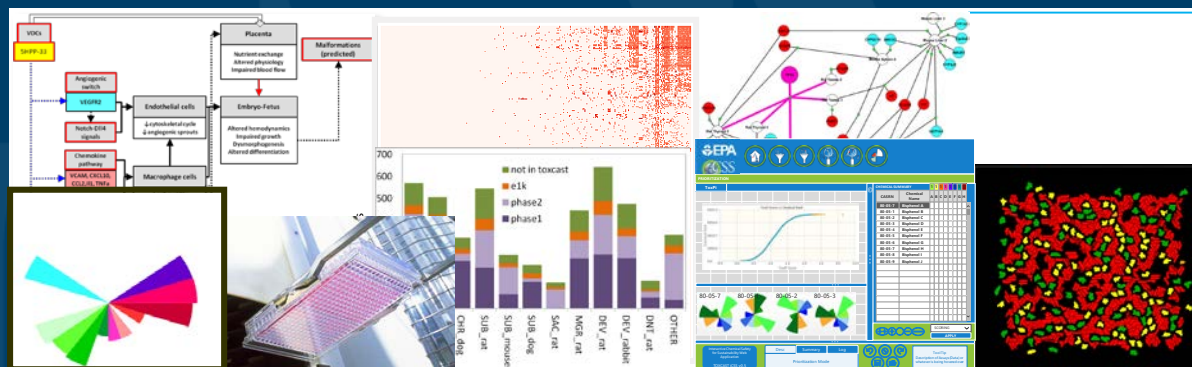


# Predictive Toxicology: Current Status and Future Outlook



EBI-EMBL Industry Programme Workshop  
Predictive Toxicology

November 16-17, 2016

Imran Shah

National Center for Computational Toxicology

*The views expressed in this presentation are those of the author[s] and do not necessarily reflect the views or policies of the U.S. Environmental Protection Agency.*

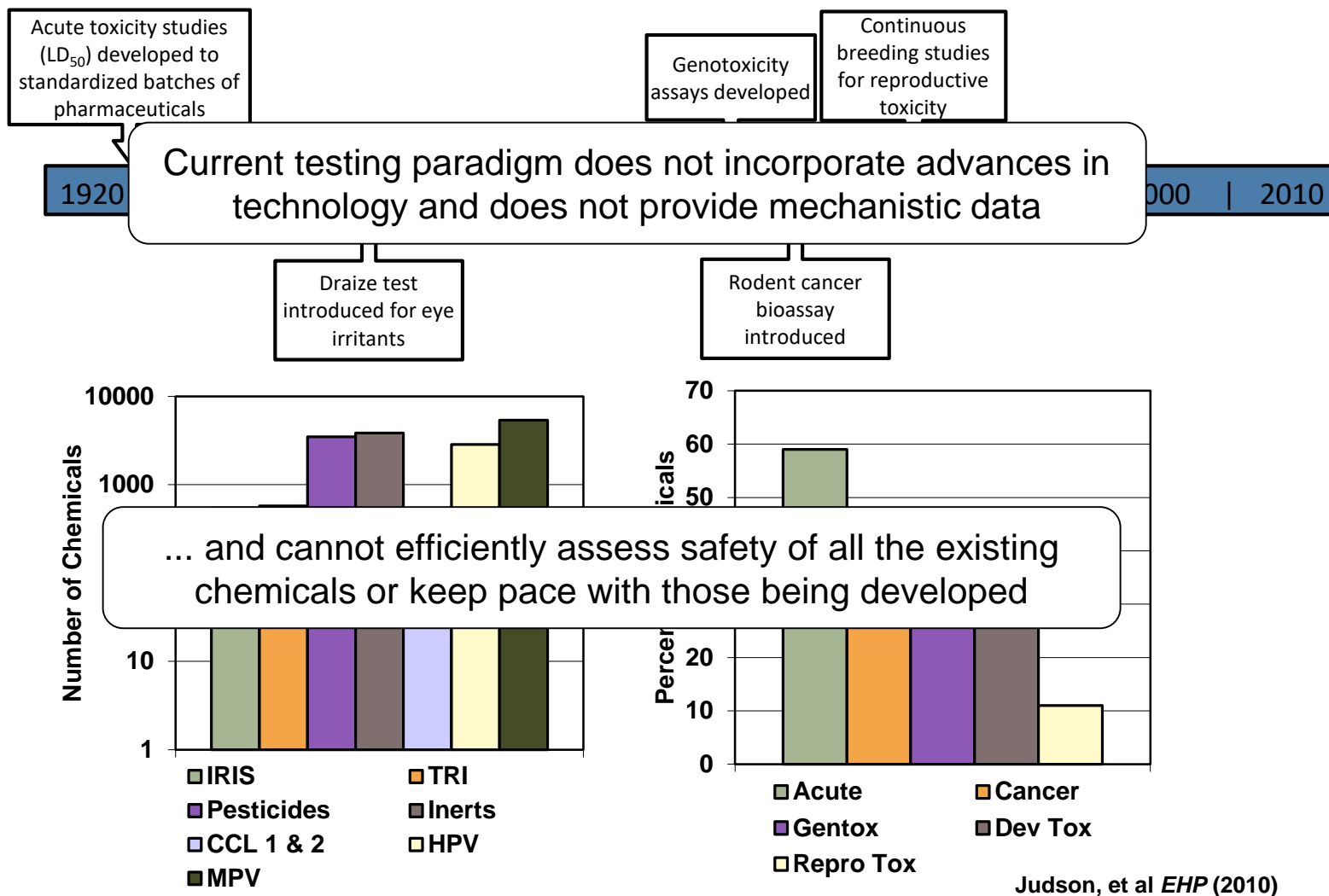
# Regulatory Agencies Need to Make A Range of Decisions on Chemicals...

- Multiple drivers shape type of assessment
  - Regulatory scope
  - Economic considerations
  - Multiple applications
- Chemical assessments are “fit-for-purpose”
  - Prioritization (e.g., EDSP, PMN, SNUR)
  - Screening-level assessments (e.g., CCL, GreenChem)
  - Provisional assessments (e.g., PPRTVs)
  - Toxicity assessments (e.g., IRIS)
  - Risk assessments (e.g., MCLs, pesticides)

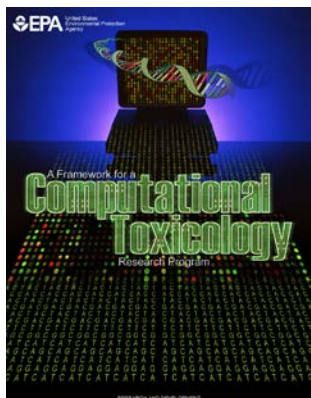
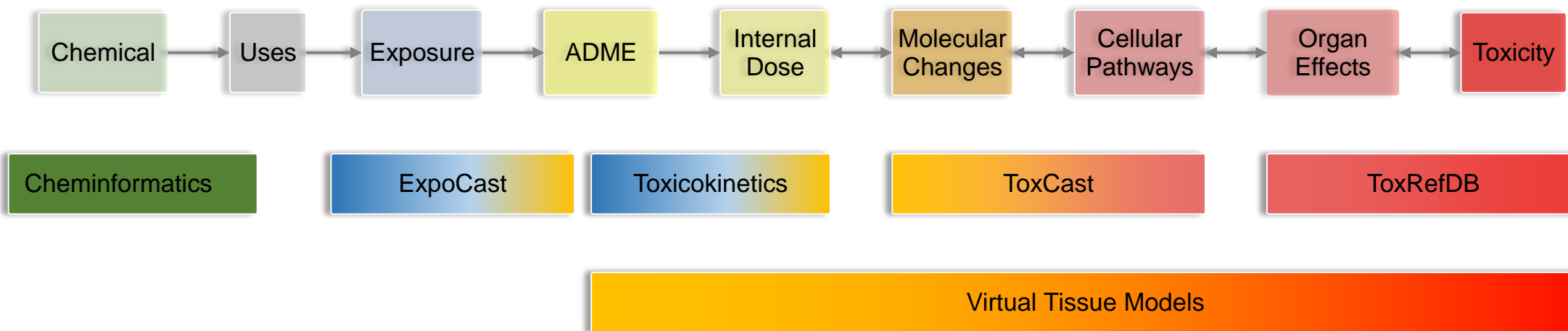
EPA	Workflow	Throughput	Data
OPPT	New chemicals: Premanufacture Notice (PMN)	~1000/yr (90d/chem)	III (II)
	Existing chemicals: Significant New Use Rule (SNUR)	~84,000 total	
	Current Chemical Risk Assessments	~10 total	I
	DFE / Green Chemistry	~2500	I, II, III
OPP	Pesticide registration (PR)	~10 new/yr ~50 old/yr	I
	Pesticide re-registration	~1000/yr 24,576 total	I
OW	Chemical Contaminant List	6yr / ~6,000 total	I, II, III
	RegDet on CCL	Every 6yr / 90 total	I
	Unregulated Contaminant Monitoring	30/5yr	I
	Drinking Water Health Advisories		II, III
OLEM	Spills Brownfields Super Fund		

I Guideline animal testing data  
 II Some *in vitro* bioactivity  
 III Chemical structure data

# Current System to Evaluate Chemicals is Antiquated and Inefficient

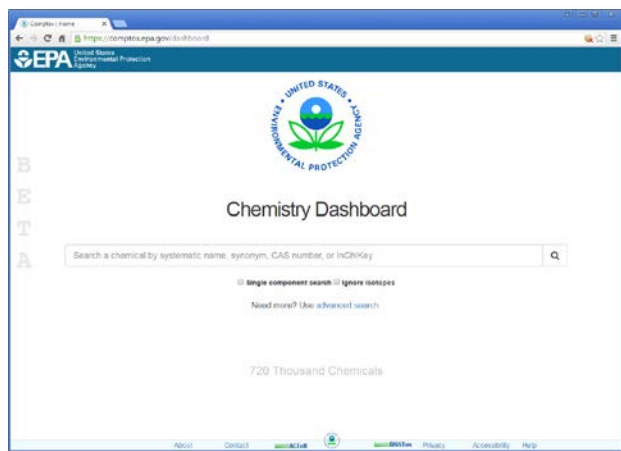


# Requires an Integrated and Multidisciplinary Solution

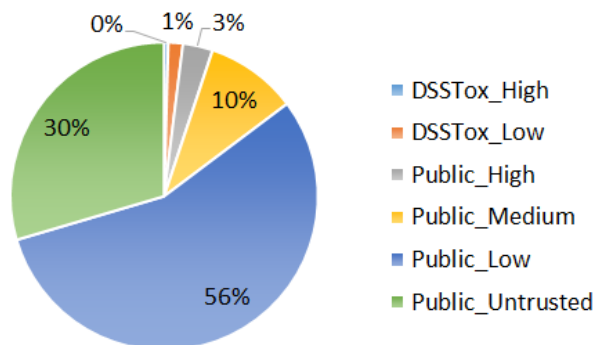


- National Center for Computational Toxicology established in 2005
- Currently staffed by ~60 employees
- Exists within the EPA's Office of Research and Development
- Home of the ToxCast and ExpoCast research efforts
- Key partner in U.S. Tox21 federal consortium

