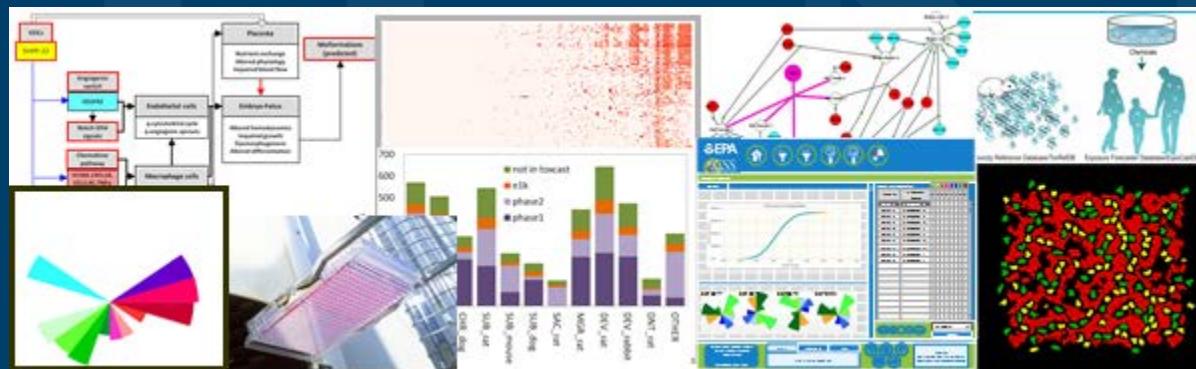


Analysis of Transcriptomic Dose Response Data in the Context of Chemical Risk Assessment

My Journey Through the Toxicogenomics Bermuda Triangle



HESI-Health Canada-McGill Workshop
May 26, 2017

Russell Thomas
Director
National Center for Computational Toxicology

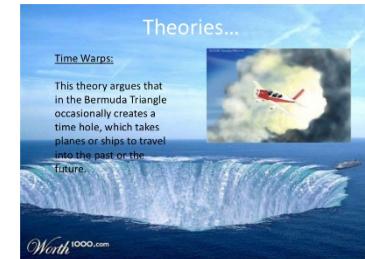
Myth of the Real Bermuda Triangle



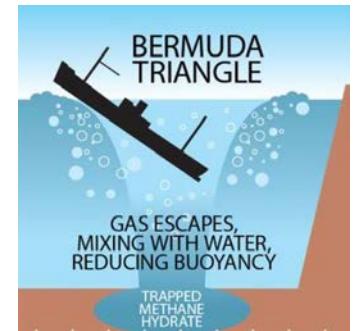
<http://onedio.co/content/new-mind-blowing-claim-about-the-bermuda-triangle-13519>



http://www.crystalinks.com/bermuda_triangle.html

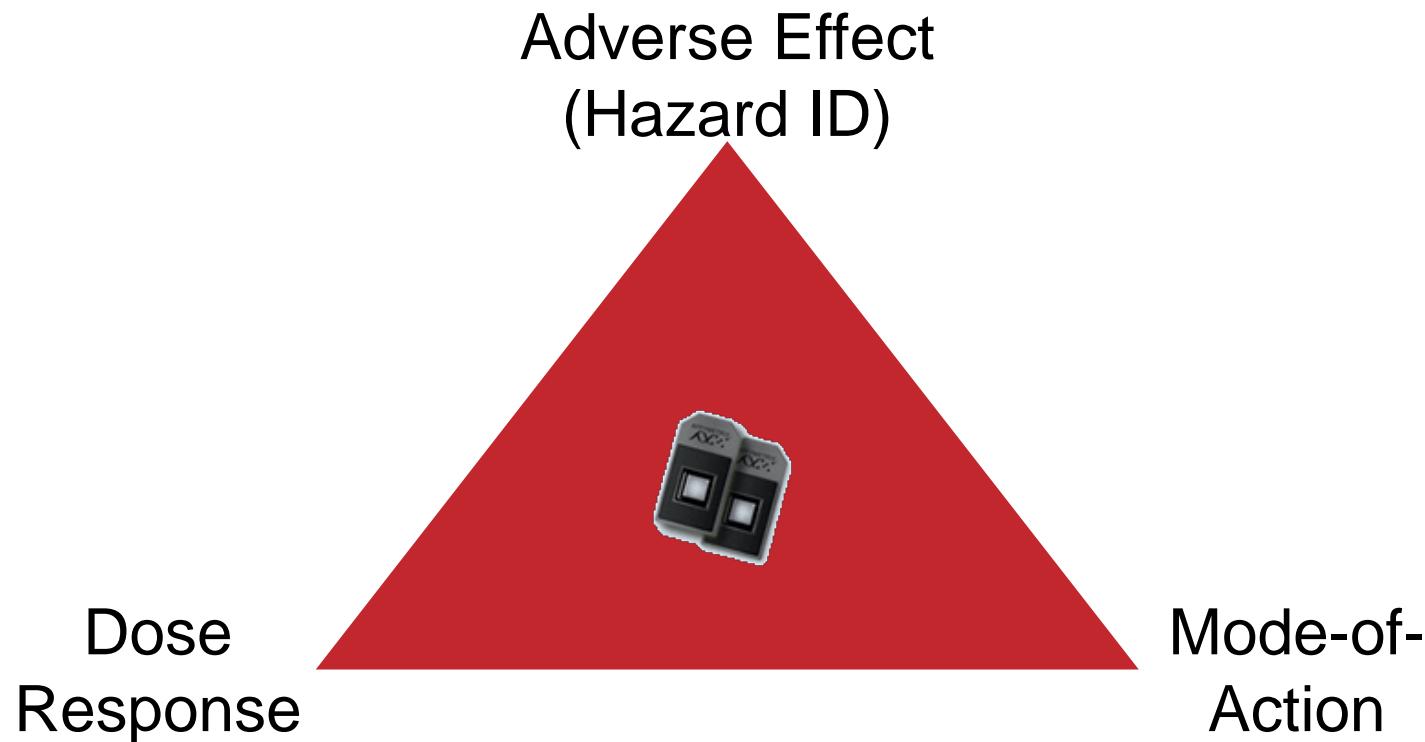


<https://www.slideshare.net/jayakanthan75/bermuda-triangle-ppt>

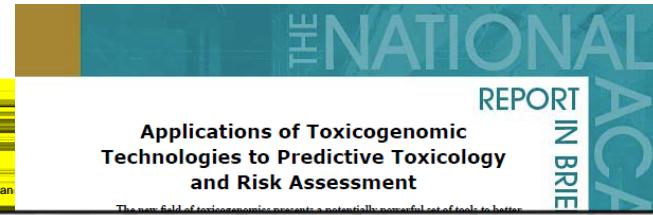


<http://www.mensxp.com/special-features/today/29862-the-mystery-behind-the-disappearances-in-the-bermuda-triangle-may-just-have-been-solved.html>

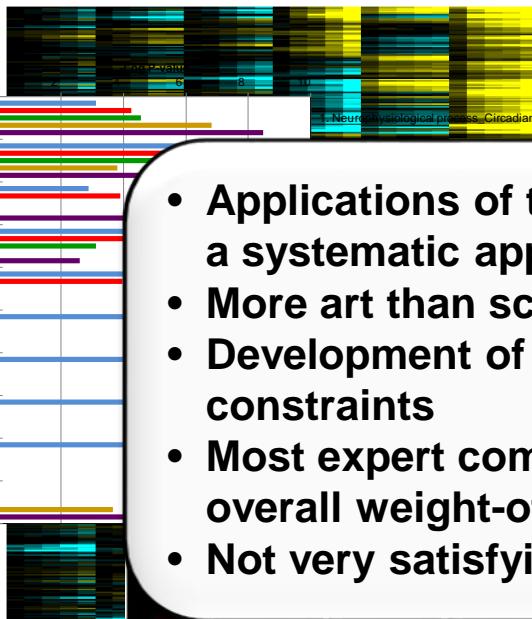
Bermuda Triangle of Toxicogenomics and Hazard Assessment



Initial Promise of Toxicogenomics Focused on Inferring MOA



- Applications of toxicogenomics to identify MOA have generally lacked a systematic approach
- More art than science
- Development of large reference databases difficult due to cost constraints
- Most expert committees/reports defaulted to using it as part of an overall weight-of-evidence
- Not very satisfying



uses of existing data sources, and study data in new ways, perhaps on a scale approaching that of the Human Genome Project. Toxicogenomics also raises some ethical issues encoded by genes.

Metabolomics is the study of the products of biological processes. Such products change in response to such things as nutrition, stress, and disease states.

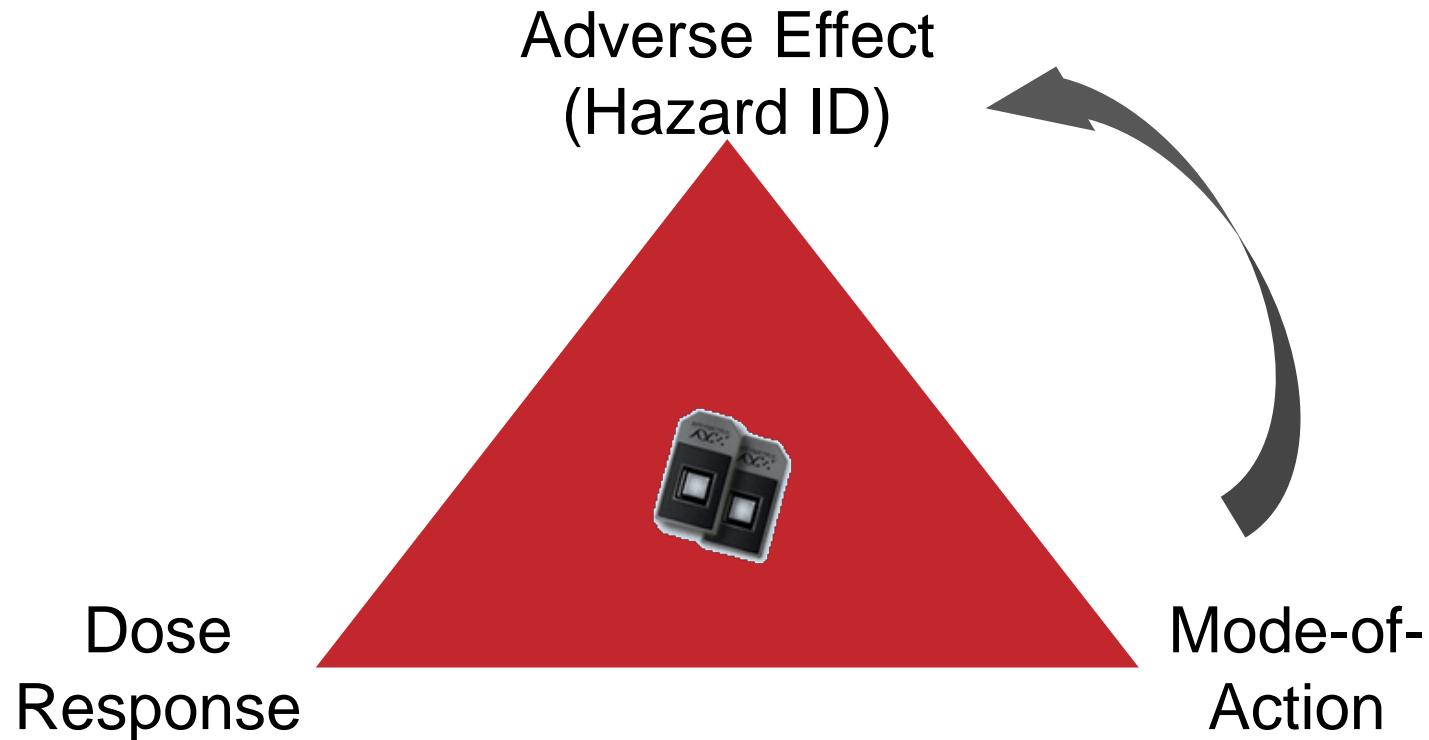
THE NATIONAL ACADEMIES
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MOA Generator



Bermuda Triangle of Toxicogenomics and Hazard Assessment



Developing Statistical Classification Models to Predict Hazard

The MicroArray Quality Control (MAQC)-II study of common practices for the development and validation of microarray-based predictive models

MAQC Consortium*

Gene expression data from most of these predictions has not been used to generate predictive models for rodent carcinogenesis, or of breast cancer. Combinations of analytical methods and endpoints and, to minimize performance depended largely on the performance. The conclusions and independent investigation.

