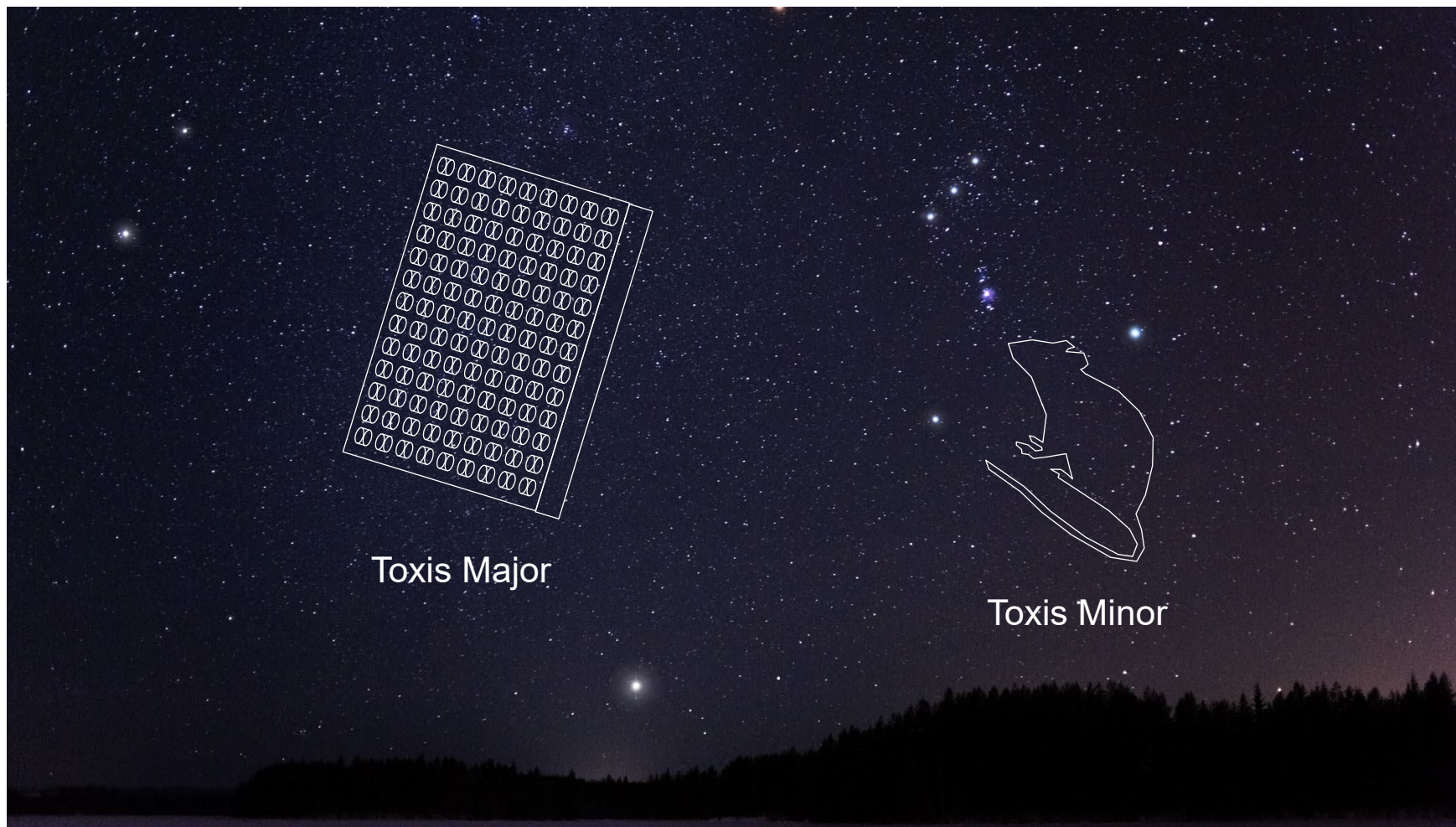
		Values
Toxicity (Prior)		
	$\text{Log}_{10} \mu_{\text{tox}}$	1.699
Uncertainty in μ_{tox} (Orders of Magnitude or OM)	7	
	$\text{Log}_{10} (\sigma_{\text{tox}})$	0.697
Exposure		
	$\text{Log}_{10}(\mu_{\text{exp}})$	-2.87
Uncertainty in μ_{exp} (OM)	1.74	
	$\text{Log}_{10}(\sigma_{\text{exp}})$	0.34
Toxicity Post Test A – NAM		
Uncertainty in measured μ_{tox} (OM)	2.74	
Delay	1 yrs	
Cost	\$50,000	
Toxicity Post Test B – Traditional		
Uncertainty in measured μ_{tox} (OM)	1.76	
Delay	8 yrs	
Cost	\$5,000,000	

- Exposure estimates derived from NHANES biomonitoring data for U.S. adults (Reyes and Price, 2019)
- Toxicity estimates from published standards using methods from Chiu and Slob (2015)
- Evaluated Benefit Cost Decision Maker (BCDM) and Target Risk Decision Maker (TRDM)
- Chronic health effects (mortality)



P. Price, Preliminary Data

Moving from a Paradox to a Practical Solution



Acknowledgements

Center for Computational Toxicology and Exposure (CCTE) Staff

Tox21 Colleagues:

NTP
FDA
NCATS

EPA Colleagues:

CEMM
CPHEA
CESER

Collaborative Partners:

Unilever
A*STAR
ECHA
EFSA
Health Canada



Research Triangle Park, NC



Cincinnati, OH



Duluth, MN



Washington, DC



Athens, GA



Gulf Breeze, FL