# Need to Start with a High Quality Chemistry Foundation

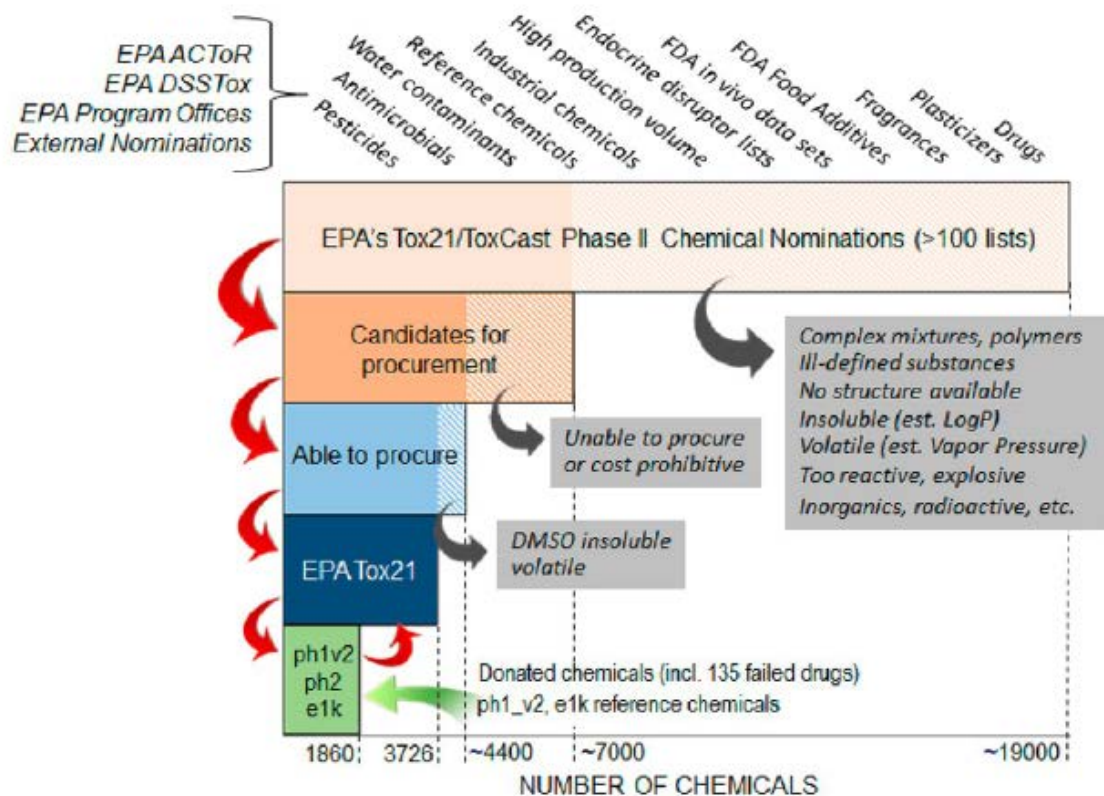


<https://comptox.epa.gov>

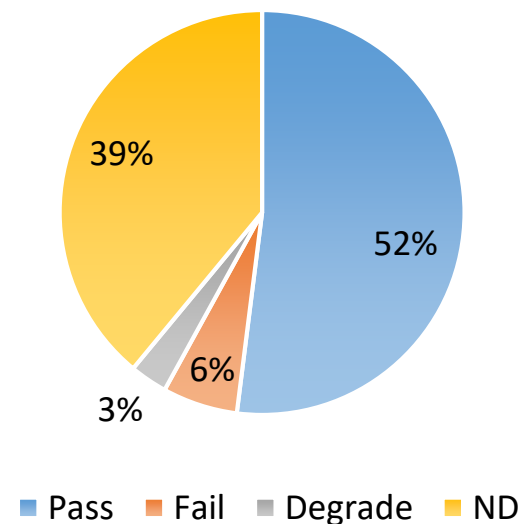


- Developing a centralized resource for curated chemical structure, identifier, and physical chemical properties of >700K unique substances with data quality flags
- Expand and curate training sets for QSAR models for phys-chem, environmental fate, and toxicological properties
- Use the centralized chemical resource as the foundation for an integrated hazard, bioactivity, pharmacokinetics, and exposure information

# Need to Start with a High Quality Chemistry Foundation



## Analytical QC of Chemical Library



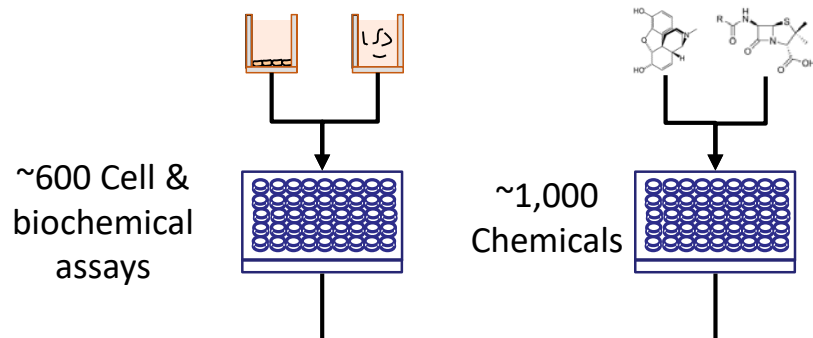
Pass = C (75%) or greater

Fail = D, F, Ac, Bc, Cc

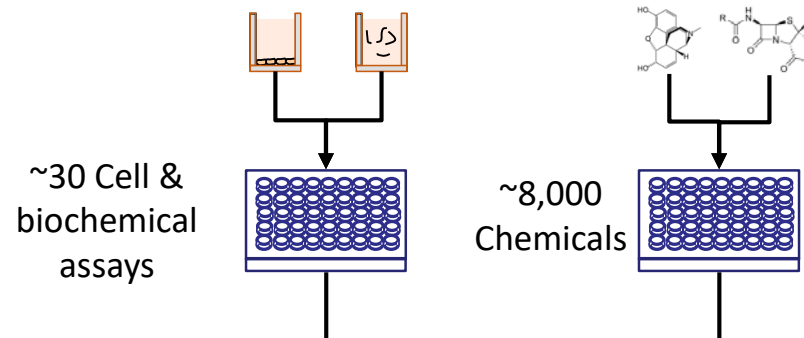


# High-Throughput Bioactivity Screening

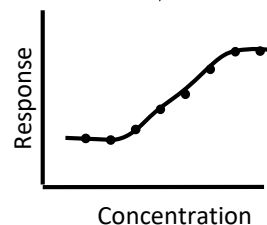
## ToxCast



## Tox21



Set	Chemicals	Assays	Completion
ToxCast Phase I	293	~600	2011
ToxCast Phase II	767	~600	2013
ToxCast Phase III	1001	~100	Ongoing
E1K (endocrine)	880	~50	2013



# ToxCast Incorporates a Diverse Array of High-Throughput *In Vitro* Assays

## Assay Provider

ACEA  
Apredica  
Attagene  
BioReliance  
BioSeek  
CeeTox  
CellzDirect  
Tox21/NCATS  
NHEERL MESC  
NHEERL Zebrafish  
NovaScreen (Perkin Elmer)  
Odyssey Thera  
Vala Sciences

## Biological Response

cell proliferation and death  
cell differentiation  
Enzymatic activity  
mitochondrial depolarization  
protein stabilization  
oxidative phosphorylation  
reporter gene activation  
gene expression (qNPA)  
receptor binding  
receptor activity  
steroidogenesis

## Target Family

response Element  
transporter  
cytokines  
kinases  
nuclear receptor  
CYP450 / ADME  
cholinesterase  
phosphatases  
proteases  
XME metabolism  
GPCRs  
ion channels

## Assay Design

viability reporter  
morphology reporter  
conformation reporter  
enzyme reporter  
membrane potential reporter  
binding reporter  
inducible reporter

## Readout Type

single  
multiplexed  
multiparametric

## Cell Format

cell free  
cell lines  
primary cells  
complex cultures  
free embryos

## Species

human  
rat  
mouse  
zebrafish  
sheep  
boar  
rabbit  
cattle  
guinea pig

## Tissue Source

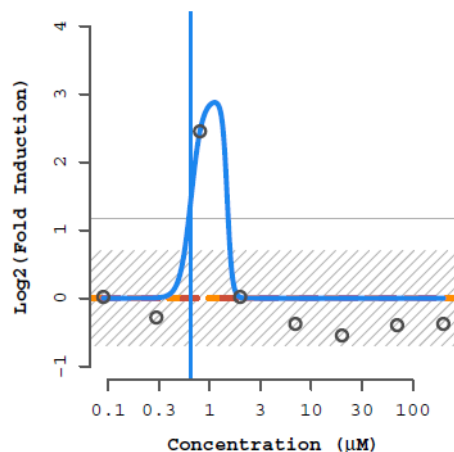
Lung	Breast
Liver	Vascular
Skin	Kidney
Cervix	Testis
Uterus	Brain
Intestinal	Spleen
Bladder	Ovary
Pancreas	Prostate
Inflammatory	Bone

## Detection Technology

qNPA and ELISA  
Fluorescence & Luminescence  
Alamar Blue Reduction  
Arrayscan / Microscopy  
Reporter gene activation  
Spectrophotometry  
Radioactivity  
HPLC and HPEC  
TR-FRET



# Efforts to Ensure HTS Data Quality and Increase Transparency



ASSAY: ARID117 (ATQ\_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	GNLS
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