- Most studies show 60-85% accuracy for predicting cancer-related endpoints
- Only a limited number of tissues have been evaluated
- Requires >20 compounds with adequate redundancy and diversity in mechanisms to have a robust training set (Thomas et al., 2009)
- >30 organs show tumor responses in NTP database with ~50% having >10 chemicals in at least one species/sex
- Difficult to justify as a comprehensive screen for rodent carcinogenicity

TOXICOLOGICAL SCIENCES 124(1), 54–74 (2011)
doi:10.1093/toxsci/kfr202
Advance Access publication August 2, 2011

Development and Evaluation of a Genomic Signature for the Prediction of Carcinogenicity in Mice Using Rat Hepatoma Cells

The Journal of

Vol.37, No.1

Drug

Toxicology

Advances

Discrimination for Genotoxic and Nongenotoxic Carcinogens by Gene Expression Profiling in Primary Mouse Hepatocytes Improves the Prediction of Carcinogenic Potential

Karen Mathijssen,*† Karen J. J. Brauers,* Danyel G. J. Jennen,*† Andre Boorsma,*‡ Marcel H. M. van Herk,* Ralph W. H. Gottschalk,* Jos C. S. Kleinjans,*† and Joost H. M. van Delft*†‡

*Department of Health Risk Analysis and Toxicology, Faculty of Health, Medicine and Life Sciences, Maastricht University, 6200 MD Maastricht, The Netherlands; †Netherlands Toxicogenomics Center, 6200 MD Maastricht, The Netherlands; and ‡TNO Quality of Life, 3700 AJ Zeist, The Netherlands

Received June 12, 2009; accepted September 11, 2009

*The Hamner Institutes for Health Sciences, Research Triangle Park, North Carolina 27709; †SAS Institute Inc., Cary, North Carolina 27513; ‡Vavilov Institute of General Genetics, Moscow B333, 117809, Russia; and §GeneGo, Inc., St Joseph, Michigan 49085

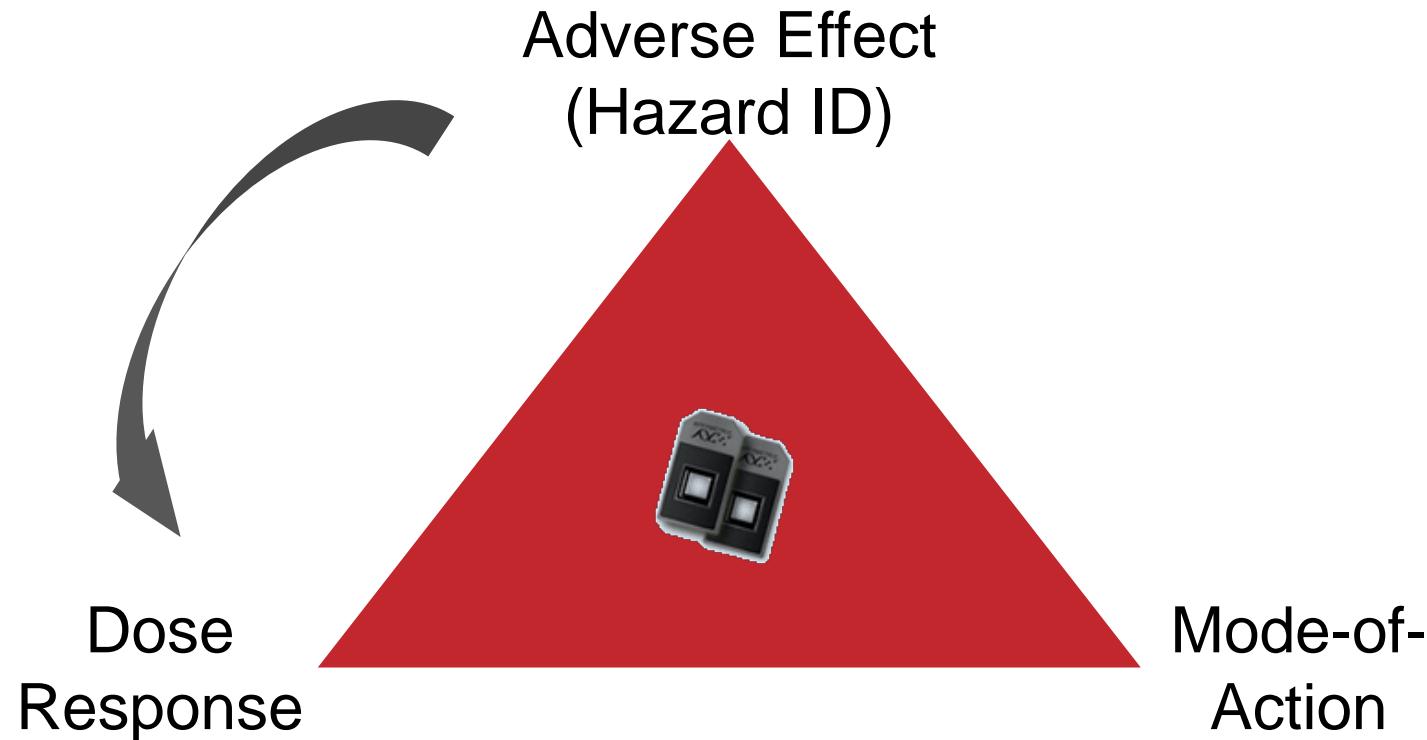
Prediction of carcinogenic potential by a toxicogenomic approach using rat hepatoma cells

Kazunari Tsujimura,^{1,2} Makoto Asamoto,^{1,3} Shugo Suzuki,¹ Naomi Hokaiwado,¹ Kumiko Ogawa¹ and Tomoyuki Shirai¹

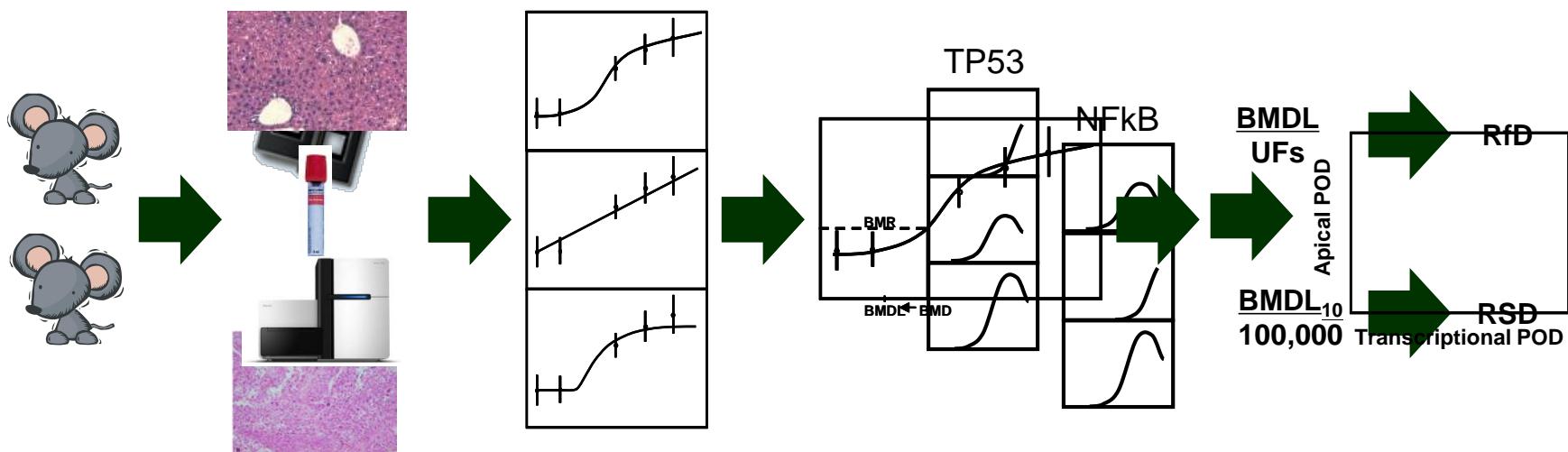
¹Department of Experimental Pathology and Tumor Biology, Nagoya City University, Graduate School of Medical Sciences, 1 Kawasumi, Mizuho-cho, Mizuho-ku Nagoya 467-8601, Japan; ²Chemicals Evaluation and Research Institute, Hita Laboratory, Tokyo 112-0004, Japan

(Received April 4, 2006/Revised June 1, 2006/Accepted June 8, 2006/Online publication August 11, 2006)

Bermuda Triangle of Toxicogenomics and Hazard Assessment



Using Transcriptomics to Define Dose Response



Subchronic Animal
Chronic Animal
Studies

Transcriptional
Responses
Considered
Adverse

Fit Data With
Statistical
Models

Identify Point Soles By
Departing (high BMD)
and Calculate
Summary Value
(usually the median
pathway BMD)

Dose-Response In Vivo
Exposures and
Transcriptional
PODs

Thomas *et al.*, *Tox Sci.*, 2011
Thomas *et al.*, *Mut Res.*, 2012
Thomas *et al.*, *Tox Sci.*, 2013

In Vivo Study to Assess Transcriptional and Apical Correlation

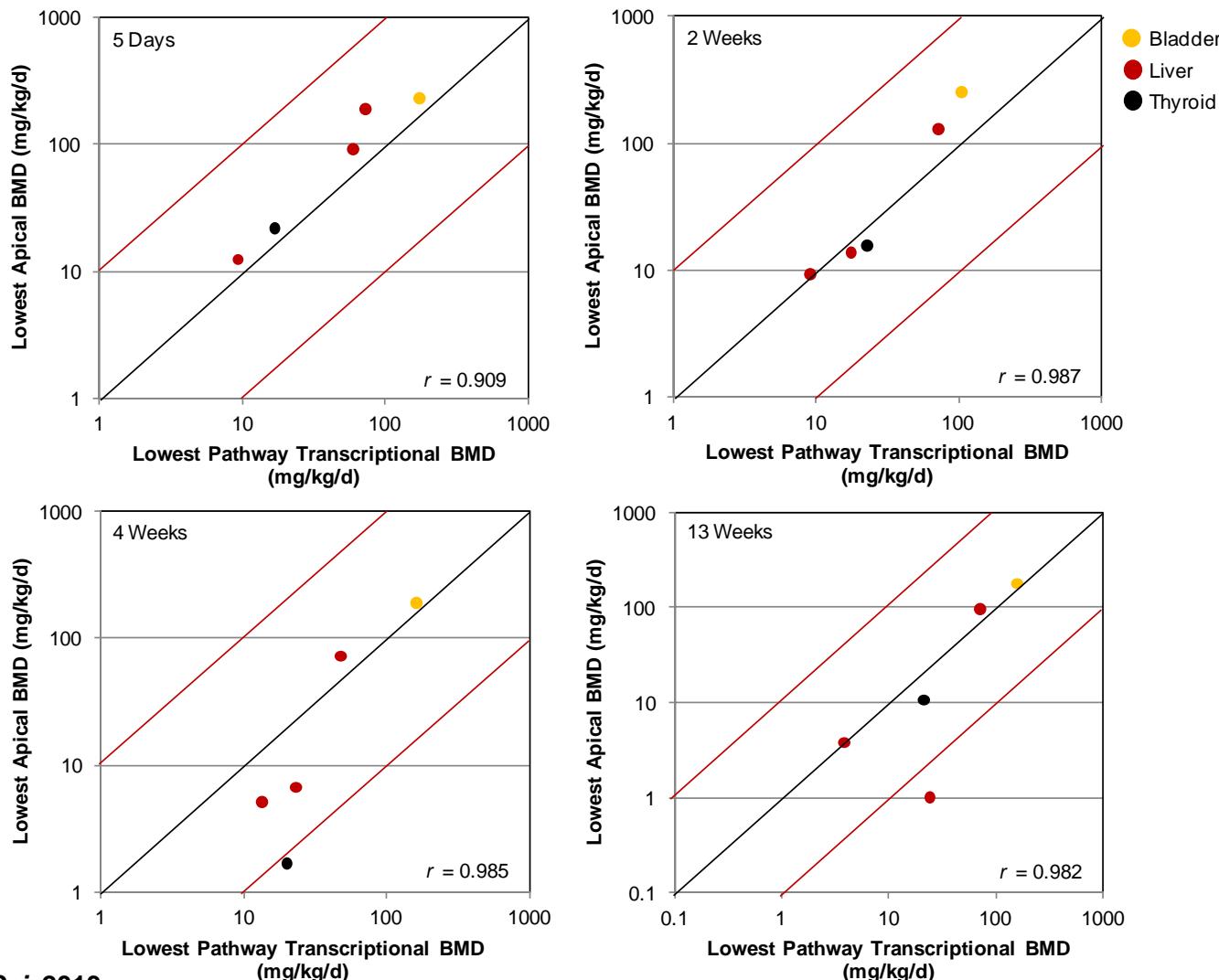
Chemical	Route	Doses ^c	Rodent Model	Time Point	Target Tissue
1,2,4-Tribromobenzene ^a	Gavage	2.5, 5, <u>10</u> , 25, 75 mg/kg	Male Sprague Dawley rats	5 d, 2, 4, 13 wks	Liver
Bromobenzene ^a	Gavage	25, (<u>50</u>), <u>100</u> , <u>200</u> , 300, <u>400</u> mg/kg	Male F344 rats	5 d, 2, 4, 13 wks	Liver
2,3,4,6-Tetrachlorophenol ^a	Gavage	10, <u>25</u> , 50, <u>100</u> , <u>200</u> mg/kg	Male Sprague Dawley rats	5 d, 2, 4, 13 wks	Liver
4,4'-Methylenebis (N,N-dimethyl) benzenamine ^b	Feed	50, 200, <u>375</u> , 500, <u>750</u> ppm	Male F344 rats	5 d, 2, 4, 13 wks	Thyroid ^b
N-Nitrosodiphenylamine ^b	Feed	<u>250</u> <u>1000</u> <u>2000</u> <u>3000</u> <u>4000</u> ppm	Female F344 rats	5 d, 2, 4, 13 wks	Bladder ^b
Hydroquinone ^b	Measured apical (histological and organ weight; n = 10) and gene expression changes (n = 5) at each dose and time point in the target tissue.				
1,4-Dichlorobenzene ^b	Inhalation	25, <u>75</u> , <u>300</u> , 800, <u>1200</u> ppm	Female B6C3F1 mice	13 wks	Liver
Propylene glycol mono-t-butyl ether ^b	Inhalation	25, <u>75</u> , <u>300</u> , 800, <u>1200</u> ppm	Female B6C3F1 mice	13 wks	Liver
1,2,3-Trichloropropane ^b	Gavage	2, <u>6</u> , <u>20</u> , 40, <u>60</u> mg/kg	Female B6C3F1 mice	13 wks	Liver
Methylene Chloride ^b	Inhalation	100, 500, <u>2000</u> , 3000, <u>4000</u> ppm	Female B6C3F1 mice	13 wks	Liver, Lung
Naphthalene ^b	Inhalation	0.5, 3, <u>10</u> , 20, <u>30</u> ppm	Female B6C3F1 mice	13 wks	Lung
1,4-Dichlorobenzene ^b	Gavage	100, <u>300</u> , 400, 500, <u>600</u> mg/kg	Female B6C3F1 mice	13 wks	Liver

^aChemicals in IRIS database for non-cancer endpoints only

^bChemicals previously tested by the U.S. National Toxicology Program

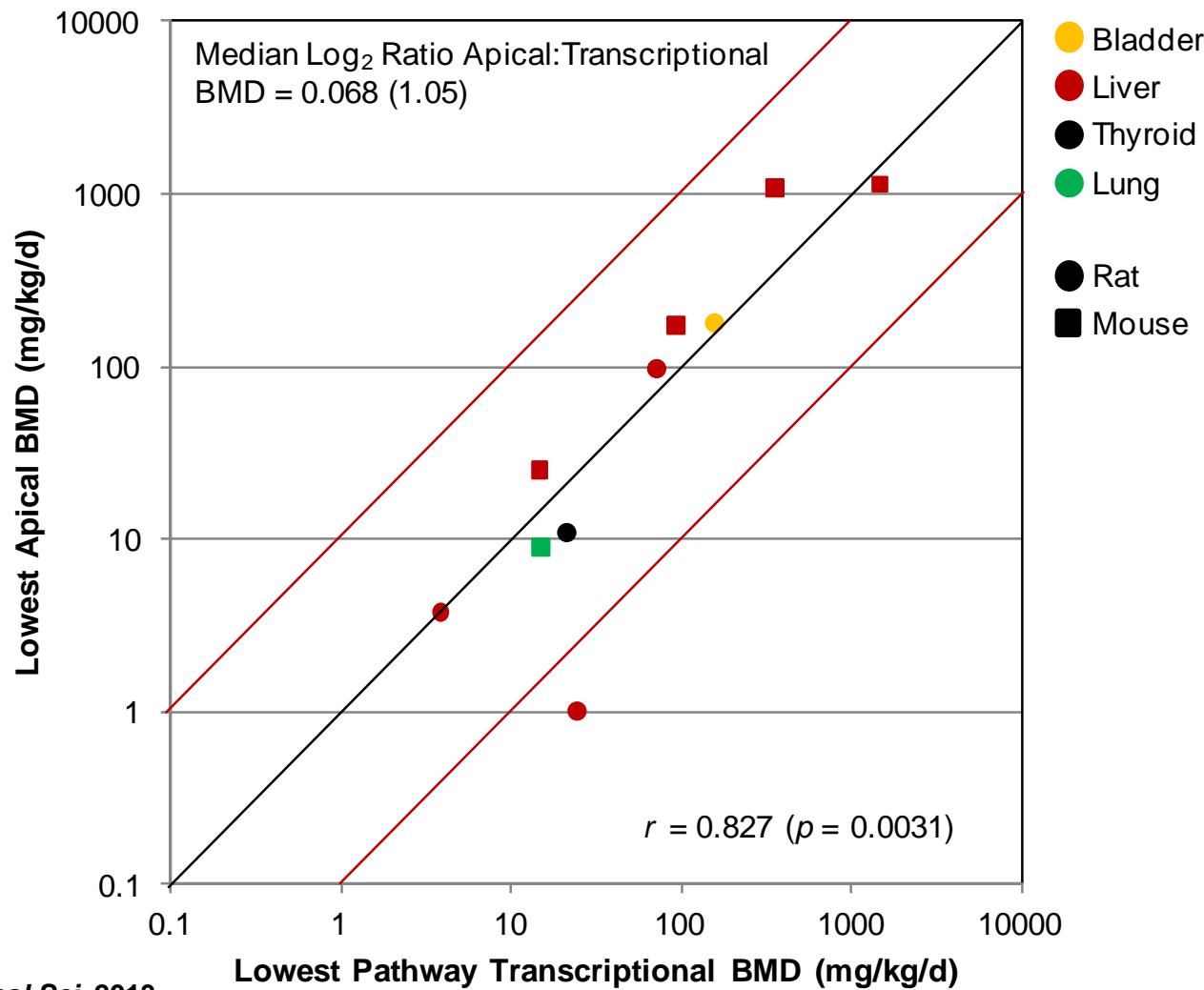
^cUnderlined doses used in NTP two-year rodent bioassay or IRIS database

Temporal Changes Between Transcriptional and Non-Cancer PODs



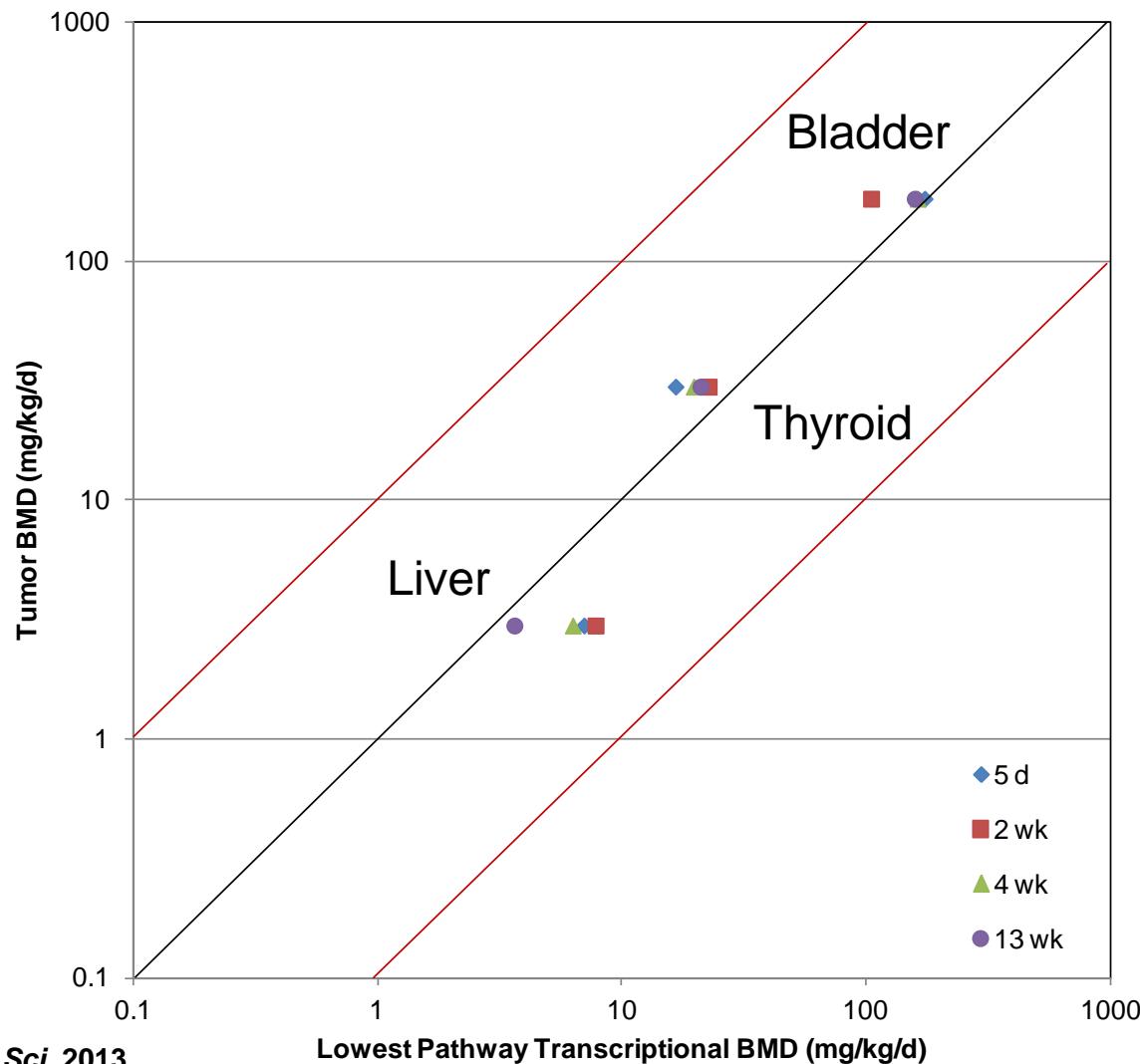
Thomas et al., *Toxicol Sci*, 2013

Combined Correlation Between Non-Cancer and Transcriptional PODs

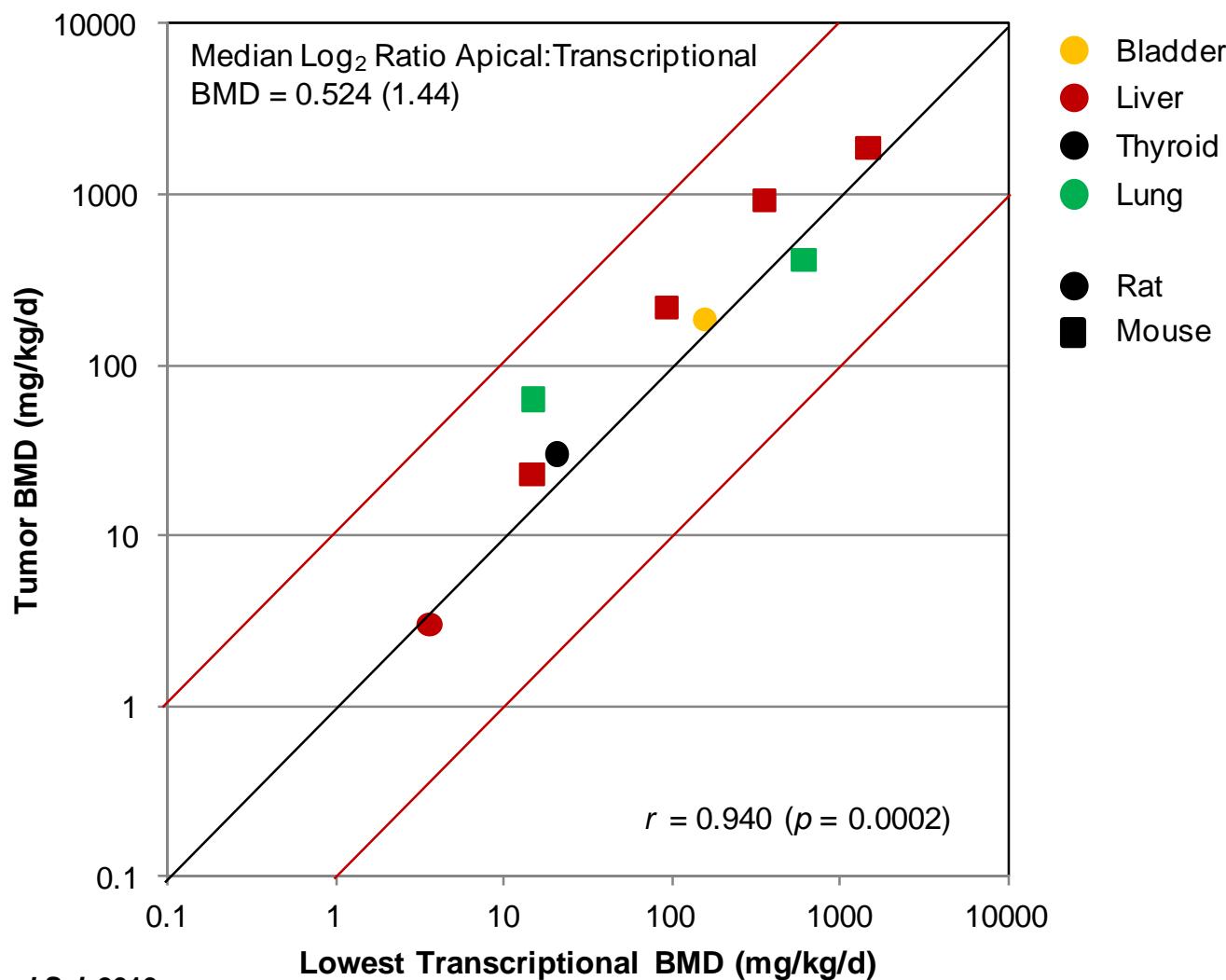


Thomas et al., *Toxicol Sci*, 2013

Temporal Changes Between Transcriptional and Cancer PODs

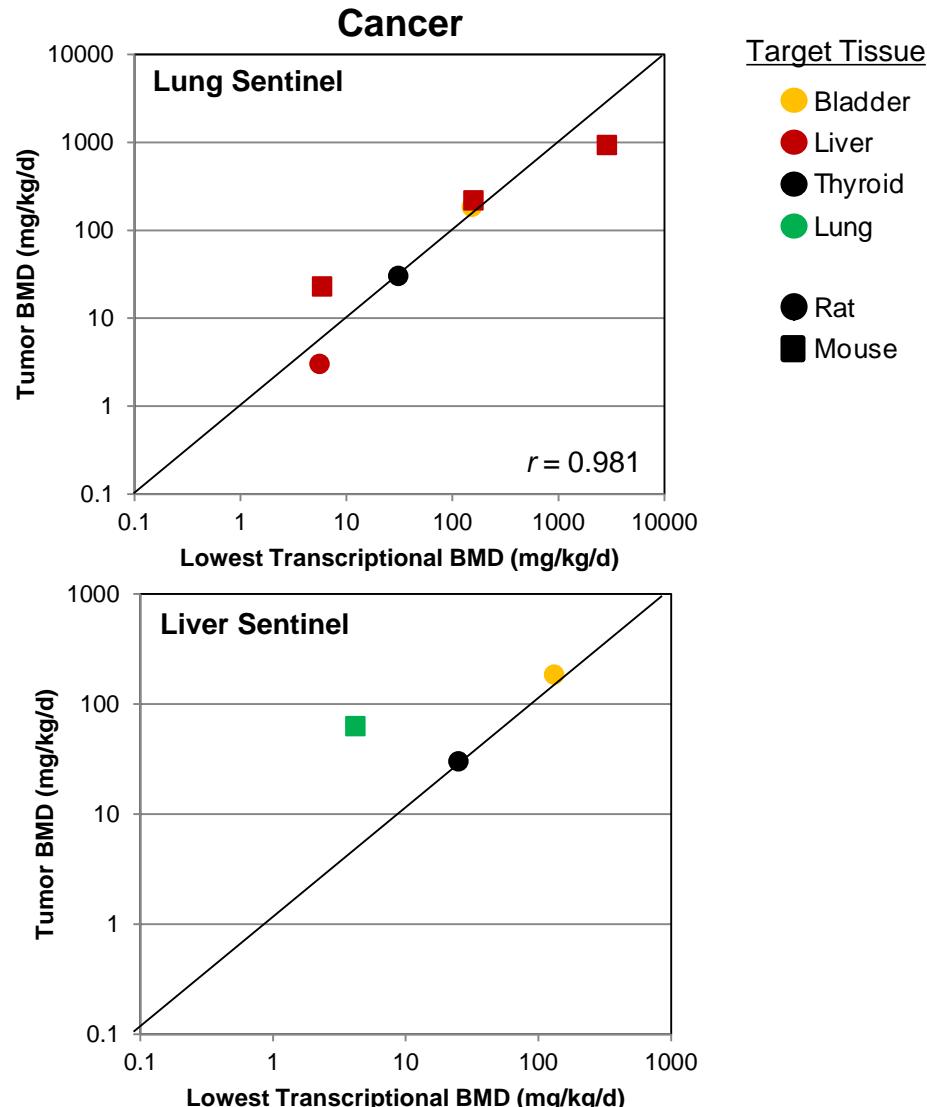
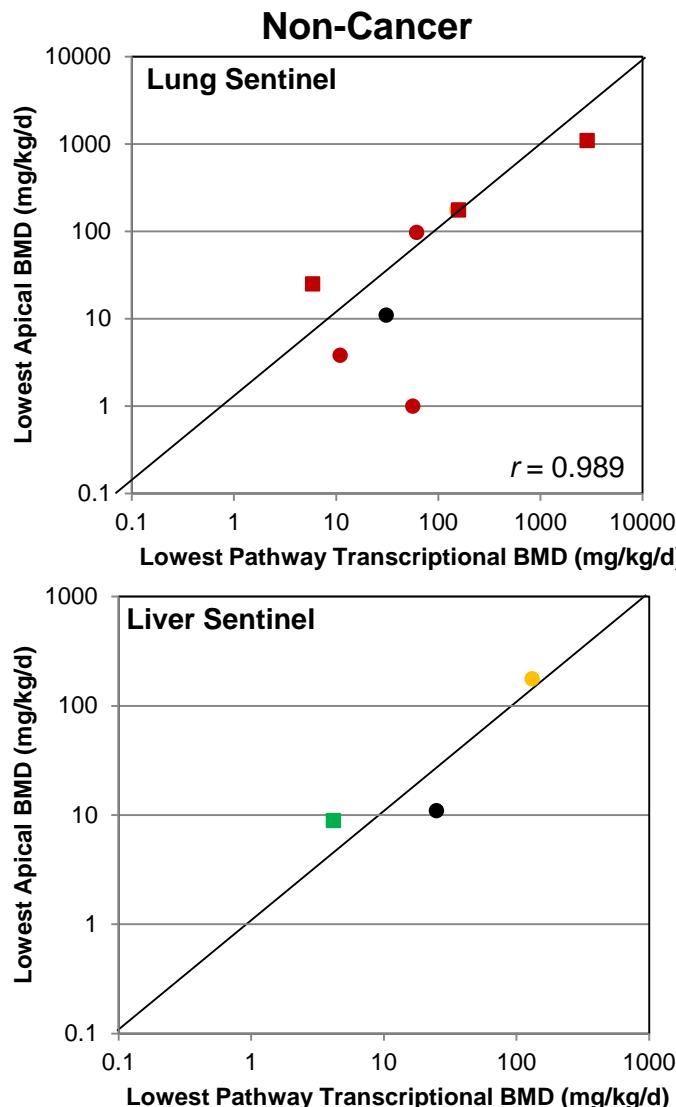