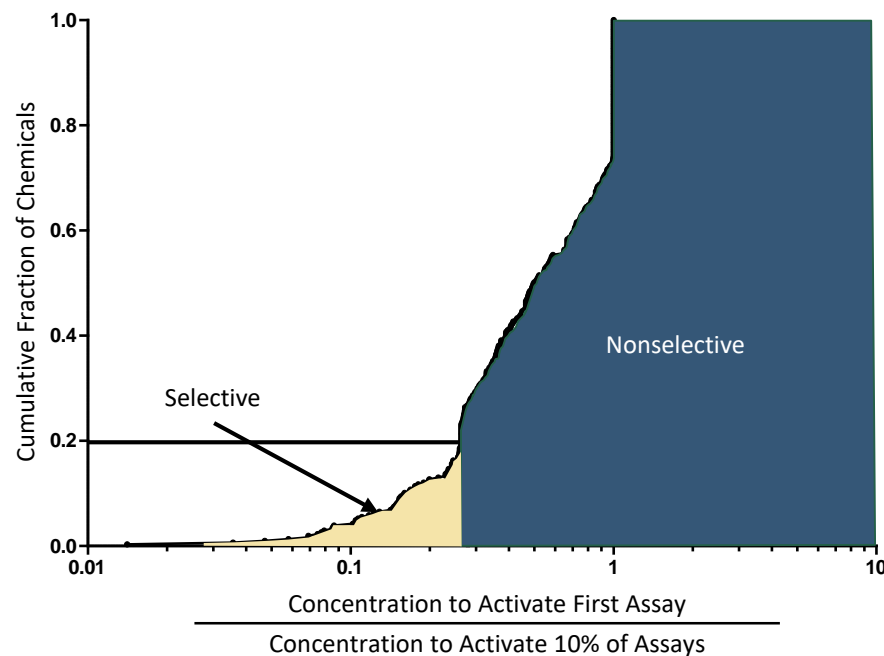
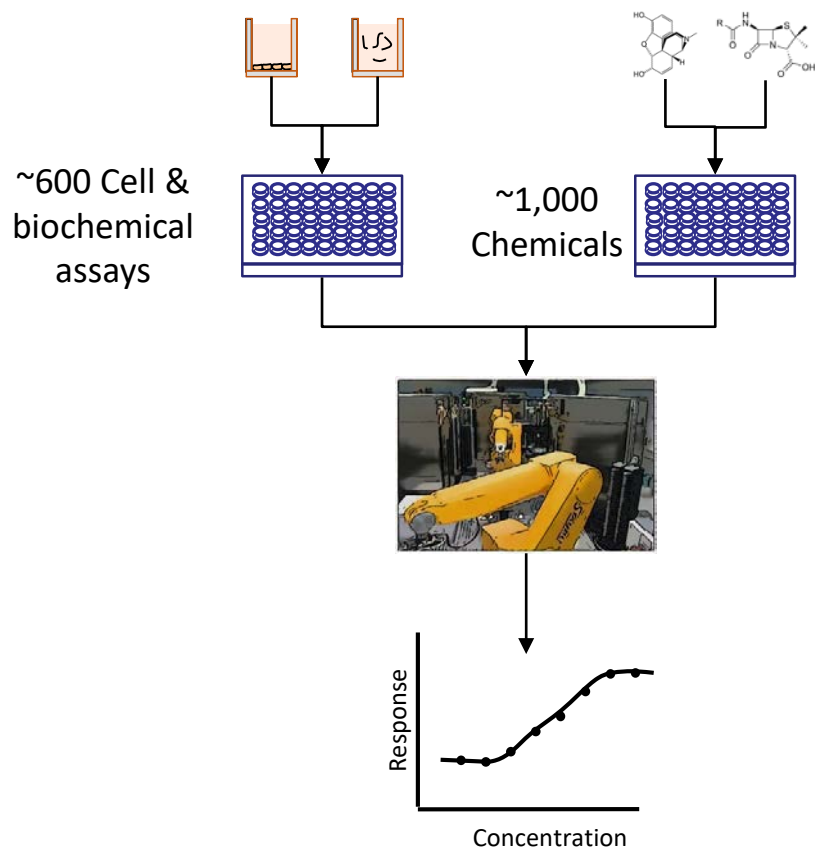
FLAGS:

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)
- Transparent ToxCast data analysis pipeline
  - Data quality flags to indicate concerns with chemical purity and identity, noisy data, and systematic assay errors
  - Publicly available as an R package
- 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

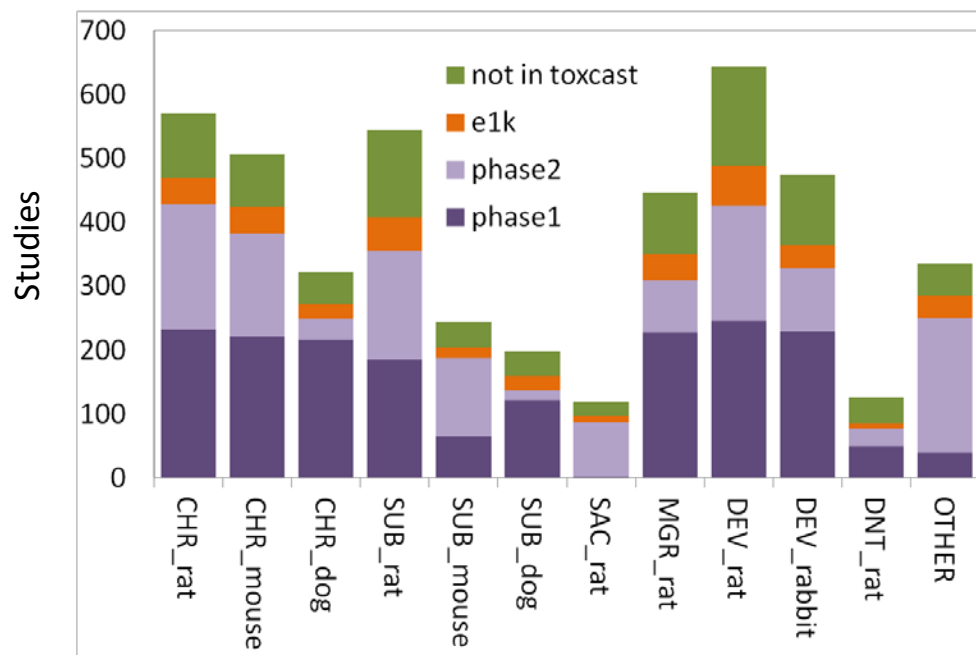
# Most Environmental Chemicals are Nonselective for Biological Targets

## ToxCast



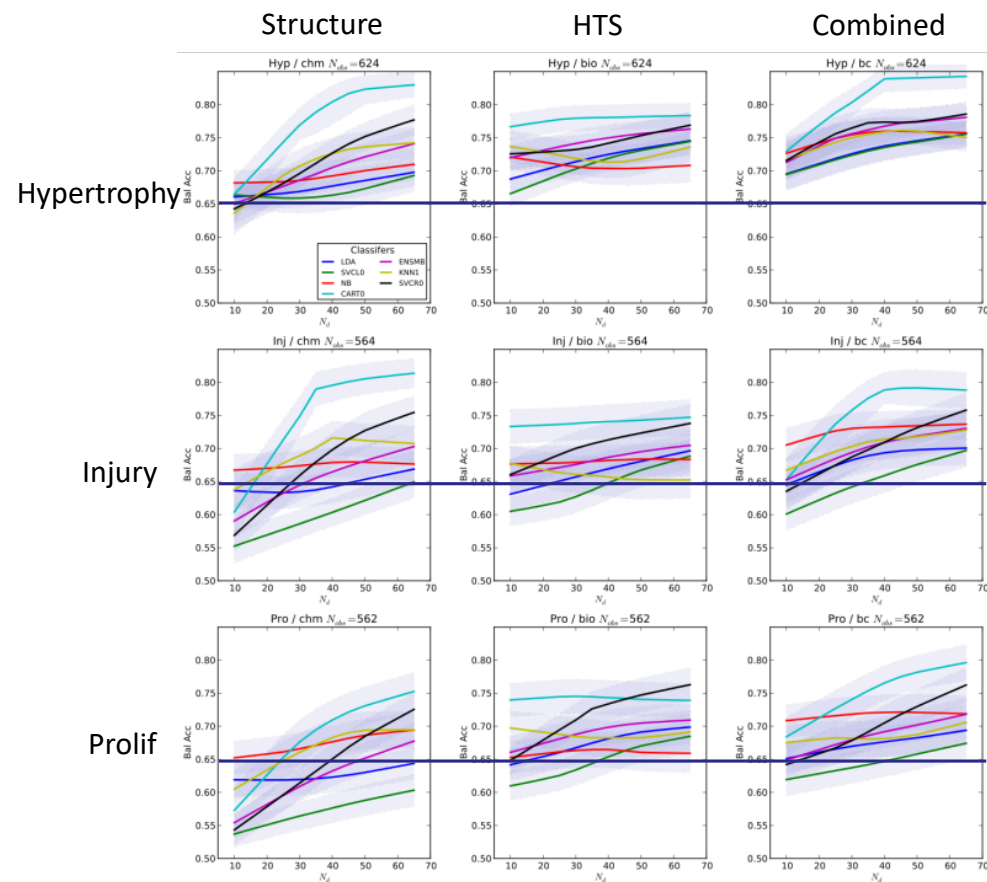
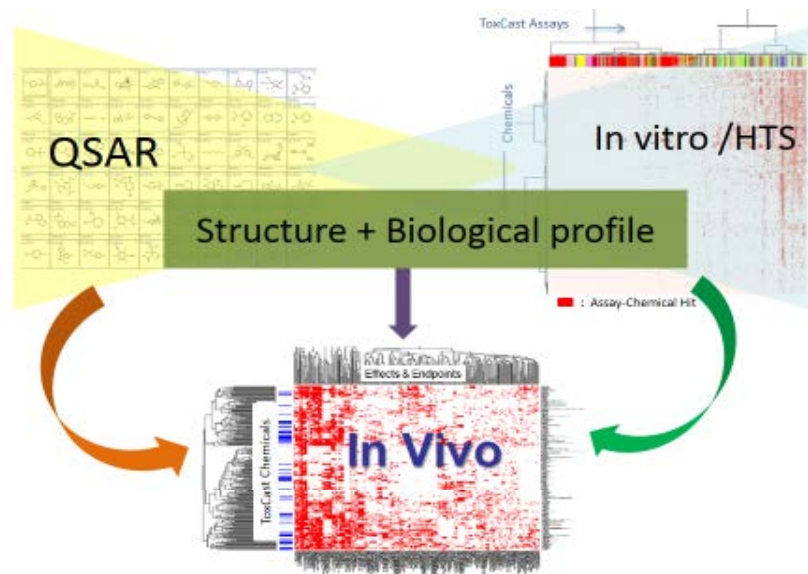
# ToxRefDB: Digitizing Legacy *in vivo* Toxicology Data

- ToxRefDB holds *in vivo* endpoint data from animal toxicology studies (DERs, NTP, open literature, pharma)
- Currently at 5567 studies on 1049 unique chemicals

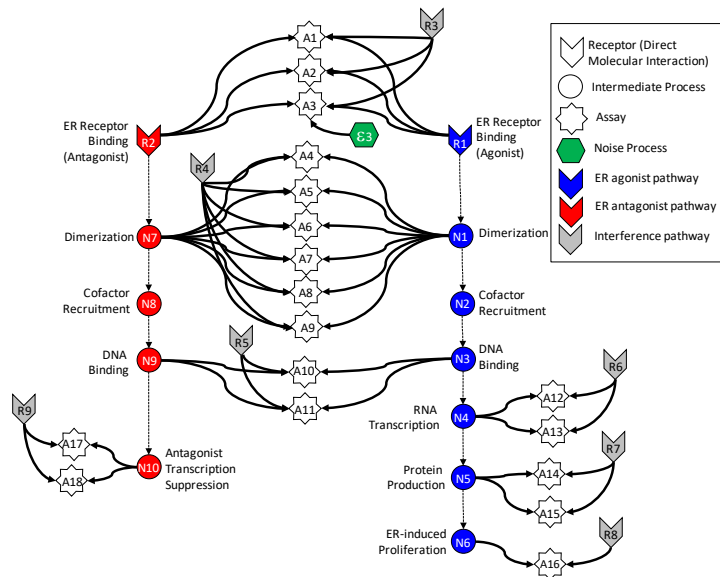


Data Source	Study Count
EPA OPP DER	3279
Open Literature	731
National Toxicol Program	666
Sanofi_Pharma	222
Unpublished Submissions	50
GSK Pharma	38
Health Canada PMRA DER	23