Combined Correlation Between Cancer and Transcriptional PODs



Thomas et al., *Toxicol Sci*, 2013

Correlation With Transcriptional PODs in Sentinel Tissue



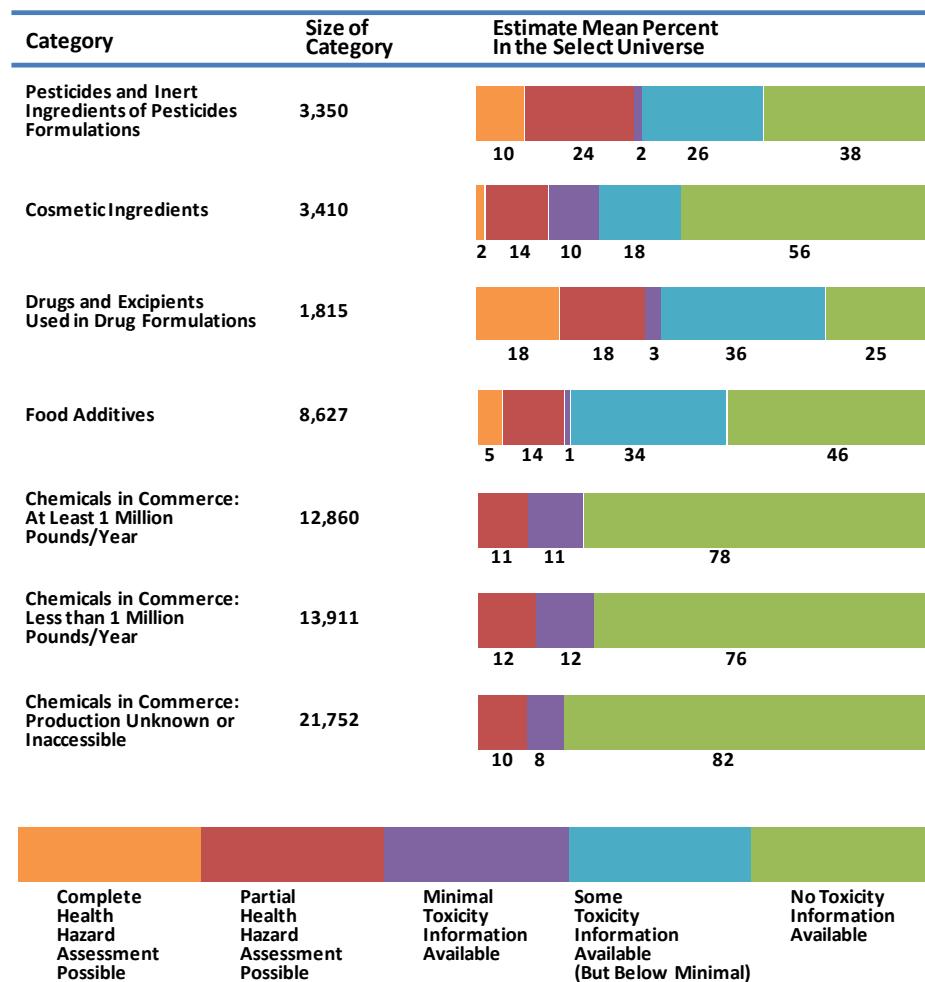
What Did These Studies Tell Us...

- Transcriptomic dose response alterations correlate with both noncancer- and cancer-related apical endpoints
- Correlation between apical responses and transcriptional no effect levels appear stable over time
- The average ratio apical and transcriptional points-of-departure for the most sensitive response was less than two-fold
- Transcriptional points-of-departure in sentinel tissues may provide reasonable surrogates for those in the target tissue

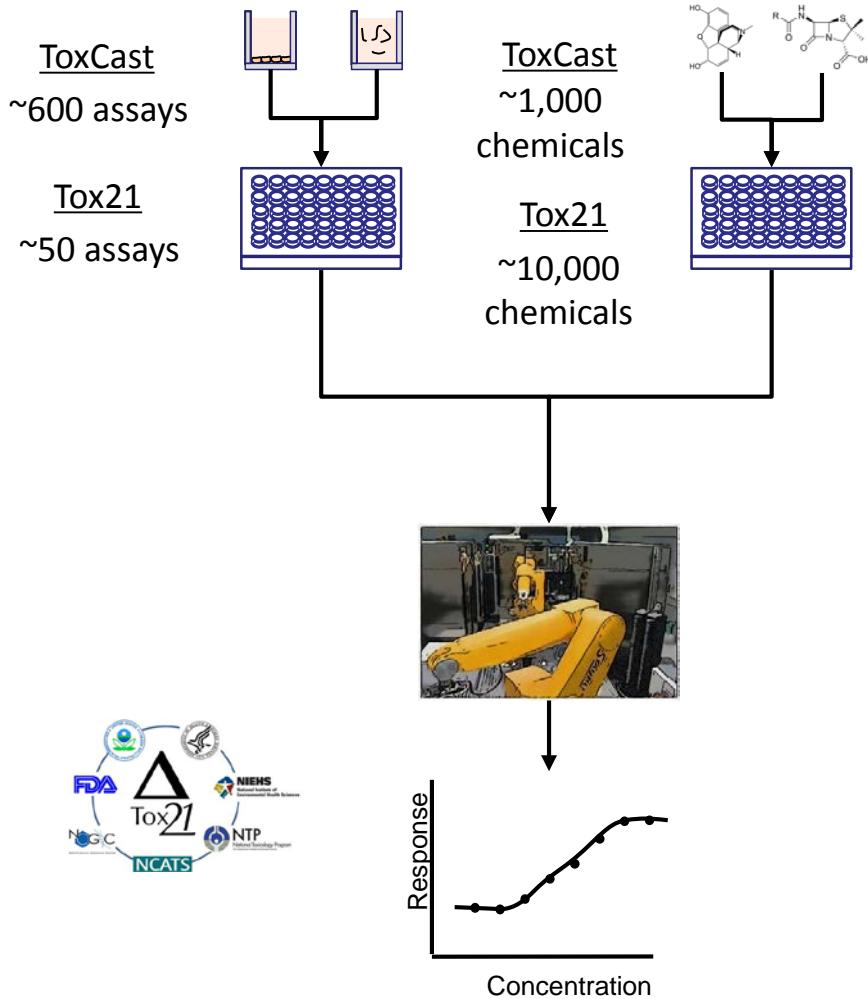
But, There are Other Challenges for EPA to Consider

- 1984 US NRC report
- Major challenge is too many chemicals and not enough data
- Total = 65,725
- No tox data = 46,000

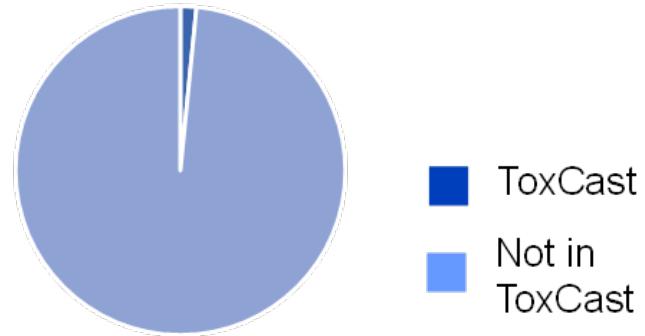
US National Research Council, 1984



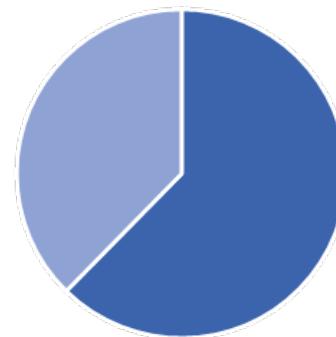
Using High-Throughput Screening to Address Data Gap



Gene Coverage



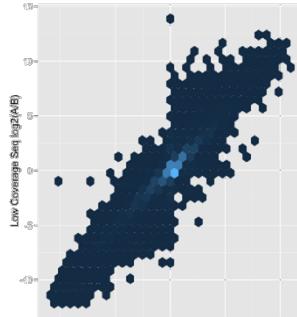
Pathway Coverage*



*At least one gene from pathway represented

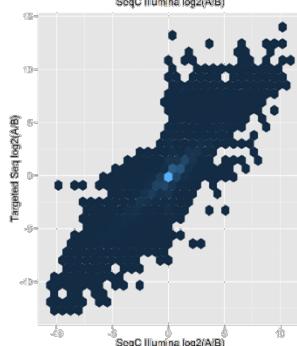
Searching for a High-Throughput Toxicogenomics Platform

Technical Comparison



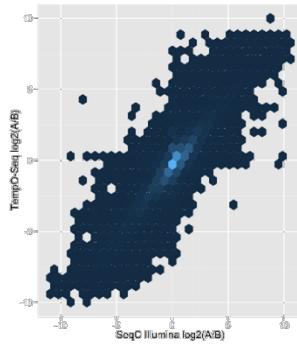
TruSeq

r^2 0.74



Requirements:

- Low cost
- Whole genome



Low Coverage

r^2 0.83

Functional Comparison

TruSeq

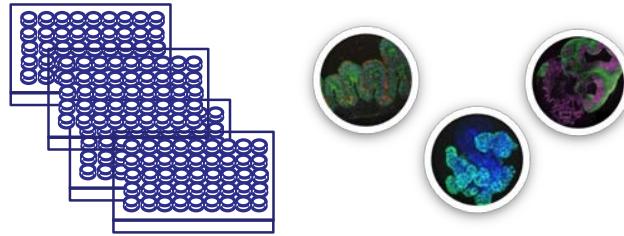
0/5 (0%)

DSeq
(30%)

Low Coverage

0/5 (0%)

Developing a Portfolio of TGX Experimental and Analytical Tools



High-Throughput Toxicogenomics Screen

- Multiple cell types
- Thousands of chemicals
- Whole transcriptome (EPA)
- S1500+ (NTP)

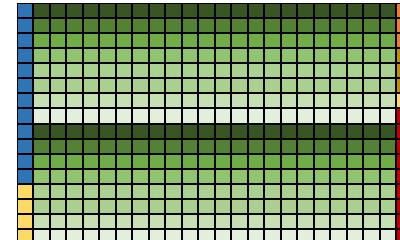
Assay Design for Rigorous QC and Performance Validation

	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19	20	21	22	23	24	
A	MAQC-A (Us)	100	100	100	100	100	100	100	100	100	100	100	100	100	100	100	100	100	100	100	100	100	100	non-treated	
B	MAQC-A (Us)	30	30	30	30	30	30	30	30	30	30	30	30	30	30	30	30	30	30	30	30	30	30	30	non-treated
C	MAQC-B (Us)	10	10	10	10	10	10	10	10	10	10	10	10	10	10	10	10	10	10	10	10	10	10	10	non-treated
D	MAQC-B (Us)	3	3	3	3	3	3	3	3	3	3	3	3	3	3	3	3	3	3	3	3	3	3	3	DMSO
E	Bulk Lysate (DMSO)	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	DMSO
F	Bulk Lysate (DMSO)	0.3	0.3	0.3	0.3	0.3	0.3	0.3	0.3	0.3	0.3	0.3	0.3	0.3	0.3	0.3	0.3	0.3	0.3	0.3	0.3	0.3	0.3	0.3	DMSO
G	Bulk Lysate (Trichostatin)	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1	DMSO [No Label]
H	Bulk Lysate (Trichostatin)	0.03	0.03	0.03	0.03	0.03	0.03	0.03	0.03	0.03	0.03	0.03	0.03	0.03	0.03	0.03	0.03	0.03	0.03	0.03	0.03	0.03	0.03	Trichostatin (1 µM)	
I	Lysis Buffer (Us)	100	100	100	100	100	100	100	100	100	100	100	100	100	100	100	100	100	100	100	100	100	100	Trichostatin (1 µM)	
J	Lysis Buffer (Us)	30	30	30	30	30	30	30	30	30	30	30	30	30	30	30	30	30	30	30	30	30	30	30	Trichostatin (1 µM)
K	MAQC-A (Them)	10	10	10	10	10	10	10	10	10	10	10	10	10	10	10	10	10	10	10	10	10	10	10	Genistein (10 µM)
L	MAQC-A (Them)	3	3	3	3	3	3	3	3	3	3	3	3	3	3	3	3	3	3	3	3	3	3	3	Genistein (10 µM)
M	MAQC-B (Them)	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	Genistein (10 µM)
N	MAQC-B (Them)	0.3	0.3	0.3	0.3	0.3	0.3	0.3	0.3	0.3	0.3	0.3	0.3	0.3	0.3	0.3	0.3	0.3	0.3	0.3	0.3	0.3	0.3	0.3	Sirolimus (0.1 µM)
O	Lysis Buffer (Them)	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1	Sirolimus (0.1 µM)
P	Lysis Buffer (Them)	0.03	0.03	0.03	0.03	0.03	0.03	0.03	0.03	0.03	0.03	0.03	0.03	0.03	0.03	0.03	0.03	0.03	0.03	0.03	0.03	0.03	0.03	Sirolimus (0.1 µM)	

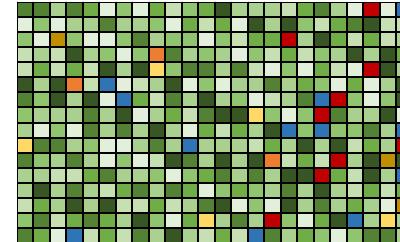
No Cells

Screen Design

- Large bank of cytogenetically and functionally characterized cells
- 8 point concentration response
- Single time point
- Parallel HCl screen

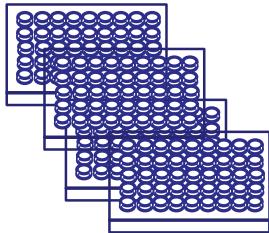


Randomized treatment



Josh Harrill, Unpublished

Developing a Portfolio of TGX Experimental and Analytical Tools



High-Throughput Toxicogenomics Screen

- Multiple cell types
- Thousands of chemicals
- Whole transcriptome (EPA)
- S1500+ (NTP)

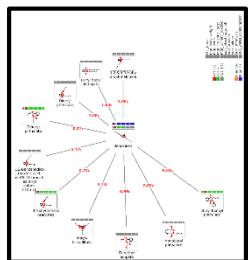
Mode of Action/MIE



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

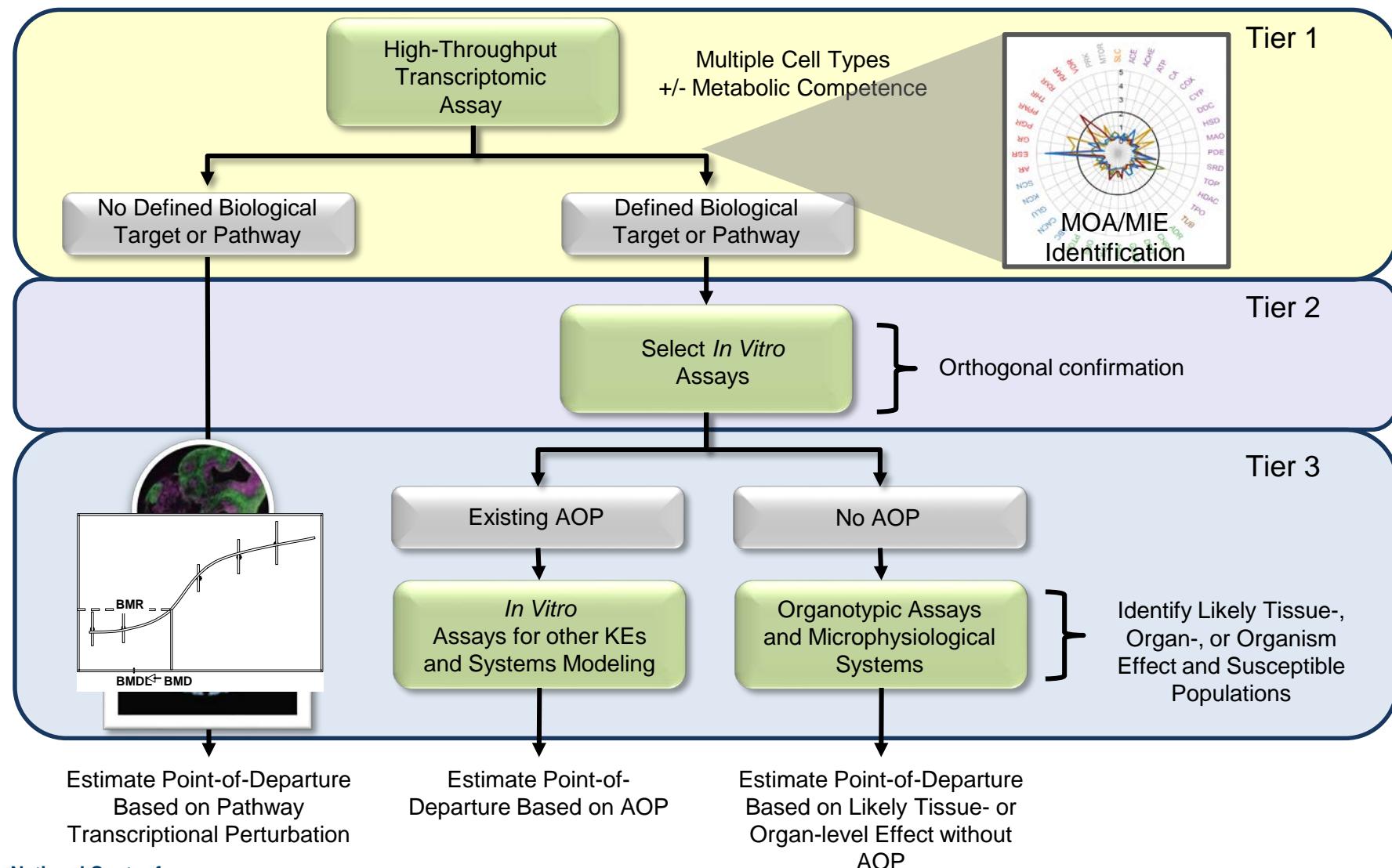
Read Across and Category Approach

- GenRA tool
- Chemical and Biological Read Across
- Quantitative estimates of uncertainty

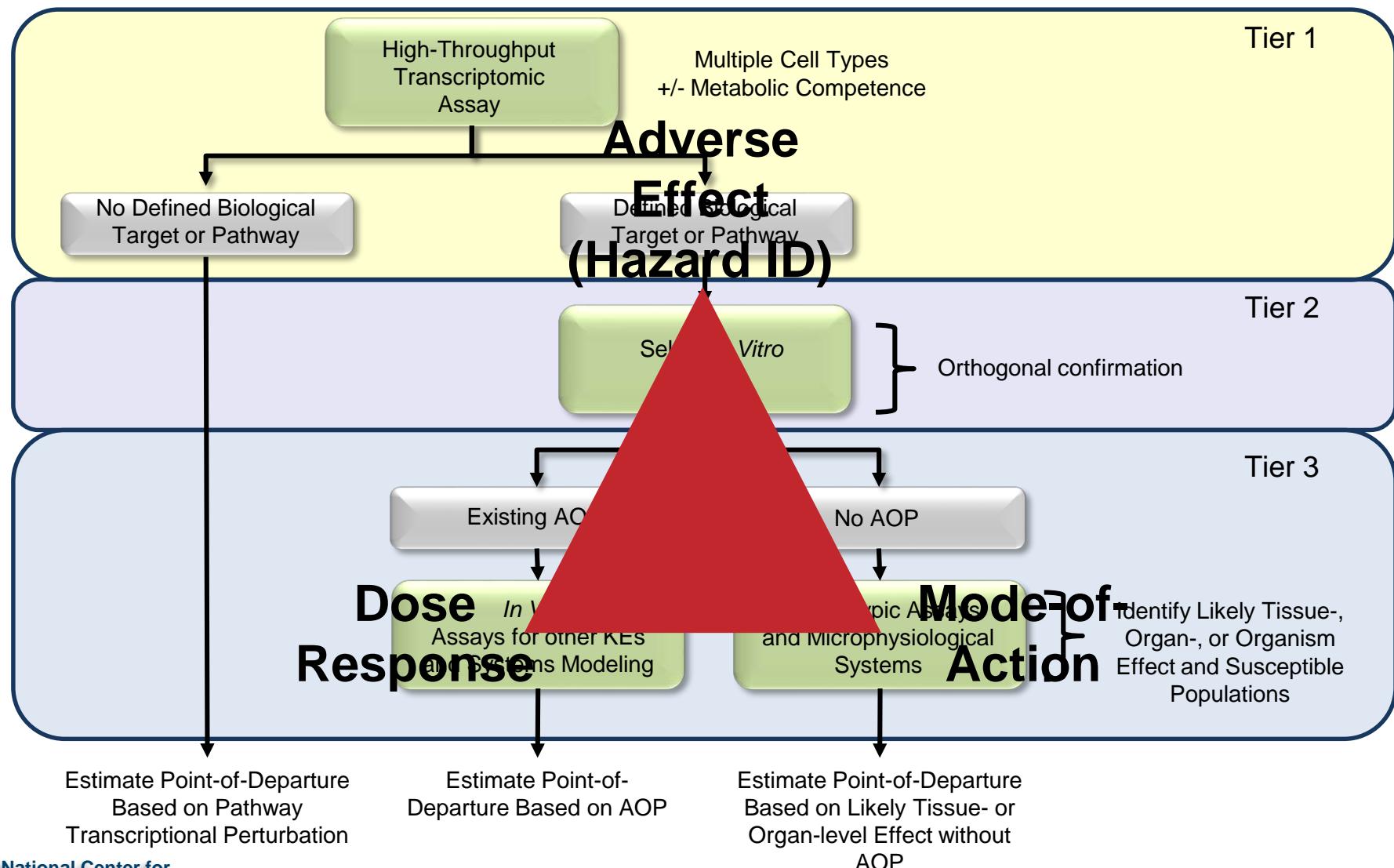


Shah et al., 2016

Integrating Components Into a Tiered Testing and Assessment Strategy



Hitting All the Points of the Toxicogenomics Bermuda Triangle



Keys to Breaking Out of the Toxicogenomics Bermuda Triangle

- Develop best practices and data use recommendations that cover a range of regulatory decisions
- Systematize and convergence on analysis approaches
- Develop flexible validation approaches based on performance criteria and reference standards (e.g., MAQC) that adapt to evolution in technology
- Characterize qualitative and quantitative uncertainty in application of the technology to specific regulatory decisions while also doing this for legacy approaches
- Continue progress in the transition apical to molecular/pathway-based endpoints as a basis for safety-related decisions

Need evolutionary leap, not incremental advances

Thank You for Your Attention!

Tox21 Colleagues:

NTP Crew
FDA Collaborators
NCATS Collaborators

NCEA Colleagues:

Scott Wesselkamper
Nina Wang
Jason Lambert
Janet Hess-Wilson
Dan Petersen
Jay Zhao

Hamner Colleagues:

Barbara Wetmore
Reetu Singh
Bethany Parks
Linda Pluta
Michael Black
Eric Healy
Longlong Yang



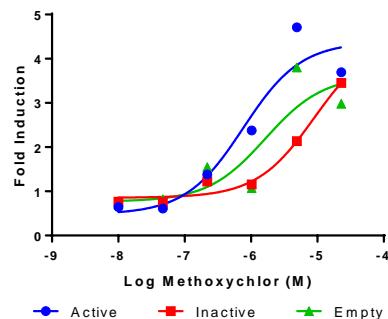
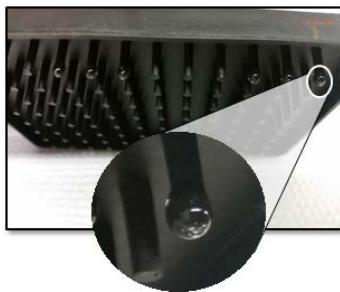
EPA's National Center for Computational Toxicology

Beginning to Address Metabolic Competence

“Extracellular” Approach



Chemicals metabolism in the media or buffer of cell-based and cell-free assays

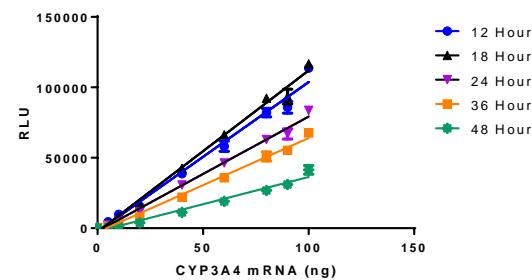
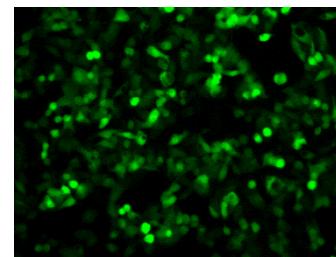


More closely models effects of hepatic metabolism and generation of circulating metabolites

“Intracellular” Approach



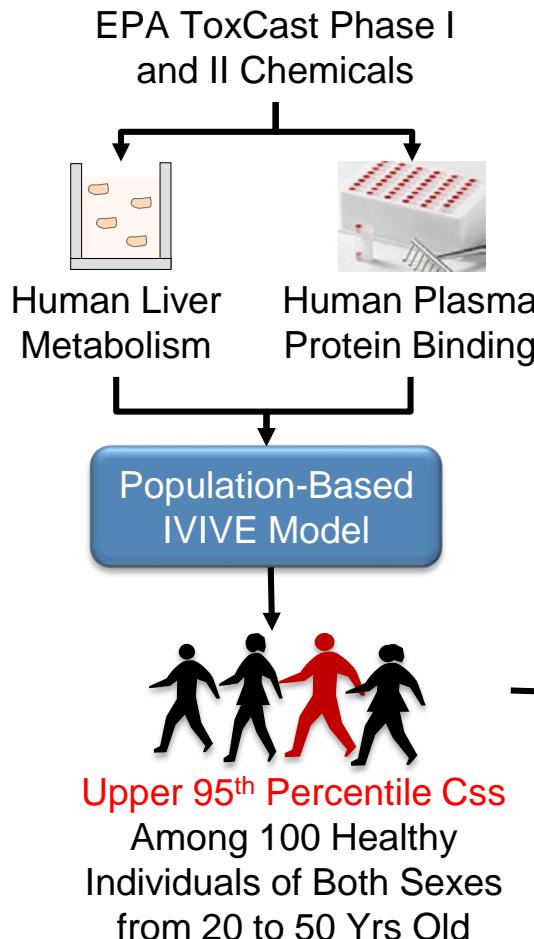
Capable of metabolizing chemicals inside the cell in cell-based assays