# Predicting Target Organ Toxicities by Machine Learning



# Developing a Pathway Model to Predict Endocrine Activity



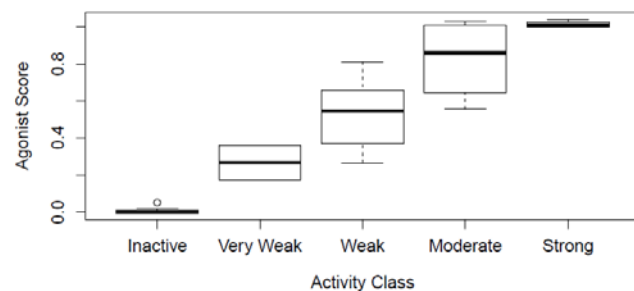
Performance for *In Vitro* Reference Chemicals

# True Pos	28
# True Neg	12
# False Pos	0
# False Neg	4
PPV	1.0
NPV	0.75
BA	0.94
Sensitivity	0.88
Specificity	1.0

Performance for *In Vivo* Uterotrophic Studies

# True Pos	28
# True Neg	12
# False Pos	1
# False Neg	1
PPV	0.97
NPV	0.92
BA	0.95
Sensitivity	0.97
Specificity	0.92

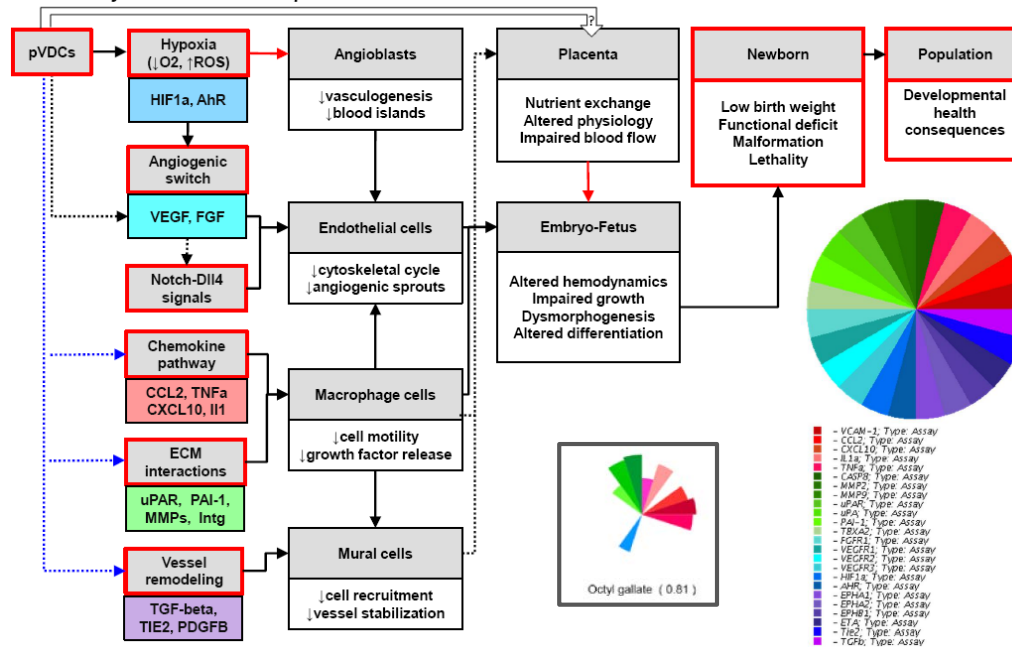
ER Pathway Model Integrating 18 *In Vitro* Assays



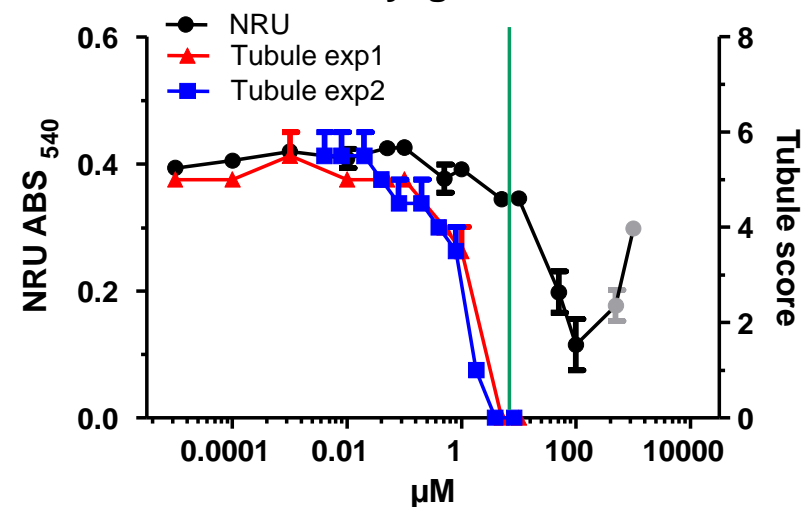
Judson *et al.*, ToxSci (in press)