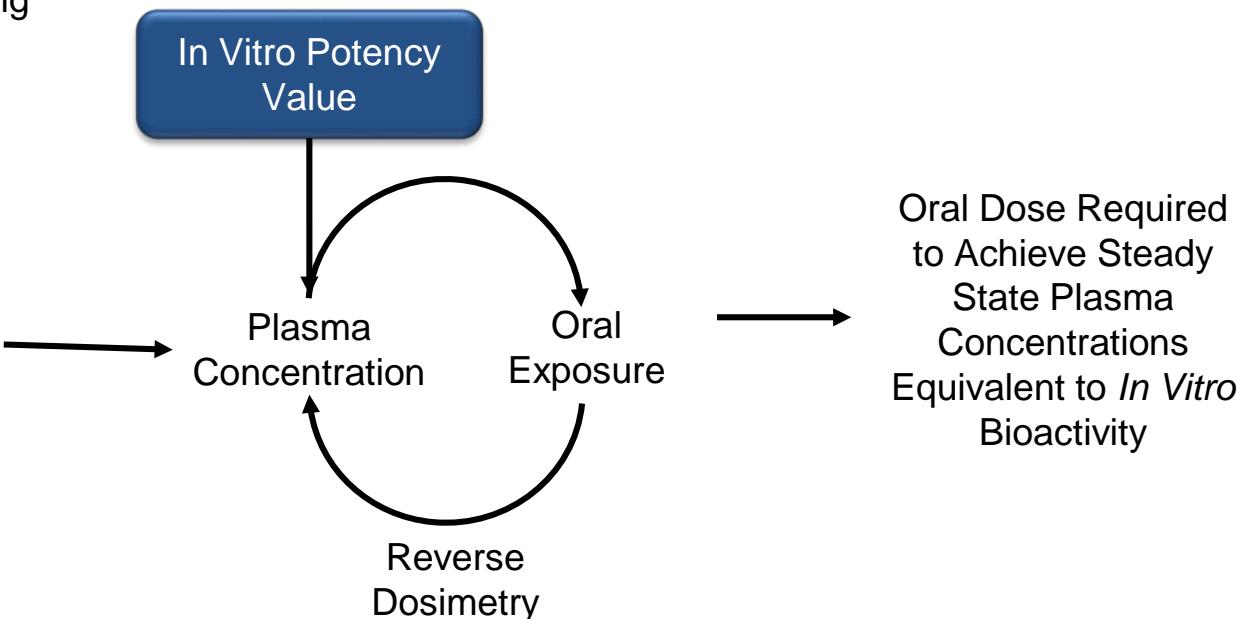
More closely models effects of target tissue metabolism

Integrated approach to model *in vivo* metabolic bioactivation and detoxification

Adding the High-Throughput Toxicokinetic Component

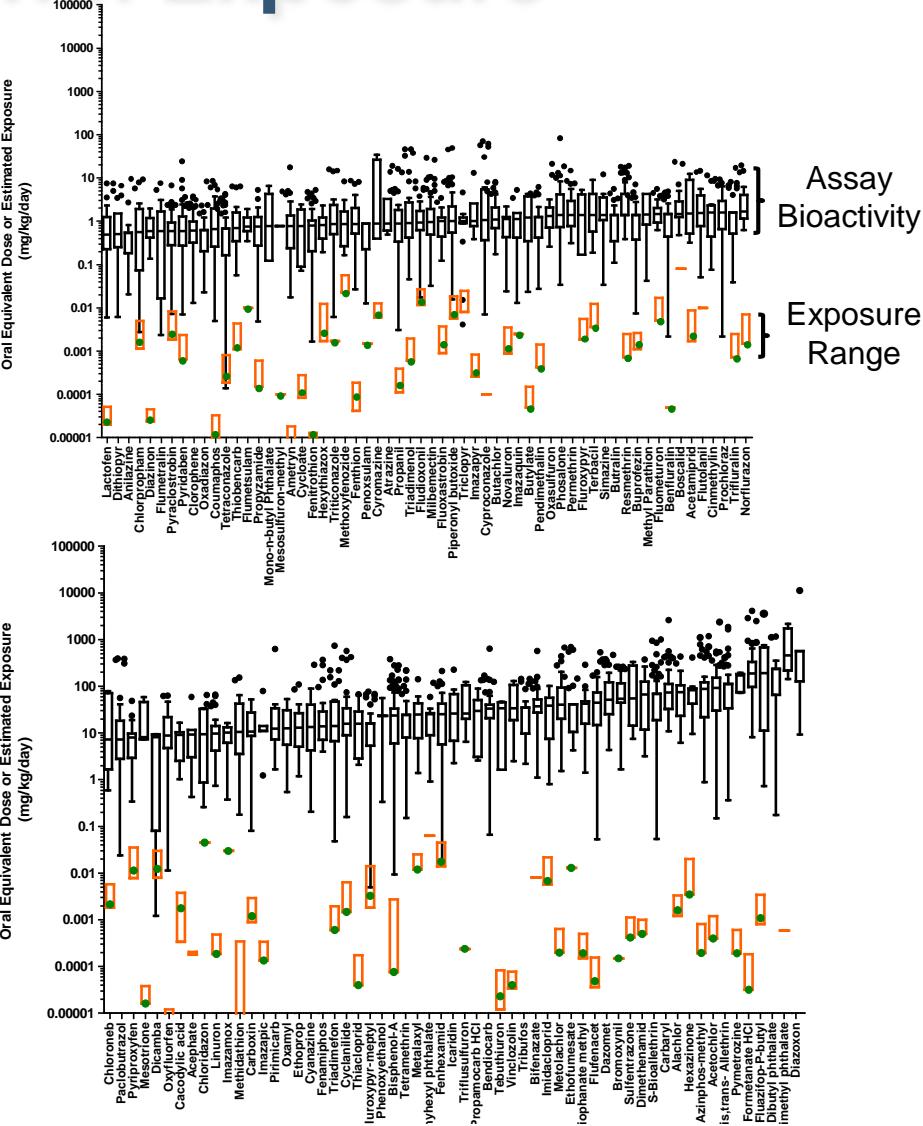
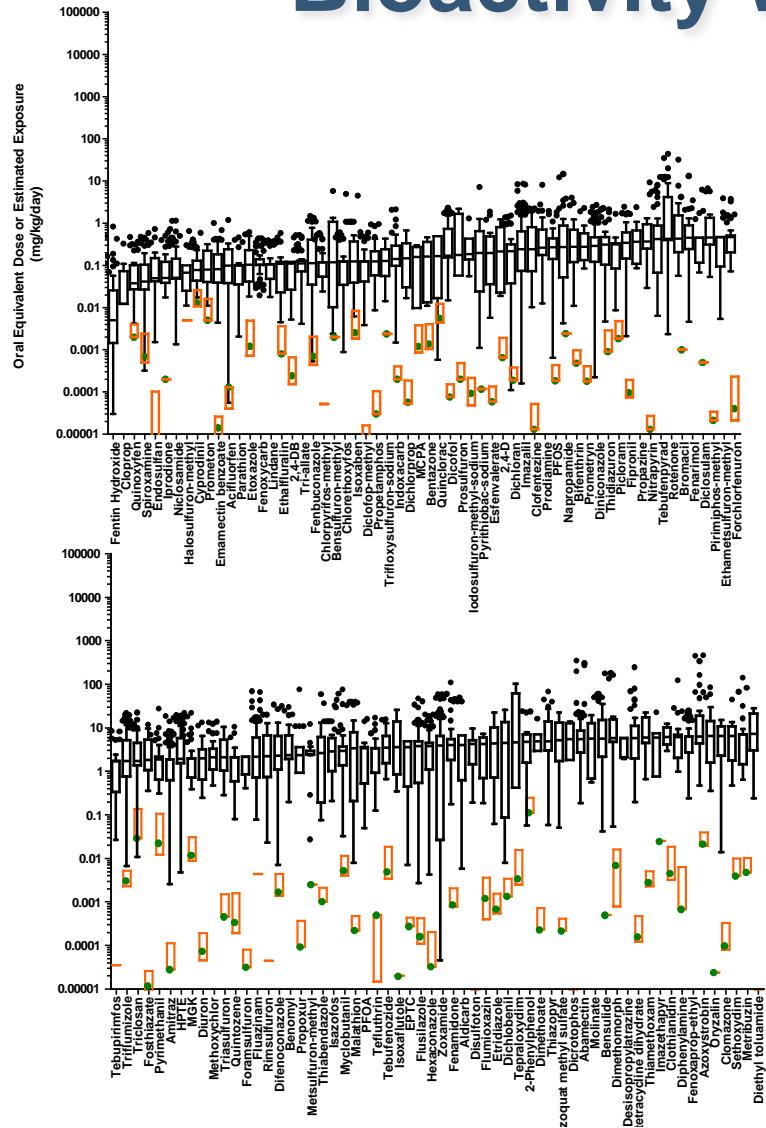


- Currently evaluated ~700 ToxCast Phase I and II chemicals
- Models available through “httk” R package (<https://cran.r-project.org/web/packages/httk/>)

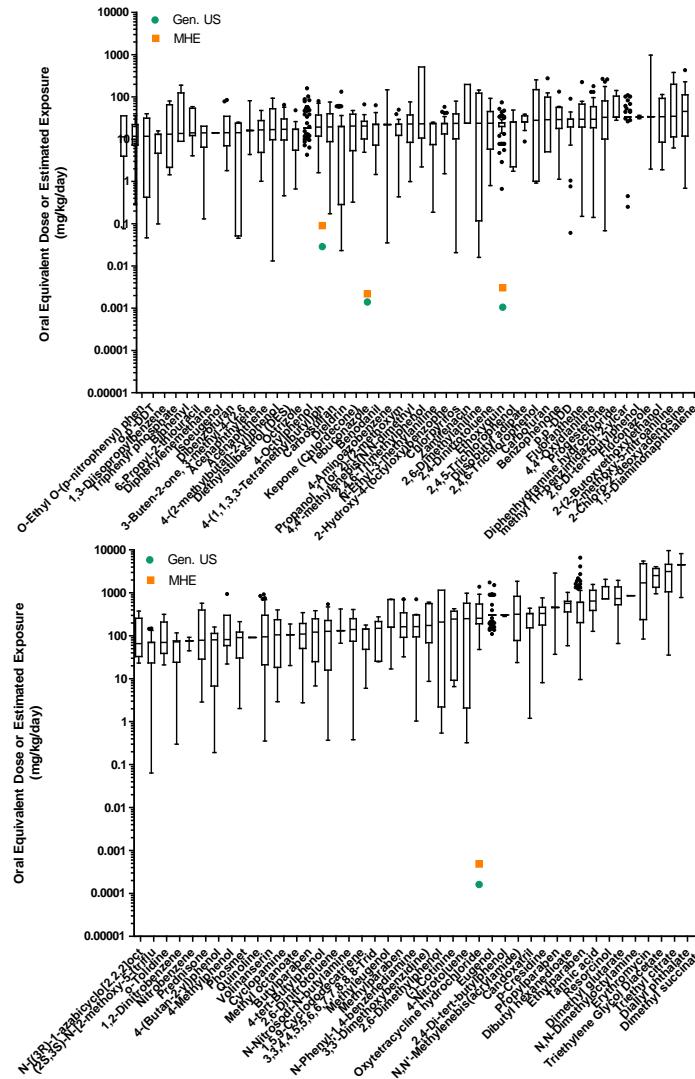
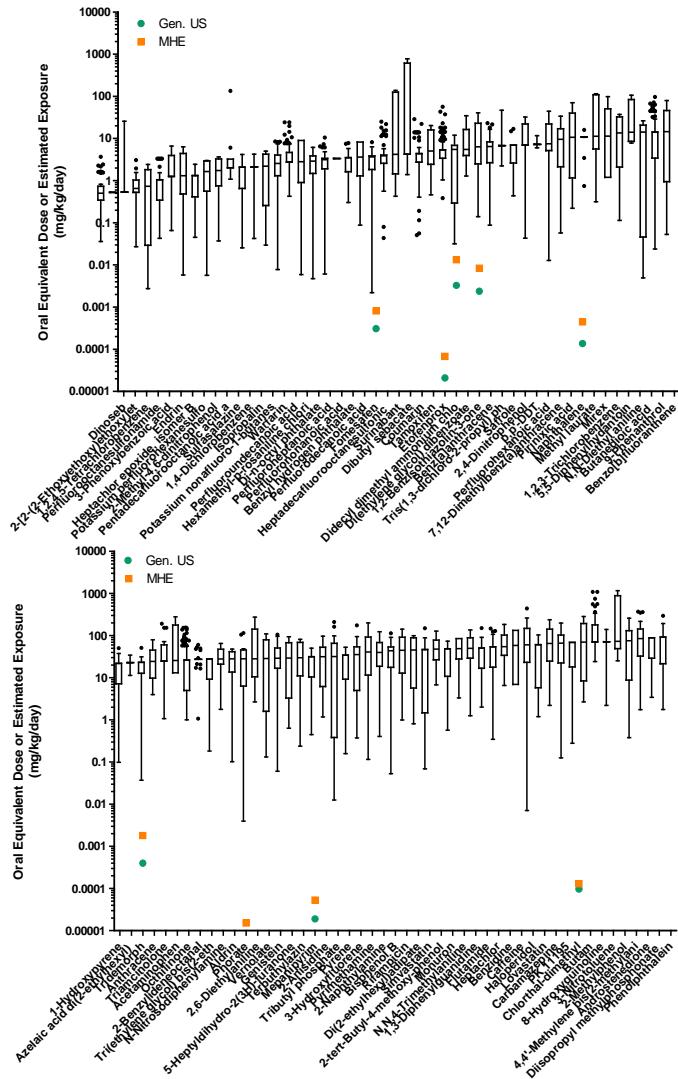


Rotroff et al., *Tox Sci.*, 2010
Wetmore et al., *Tox Sci.*, 2012

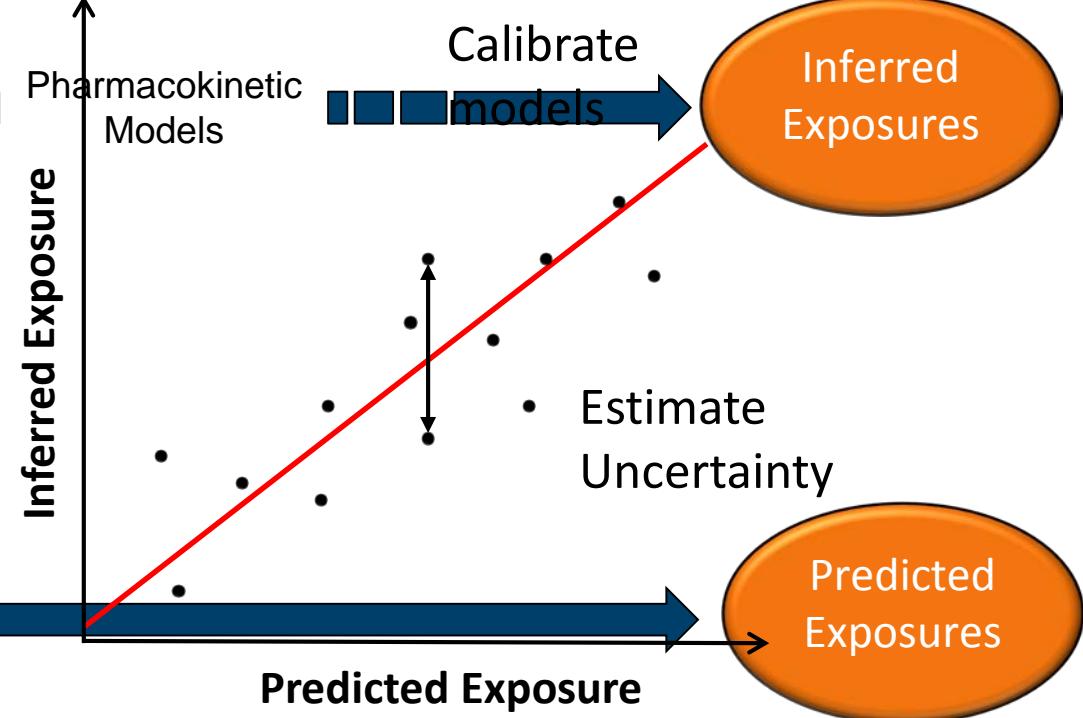
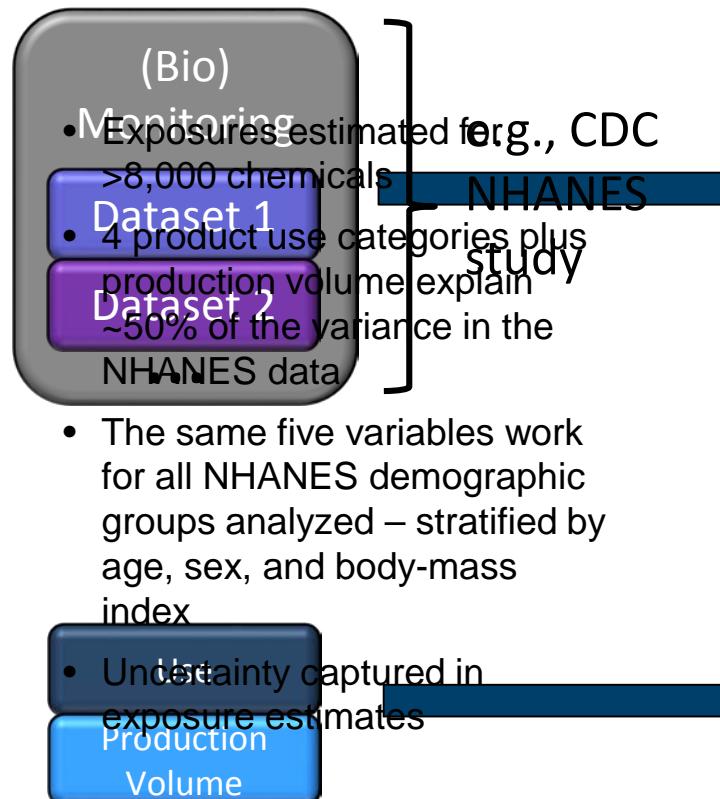
Comparing Dosimetry Adjusted Bioactivity with Exposure



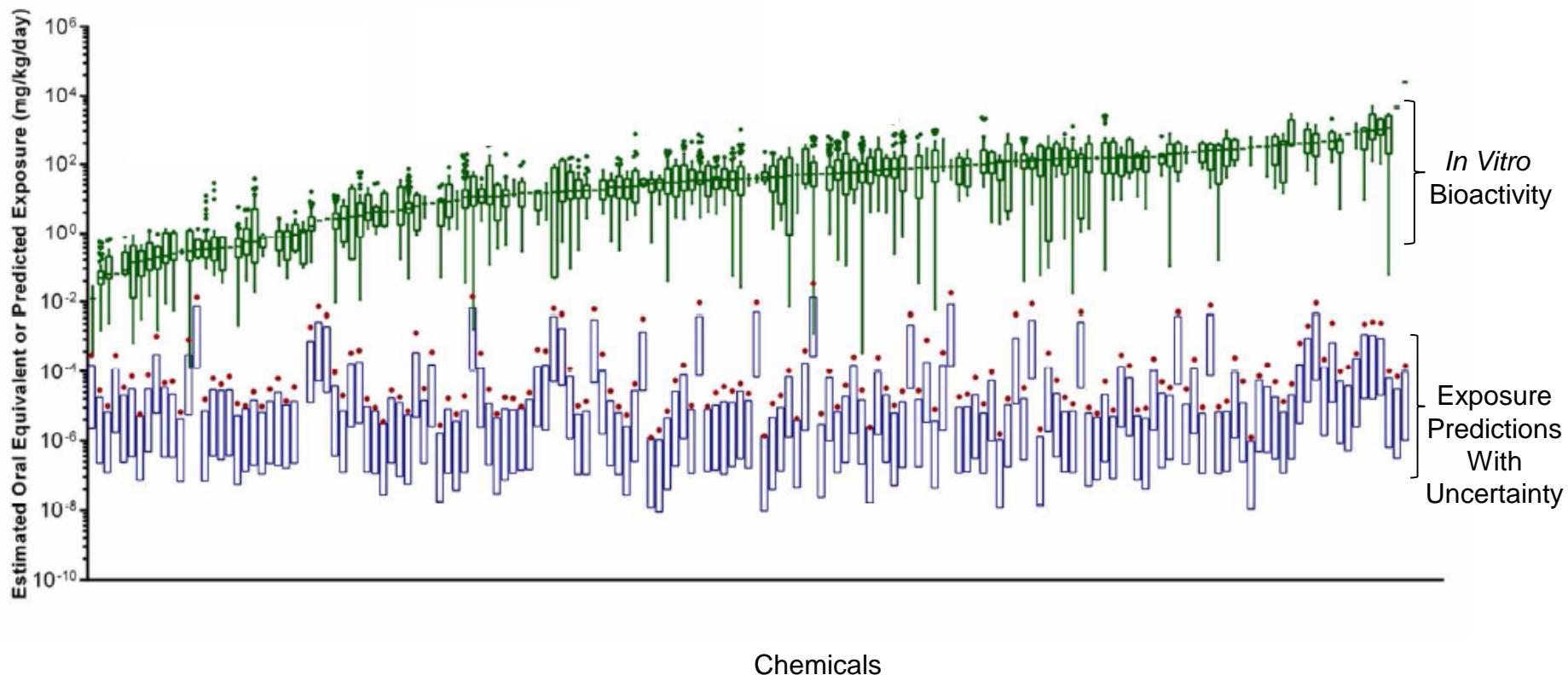
But, Exposure Information is Lacking on Most Chemicals



Adding the High-Throughput Exposure Component



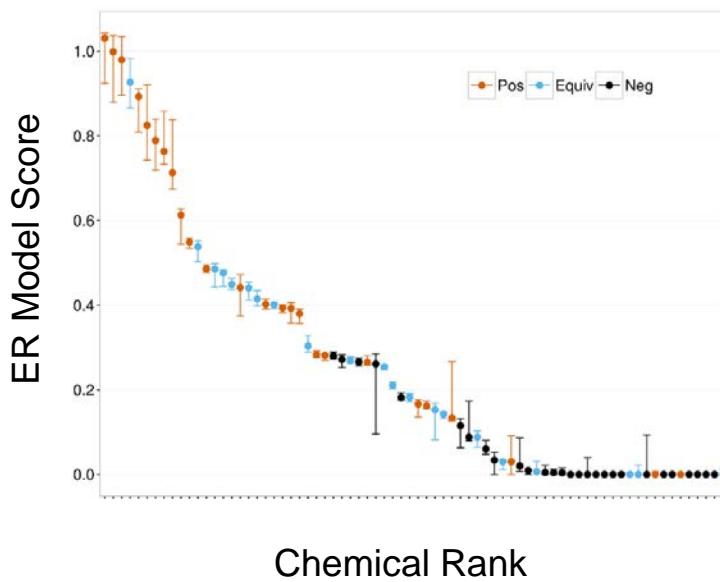
Comparing Bioactivity with Exposure Predictions for Risk Context



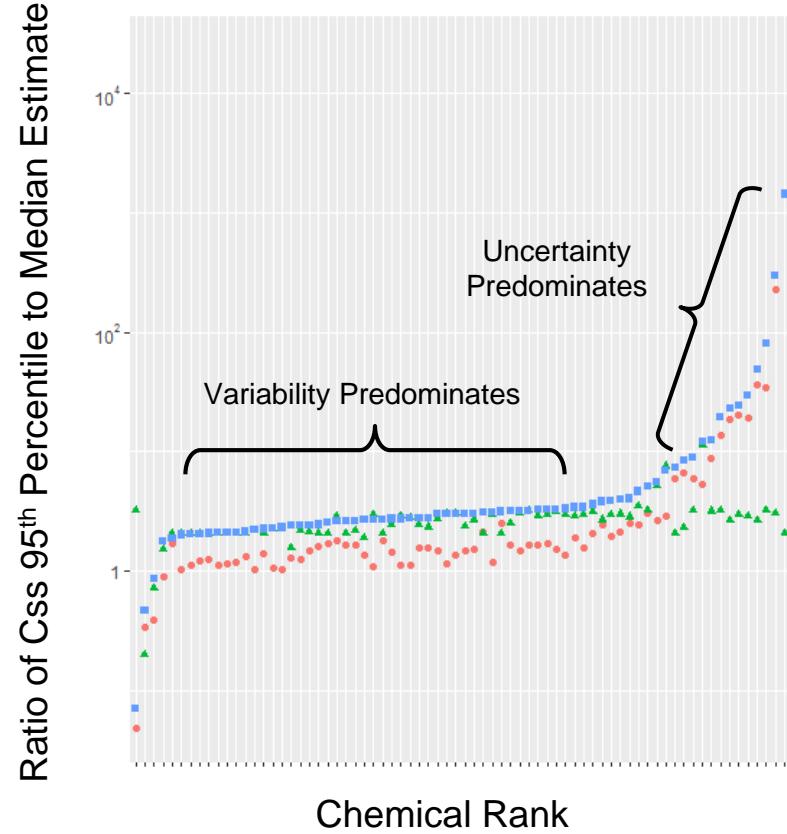
Wetmore *et al.*, *Tox Sci.*, 2015

Adding in Uncertainty and Variability for PD and PK

Propagation of Experimental Uncertainty in Models of ER Potency



Propagation of Experimental Uncertainty in High-Throughput Toxicokinetic Estimates



Covering All the Components of a 21st Century Risk Assessment

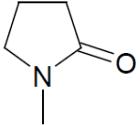
EPA
United States
Environmental Protection Agency

EPA Document# 740-R1-5002
March 2015
Office of Chemical Safety and
Pollution Prevention

TSCA Work Plan Chemical Risk Assessment

N-Methylpyrrolidone:
Paint Stripper Use

CASRN: 872-50-4



March 2015

TABLE OF CONTENTS

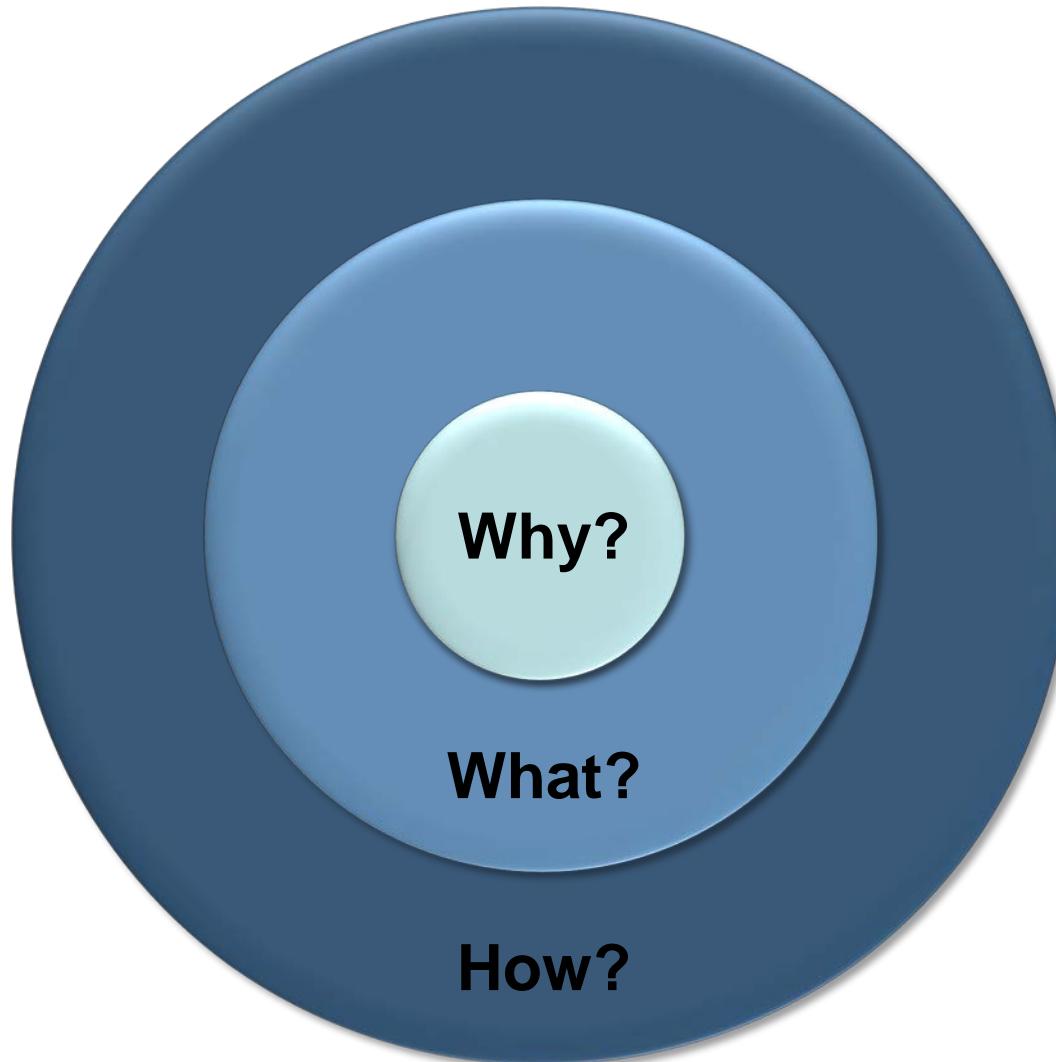
TABLE OF CONTENTS	2
AUTHORS / CONTRIBUTORS / ACKNOWLEDGEMENTS / REVIEWERS	9
ABBREVIATIONS	11
EXECUTIVE SUMMARY	14
1 BACKGROUND AND SCOPE	20
1.1 INTRODUCTION	20
1.2 USES AND PRODUCTION VOLUMES	21
1.2.1 Assessment and Regulatory History	21
1.2.2 Scope of the Assessment	23
1.3 PROBLEM FORMULATION	25
1.3.1 Physical and Chemical Properties	25
1.3.2 Environmental Fate	26
1.3.3 Conceptual Model	26
1.3.3.1 Exposure Pathways	26
1.3.3.2 Health Effects and Human Receptors	27
1.3.4 Analysis Plan	28
2 EXPOSURE ASSESSMENT	30
2.1 OCCUPATIONAL EXPOSURES	30
2.1.1 Approach and Methodology	30
2.1.1.1 Identification of Relevant Industries	31
2.1.1.2 Approach for Determining Occupational Exposure Data and Input Parameters for PBPK Modeling	32
2.1.1.3 Estimates of Occupational Exposure Parameters and Number of Exposed Workers	32
2.1.2 Use of Occupational Exposure Estimates in PBPK Modeling	35
2.2 CONSUMER EXPOSURES	37
2.2.1 Approach and Methodology	37
2.2.1.1 Consumer Dermal Exposure Assessment	38
2.2.1.2 Consumer Users and Residential Non-Users Inhalation Exposure Assessment	38
2.2.2 Model Outputs and Exposure Calculations	46
2.2.3 Use of Consumer Exposure Estimates in PBPK Modeling	46
3 HAZARD IDENTIFICATION AND DOSE-RESPONSE	48
3.1 APPROACH AND METHODOLOGY	48
3.1.1 Selection of Peer-Reviewed Assessments for Hazard Identification and Dose-Response Analysis	49
3.1.2 Hazard Summary and Hazard Identification	49
3.1.3 Selection of Developmental and Fetal Toxicity Data	60
3.1.3.1 Decreased Fetal Weight	63
3.1.3.2 Respirations and Oral NO ₂	65
3.1.3.3 Other Fetal Effects	66
3.1.3.4 Conclusion and Selection of Key Endpoints	67
3.2 DOSE-RESPONSE ASSESSMENT AND STUDY SELECTION	68
3.2.1 Identification of Studies for BMD Modeling	68
3.2.2 Derivation of Internal Doses	69
3.2.3 Dose-Effect Curves	73
3.2.4 Dose-Response	73