## ER Active Hit Rate

1431 EDSP chemicals run *in vitro*  
71 (5%) have a significant ER score



## Octyl gallate

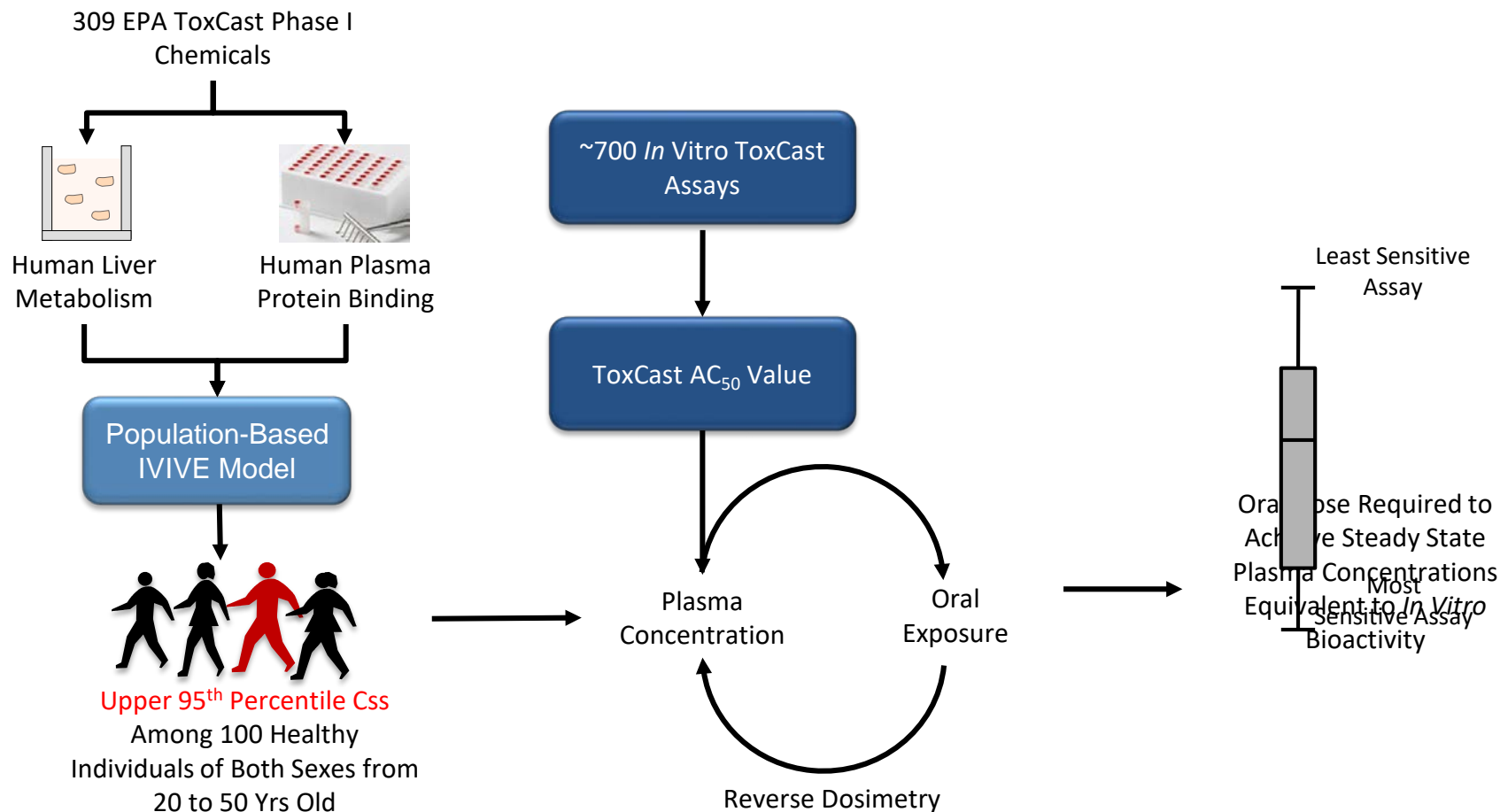


## Cell ABM of Octyl Gallate (NCCT: G Nagaraj)

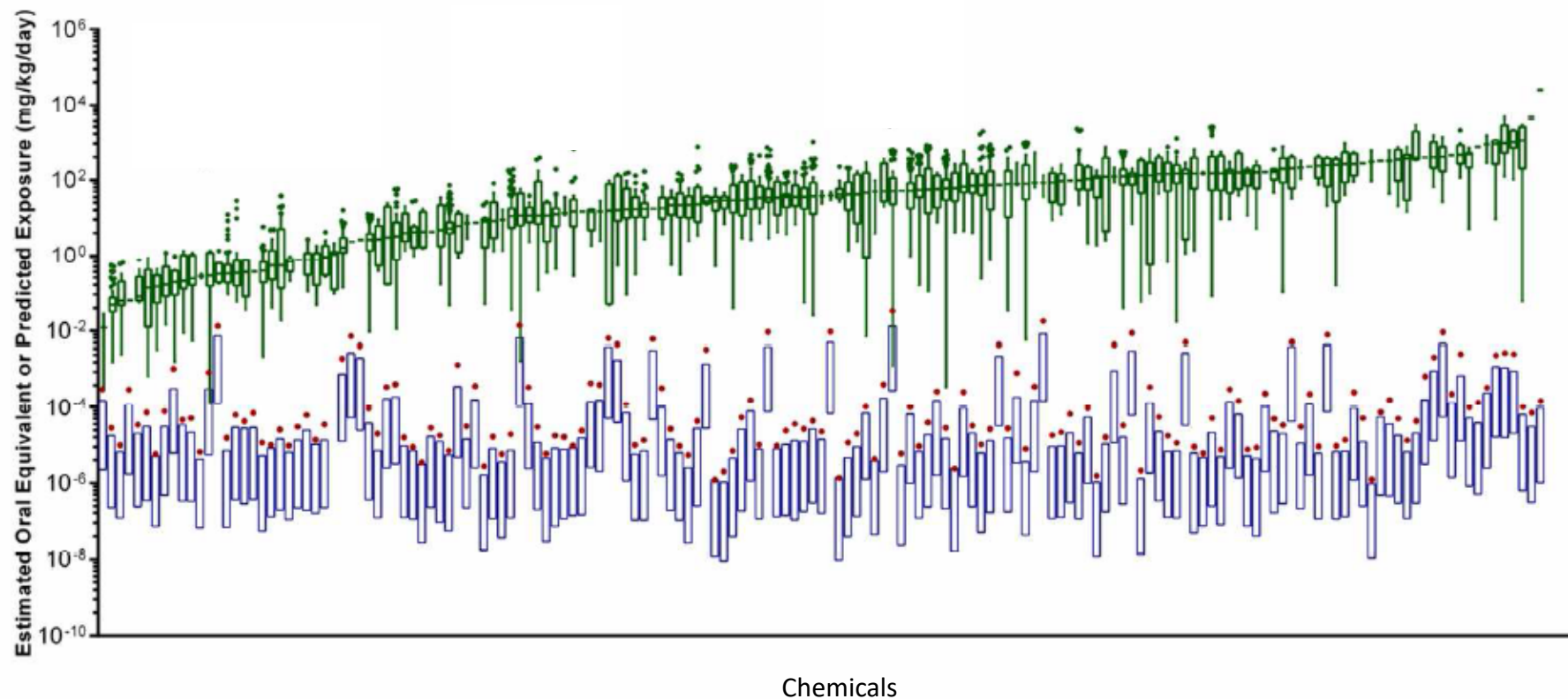




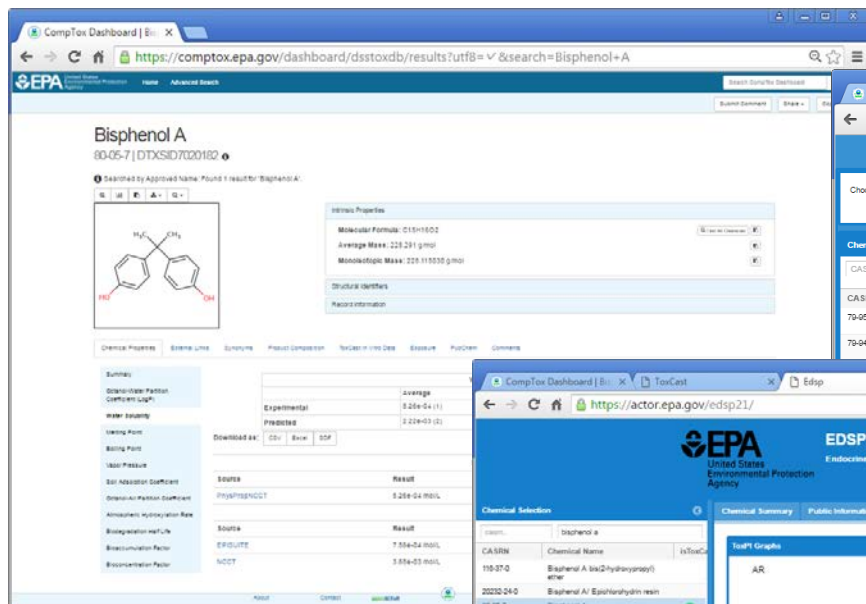
# Providing Context by Incorporating Toxicokinetics and Exposure



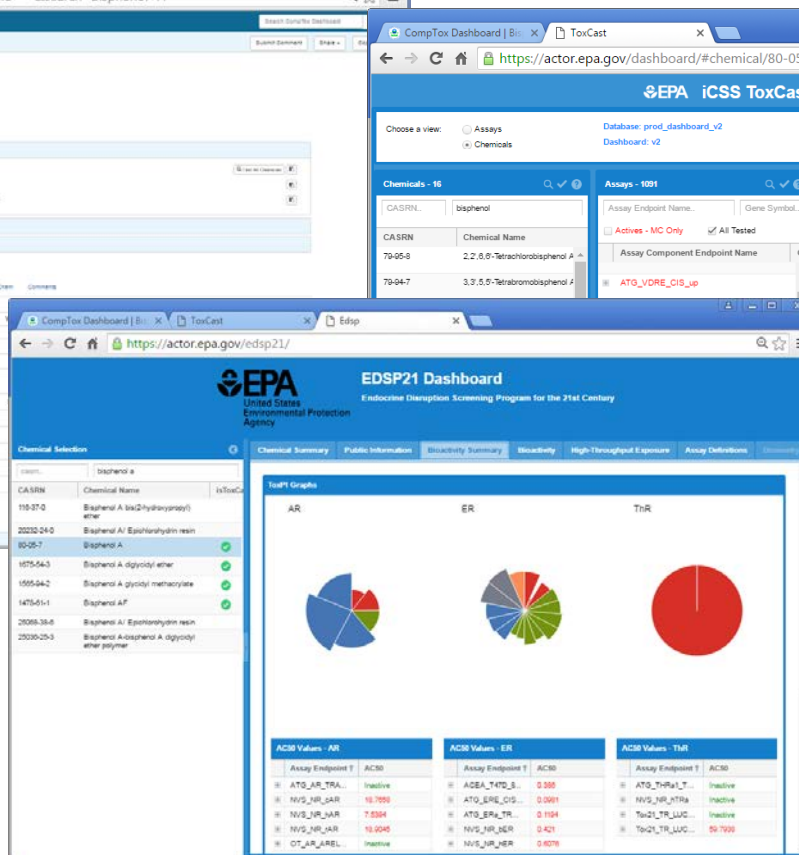
# Comparing Bioactivity with Exposure Predictions for Risk Context



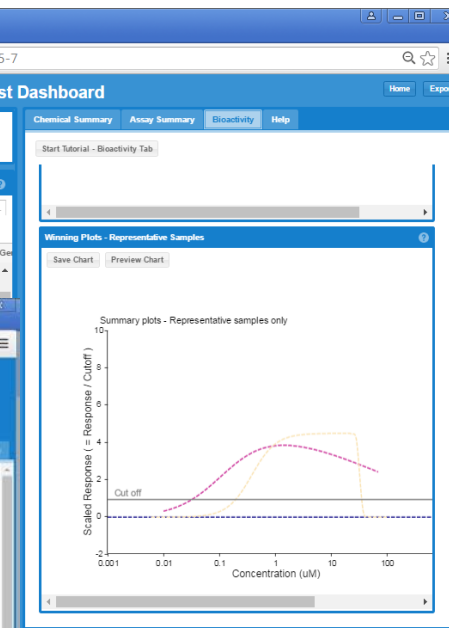
# Delivering Data to Stakeholders and Scientific Community



Chemistry Dashboard  
(<https://comptox.epa.gov/dashboard/>)

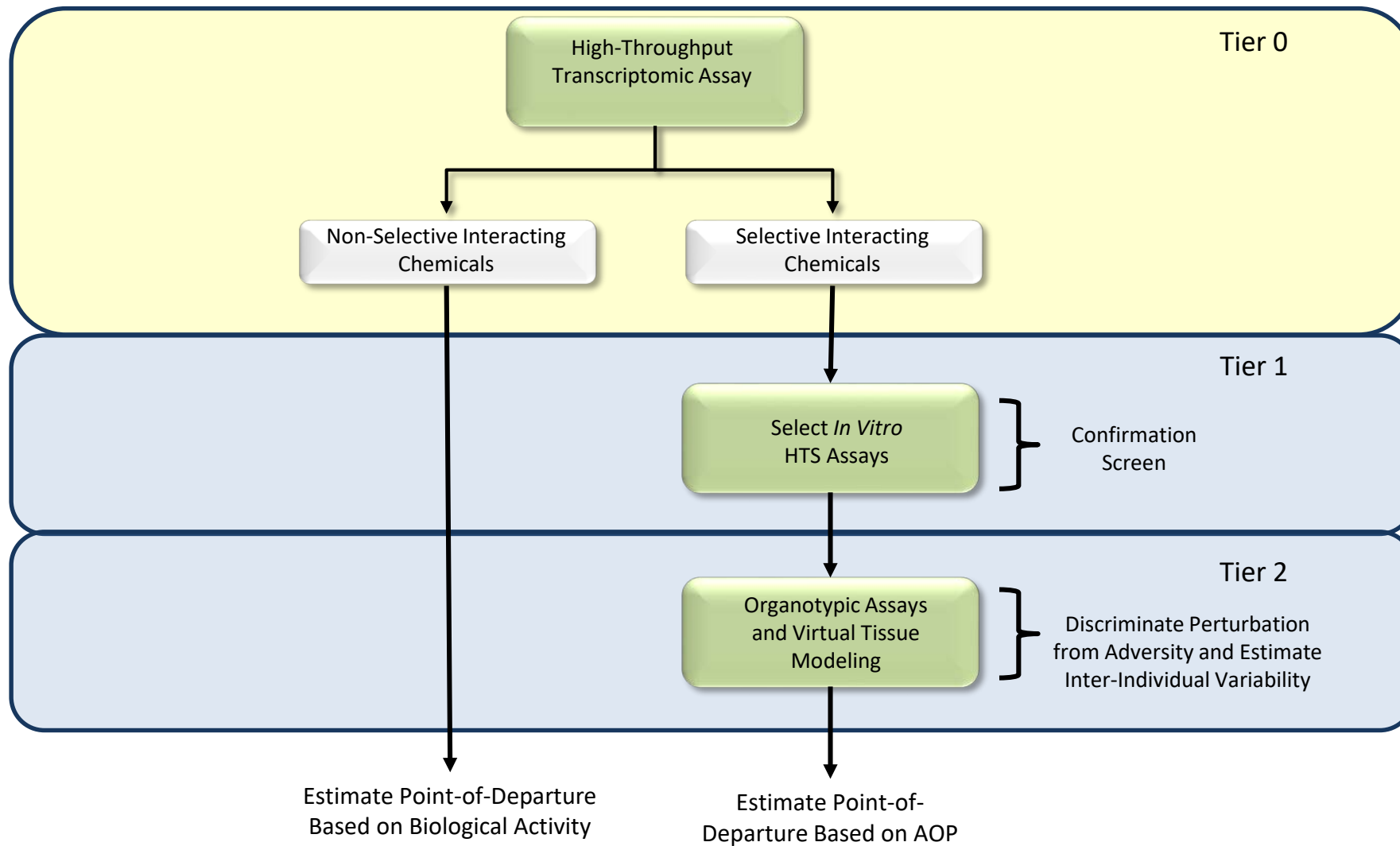


EDSP21 Dashboard  
(<https://actor.epa.gov/edsp1>)



ToxCast Dashboard  
(<https://actor.epa.gov/dashboard/>)

# Developing a Broad Hazard Screening Platform

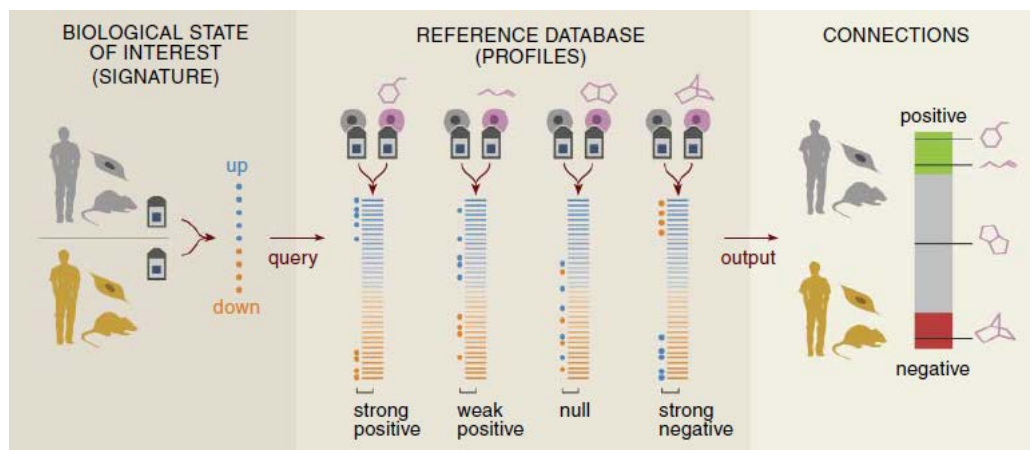


# How Would a HT Transcriptomic Platform be Deployed?

## High-Throughput Transcriptomic Assay

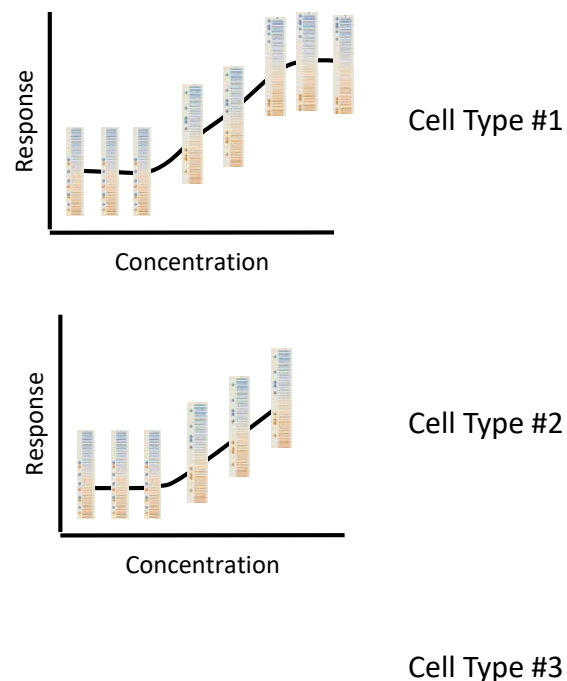
Tier 0

- **Identify predominant mechanisms as a function of concentration**
- Group chemicals by similar mechanism/bioactivity
- Identify a concentration that results in no transcriptional effects



Lamb et al. *Science* (2006)

Broad CMAPdb: 7,000 profiles; 1,309 compounds  
NIH LINCS CMAPdb: 9,000 shRNAs, 3,000 over  
expression ORFs, and 4,000 compounds in 20 cell  
types/lines (cell lines and primary cells)



...

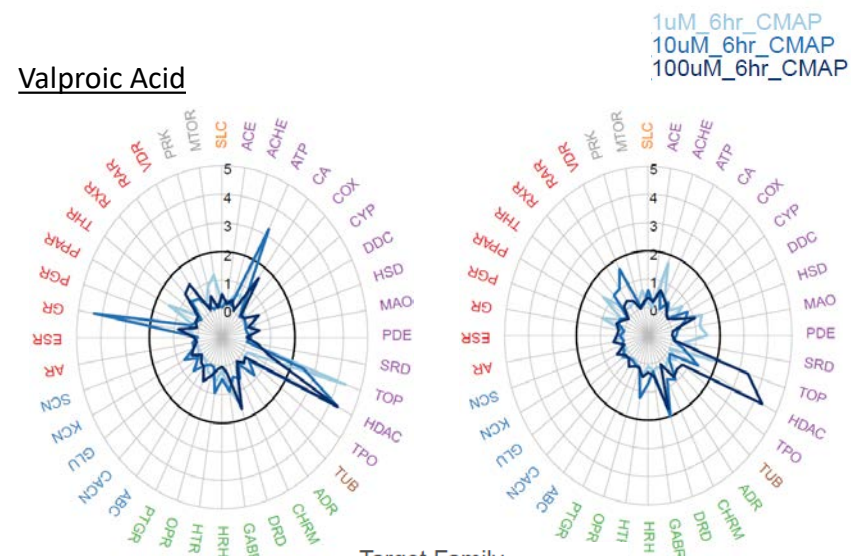
Cell Type #3

# Using HT Transcriptomics to Identify Mode-of-Action

Target Family	Total Profiles	Target Genes	Chemicals	Cell Lines
Cytokine receptors	3	1	1	3
Enzymes	336	40	112	5
Exosome	14	1	4	4
G protein-coupled receptors	585	16	192	4
Ion channels	194	8	65	3
Nuclear receptors	227	10	71	5
Protein kinases	19	8	6	4
Transporters	102	2	35	3

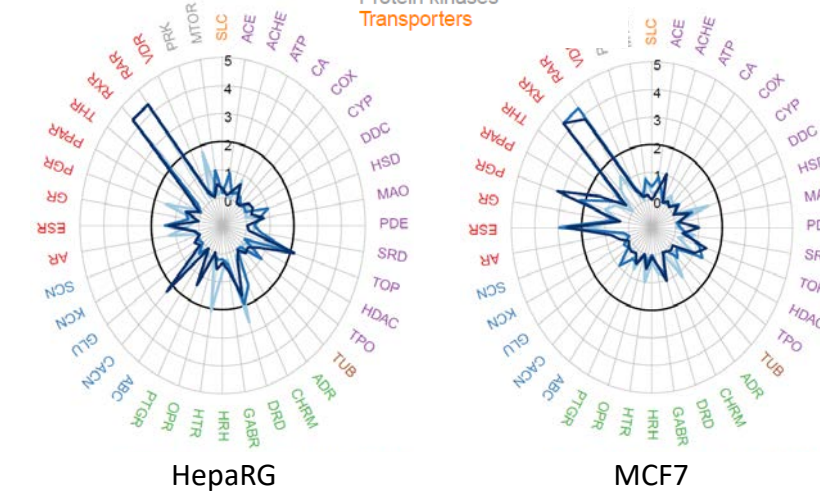
- Developed local database of Broad's CMAP data (~3,000 profiles)
- Annotated targets using KEGG (1,571 profiles)
- Significant genes identified using a z-score cutoff of 2
- Incorporated "JG" scoring method (Jiang and Gentleman 2007)
- Determine significance using a permuted rank approach across target family

## Valproic Acid



Target Family  
Enzymes  
Exosome  
G Protein-coupled receptors  
Ion channels  
Nuclear receptors  
Protein kinases  
Transporters

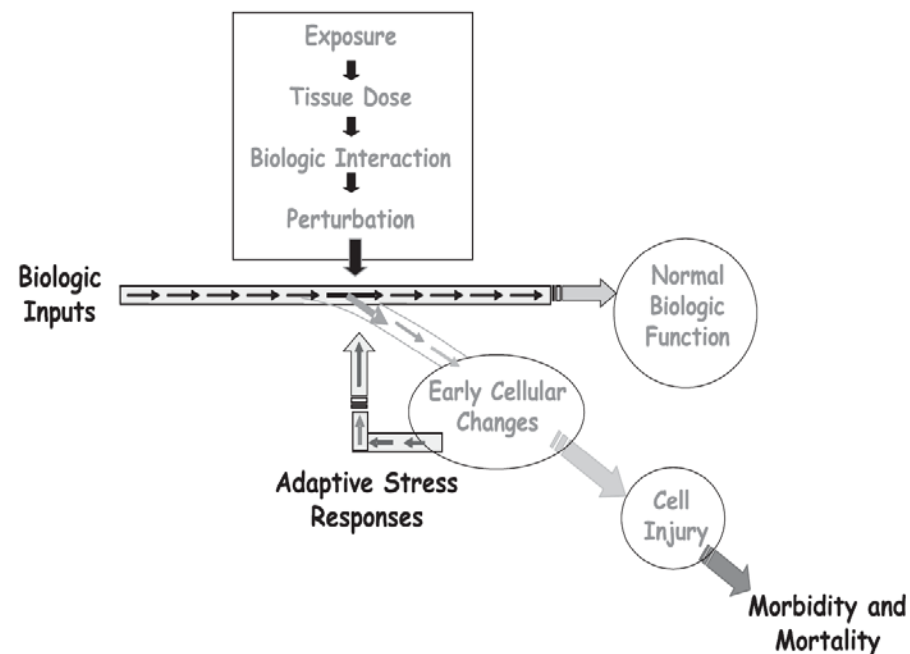
## Retinoic Acid





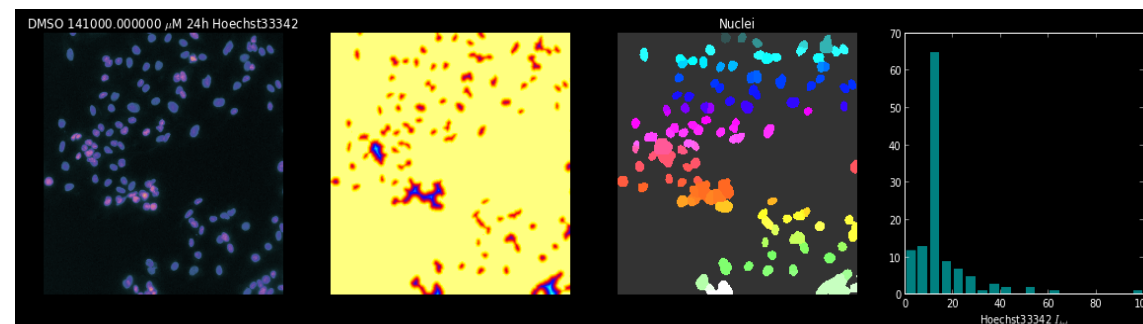
# Distinguishing Adaptation from Adversity