Hazard
Exposure
Dose Response,
Phys Chem
PK, and PODs

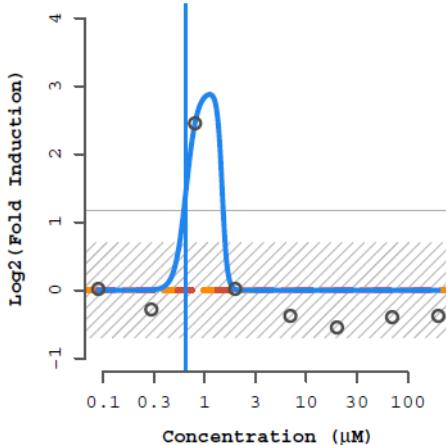
Variability ✓
Risk Summary ✓
Uncertainty ✓

3.2.5 Considerations for Incorporating Exposure Variability	78
4 HUMAN HEALTH RISK CHARACTERIZATION	80
4.1 RISK ESTIMATION APPROACH FOR ACUTE AND CHRONIC EXPOSURES	80
4.1.1 Acute Effects from Consumer Use	82
4.1.2 Subchronic Effects from Consumer Use	87
4.1.3 Risk Estimates for Chronic Occupational Exposures to NMP	90
4.2 HUMAN HEALTH RISK CHARACTERIZATION SUMMARY	94
4.3 KEY SOURCES OF UNCERTAINTY AND DATA LIMITATIONS	95
4.3.1 Key Uncertainties in the Occupational Exposure Assessment	95
4.3.2 Key Uncertainties in the Consumer Exposure Assessment	96
4.3.3 Key Uncertainties in the Acute and Subchronic Risk Assessments	99
4.3.4 Key Uncertainties in the Chronic Risk Assessment	101
4.4 RISK ASSESSMENT CONCLUSIONS	103
REFERENCES	106
APPENDICES	120
Appendix A ENVIRONMENTAL EFFECTS SUMMARY	121
A-1 ACUTE TOXICITY TO AQUATIC ORGANISMS	121
A-2 CHRONIC TOXICITY TO AQUATIC ORGANISMS	123
A-3 TOXICITY TO SEDIMENT AND SOIL ORGANISMS	123
A-4 TOXICITY TO WILDLIFE	123
A-5 SUMMARY OF ENVIRONMENTAL HAZARD ASSESSMENT	124
Appendix B CHEMICAL REPORTING DATA	125
B-1 CONSUMER USES	127
B-2 PAINT STRIPPING APPLICATIONS	128
Appendix C STATE NMP REGULATIONS	129
Appendix D OCCUPATIONAL EXPOSURE ASSESSMENT SUPPORT INFORMATION	130
D-1 SUMMARY OF DERMAL EXPOSURE PARAMETERS, INHALATION CONCENTRATIONS AND EXPOSURE REDUCTION FACTORS	130
D-2 DATA NEEDS AND DATA COLLECTION	130
D-3 INDUSTRIES THAT EMPLOY PAINT STRIPPING ACTIVITIES	133
D-4 OCCUPATIONAL PAINT STRIPPING PROCESSES AND ASSOCIATED WORKER ACTIVITIES	134
D-5 FACILITY AND POPULATION DATA AND INFORMATION	139
D-6 DERMAL EXPOSURE PARAMETERS	144
D-7 OCCUPATIONAL INHALATION EXPOSURE LITERATURE DATA	146
Appendix E CONSUMER EXPOSURE ASSESSMENT	153
E-1 ESTIMATION OF EXPOSURE PROFILES FOR PAINT REMOVERS/STRIPPERS	153
E-2 SENSITIVITY ANALYSIS FOR INHALATION SCENARIOS	165
E-3 INHALATION EXPOSURE SCENARIO INPUTS	166
E-4 INHALATION MODEL OUTPUTS AND EXPOSURE CALCULATIONS	177
E-5 MCCEM INHALATION MODELING CASE SUMMARIES	185
E-5-1 NMP Scenario 1. Coffee Table, Brush-On, Workshop, User in ROH during wait time, 0.45 ACH, 0.25 Weight Fraction	185
E-5-2 NMP Scenario 2. Coffee Table, Brush-On, Workshop, User in Workshop during wait time, 0.45 ACH, 0.5 Weight Fraction	188
E-5-3 NMP Scenario 3. Chest, Brush-On, Workshop, User in ROH during wait time, 0.18 ACH, 0.5 Weight Fraction	191

'Golden Circle' of 21st Century Risk Assessment



Regulatory Applications Require More Focus on Quality and Transparency



ASSAY: AEID117 (ATG ERA_TRANS)

NAME: Thioglycolic acid
CHID: 26141 CASRN: 68-11-1
SPID(S): TX007664
L4ID: 420385

HILL MODEL (in red):
 tp ga gw
 val: 3.1e-11 -2.15 0.416
 sd: NaN NaN NaN

GAIN-LOSS MODEL (in blue):
 tp ga gw la lw
 val: 2.93 -0.184 8 0.173 18
 sd: 3.56 0.334 9.48 5.82 814

CNST	HILL	GNLSS
AIC: 20.14	26.14	17.79
PROB: 0.23	0.01	0.76
RMSE: 0.92	0.92	0.32

MAX_MEAN: 2.45 MAX_MED: 2.45 BMAD: 0.233
 COFF: 1.17 HIT-CALL: 1 FITC: 50 ACTP: 0.77

FLAGS:
 Only one conc above baseline, active
 Borderline active

- Public release of Tox21 and ToxCast data on PubChem and EPA web site (raw and processed data)
- Publicly available ToxCast data analysis pipeline
 - Data quality flags to indicate concerns with chemical purity and identity, noisy data, and systematic assay errors
- Tox21 and ToxCast chemical libraries have undergone analytical QC and results publicly available
- Public posting of ToxCast procedures
 - Chemical Procurement and QC
 - Data Analysis
 - Assay Characteristics and Performance
- External audit on ToxCast data and data analysis pipeline
- Migrating ToxCast assay annotations to OECD 211 compliant format

Application to Regulatory Decisions for Endocrine Screening

Prioritization of the EDSP Universe of Chemicals

Prioritization of the Endocrine Program Universe of Chemical Receptor Adverse Outcome Computational Toxicology

U.S. Environmental Protection Agency Endocrine Disruptor Screening Program

Jointly developed by:

Office of Chemical Safety and Pollution Prevention
Office of Science Coordination and Policy (OSCP)
Office of Pesticide Programs (OPP)
Office of Pollution Prevention and Toxics (OPPT)

Office of Water (OW)
Washington, DC 20460

Office of Research and Development (ORD)
National Environmental and Effects Health Research
Mid-Continent Ecology Division (MED), Duluth, MN
Toxicity Assessment Division (TAD), RTP, NC 27111

National Center for Computational Toxicology (NCC)
Research Triangle Park, NC 27709

December 2012

Integrated Bioactivity and Exposure Ranking

Integrated Bioactivity and Exposure Ranking: A Computational Approach for the Identification and Screening of Chemicals in Endocrine Disruptor Screens

Environmental Protection Agency Endocrine Disruptor Screening Program

Exposure SAP White Paper

New High-throughput Methods to Estimate Chemical Exposure

Scientific Advisory Panel Meeting, July 2014

SAP December 2-5

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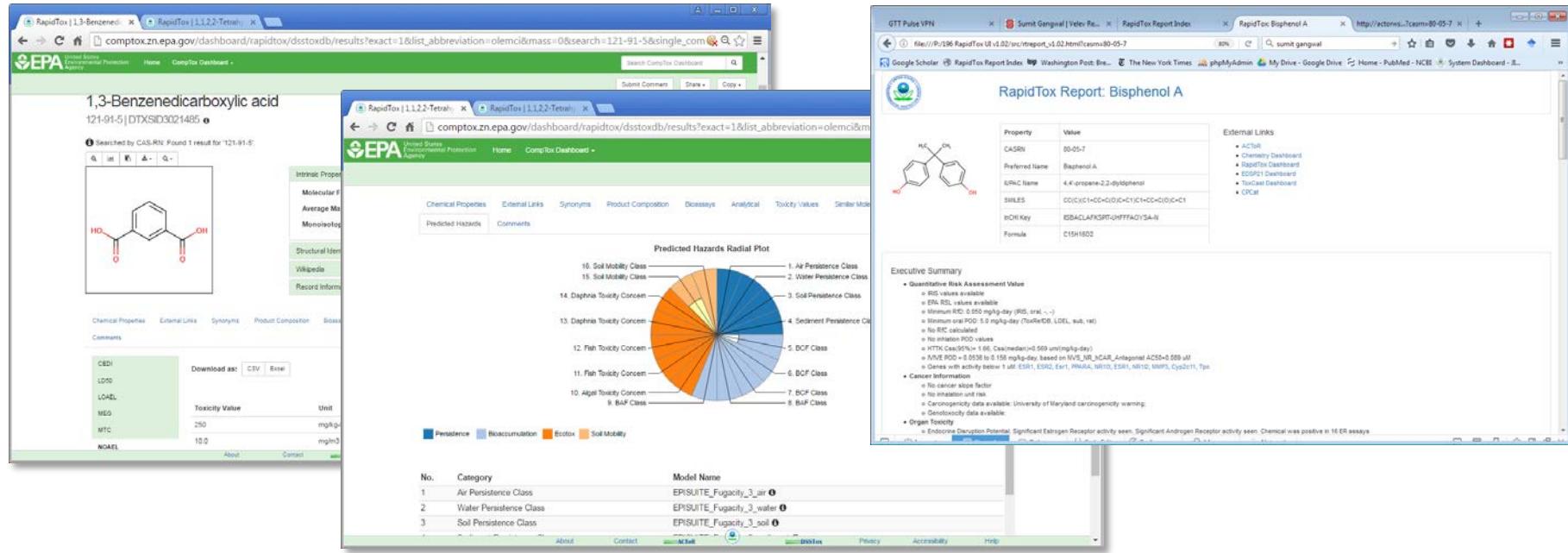
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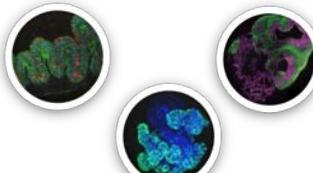
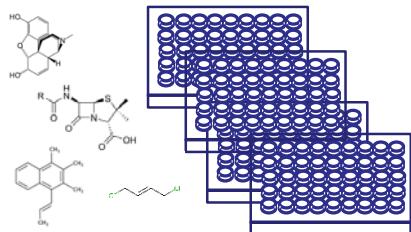
Beginning Application to Quantitative Risk Assessment Through New ‘RapidTox’ Dashboard



- Semi-automated decision support tool with dashboard interface for high-throughput risk assessments
- Integrate a range of information related to chemical properties, fate and transport, hazard, and exposure
- Transparent and interactive enough to enable expert users to review the assumptions made and refine the predictions
- Deliver quantitative toxicity values with associated estimates of uncertainty

Incorporating HT Toxicogenomics to Address This Challenge

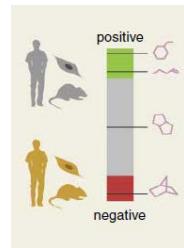
Thousands of Chemicals



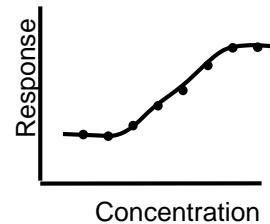
Multiple Cell Types



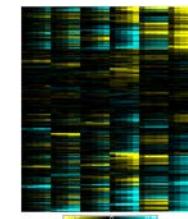
High-Throughput
Toxicogenomics



Mode-of-Action



Dose Response



Grouping and
Read Across