- ❑ **Tipping Point:** Threshold between adaptation and adversity



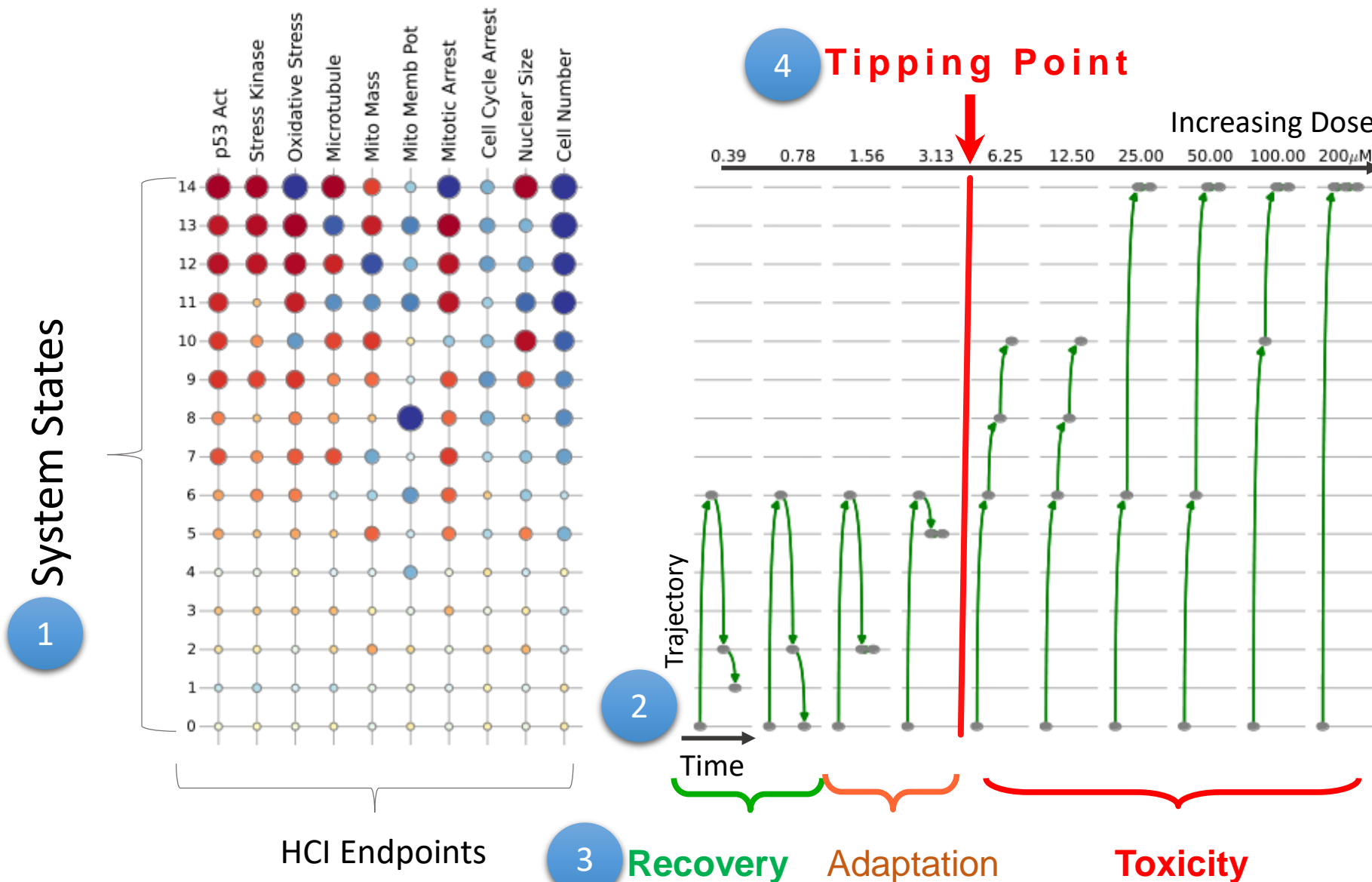
- ❑ Can we use **Tipping Point** to define a point of departure (PoD) for risk assessment ?

- ❑ Use ToxCast High Content Imaging (HCI) data to identify Tipping Points



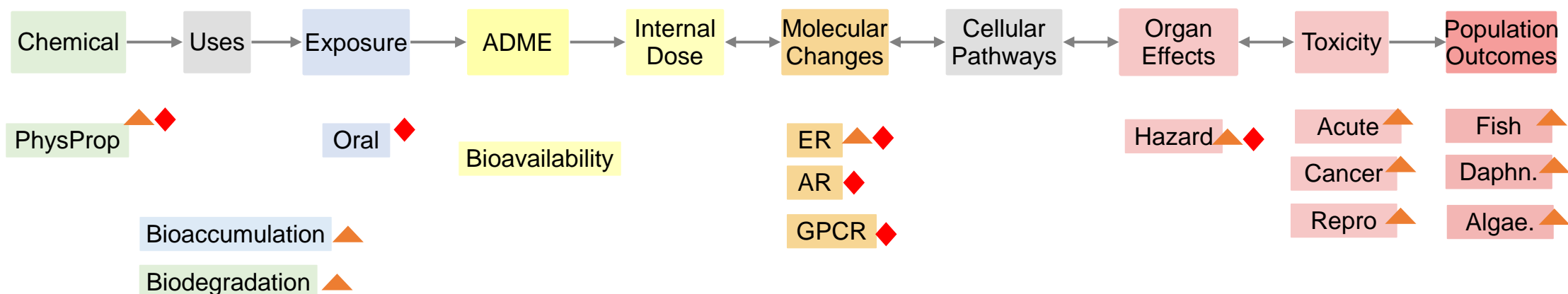
- 967 chemicals (ToxCast)
- HepG2 cells culture
- 10 concentrations
- 3 Time points
- 10 HCI Assays
- 400 plates
- 100,000 wells
- 2,400,000 images

# Tipping Point Analysis



# Thousands of chemicals have limited data!

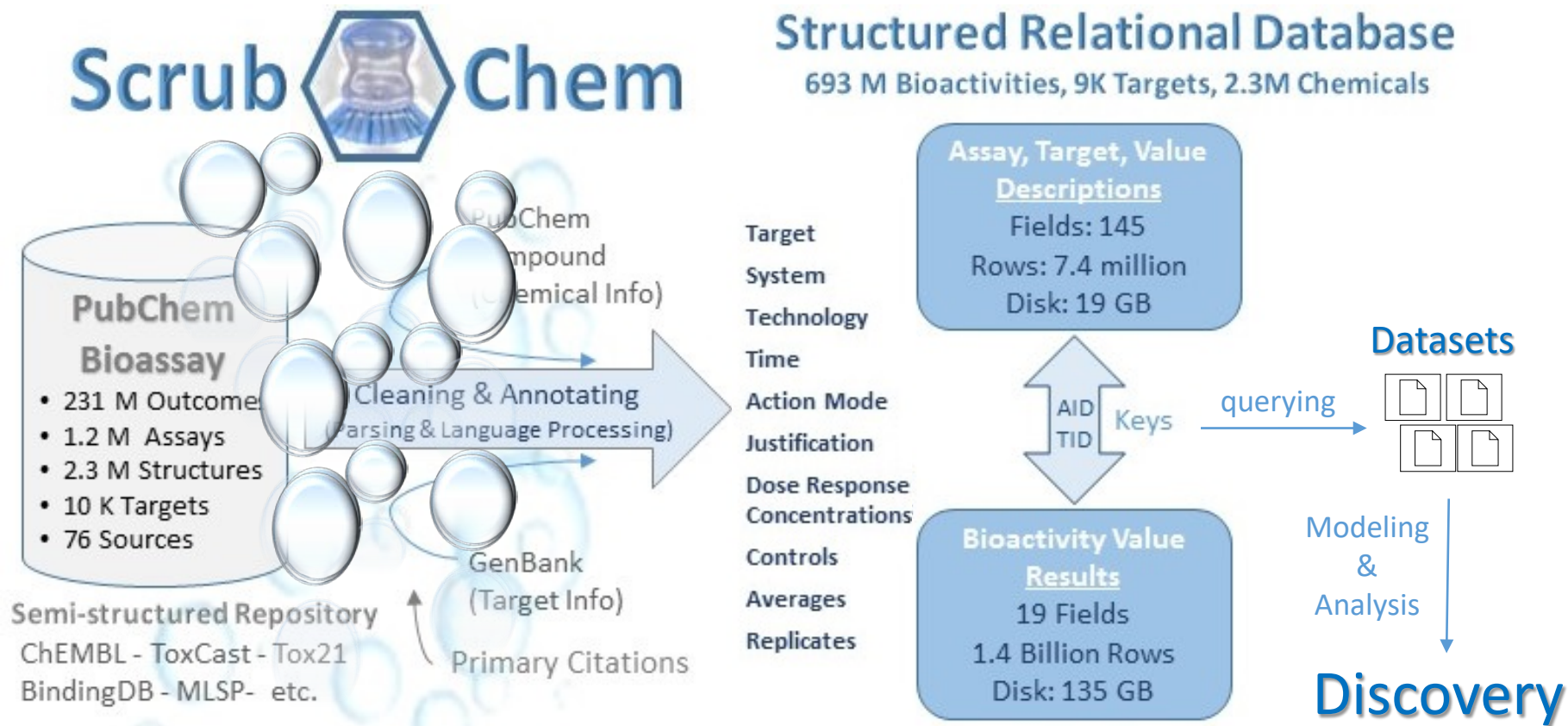
- Tens of thousands of environmental chemicals have very limited exposure/biological data
  - Need more effective tools to describe chemical properties, effects and linkages
  - Need predictive models to fill data gaps



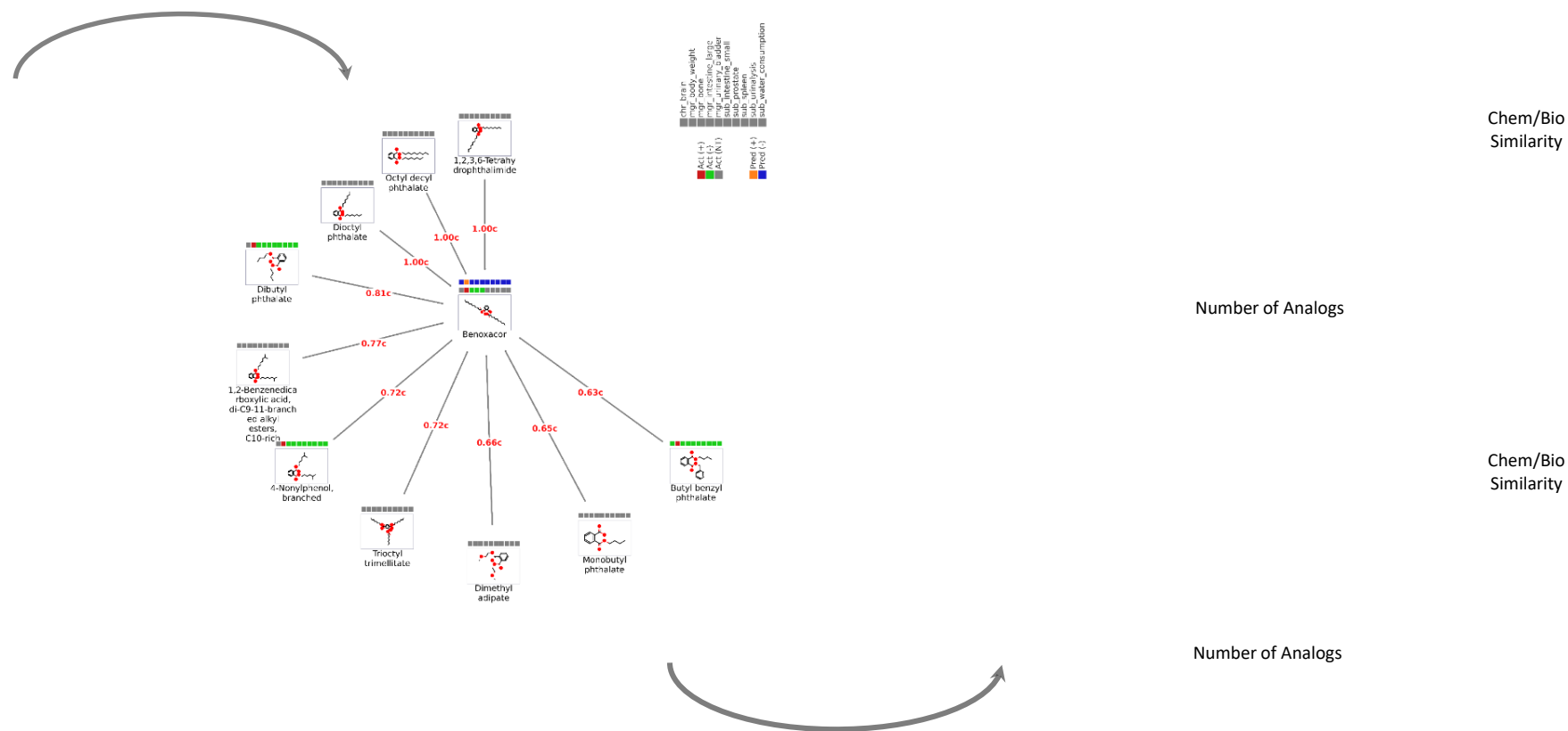
▲ existing/legacy tools: TIMES, LeadScope, ECOSAR, EPIWIN

◆ developed internally: Read-across/GenRA, Machine Learning: classification and regression

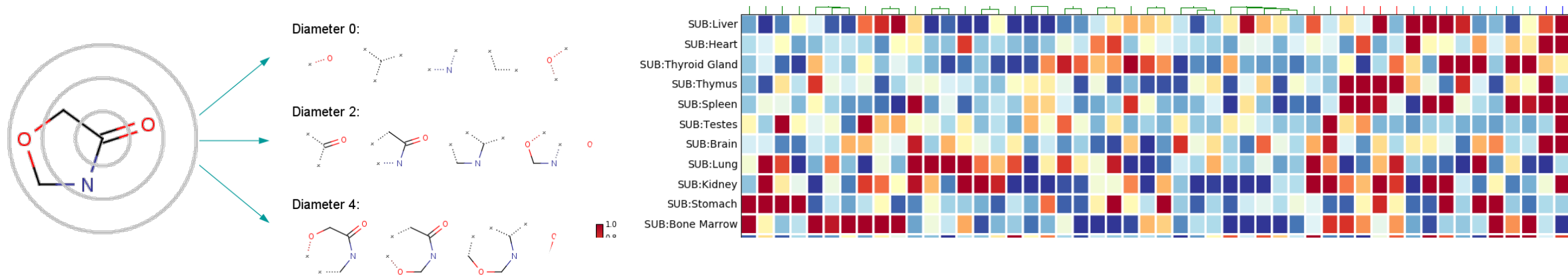
# Structuring PubChem Data for Analysis



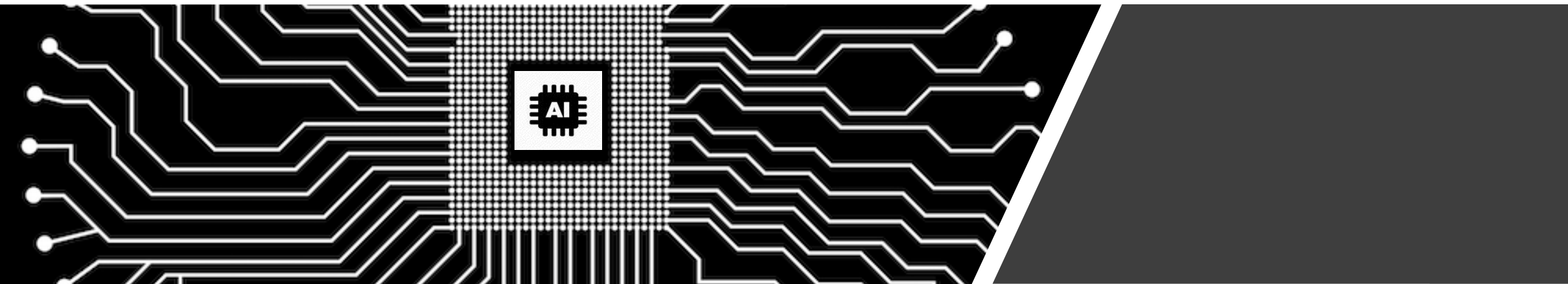
Uncertainties can be evaluated  
across the local neighbourhoods



Shah, et al. RTP 2016

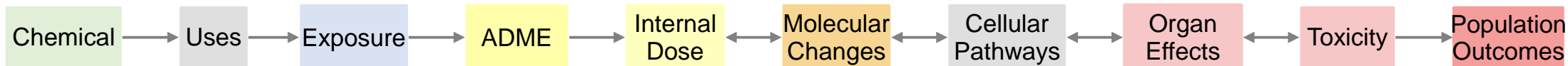


# Machine Learning to Predict Chemical Effects

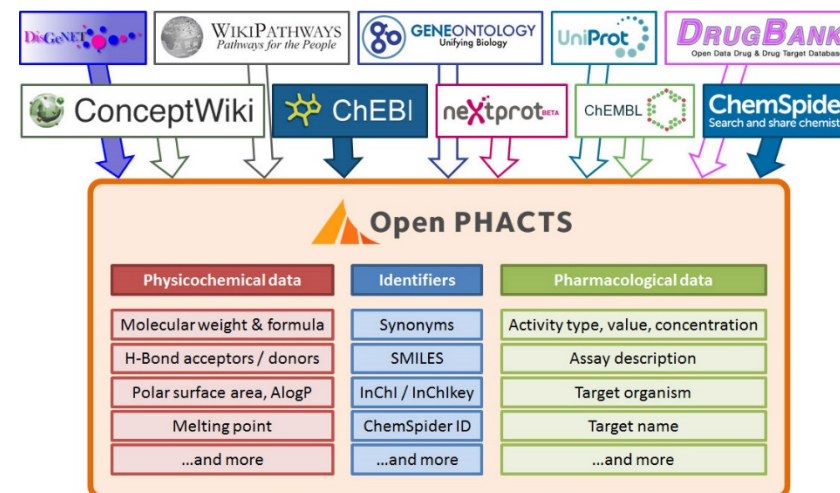
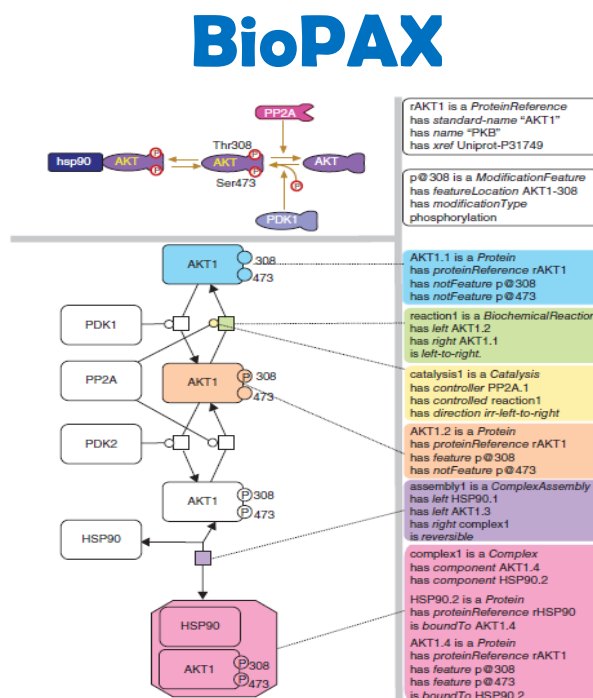
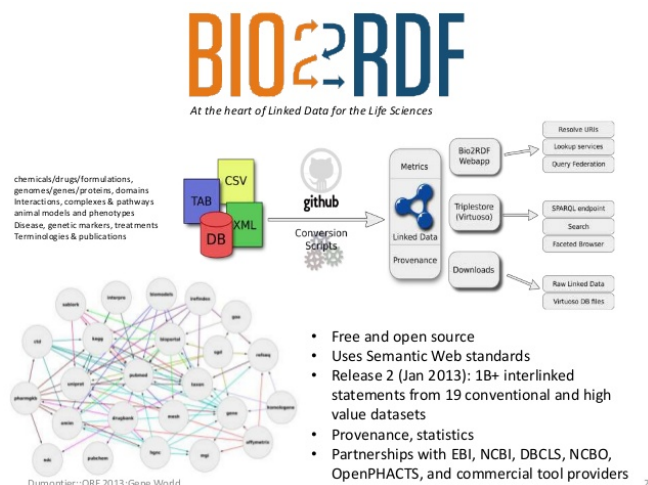




# Use Semantic Tools to Link Disparate Data



Can we use semantic tools (OWL/RDF, existing ontologies) to meaningfully integrated disparate resources ?



# Data Challenges

- Transparently sharing complex data streams – adequately capturing chemical (treatment dose and time), biological (experimental modal, assay, etc.) context to ease re-use
- Systematically integrating disparate data streams – representing linkages across molecular, cellular, tissue, organs. This is vital for relating early molecular changes to adverse (e.g. histopathological) outcomes
- Effectively extracting evidence from unstructured textual data – the literature is one of the largest resource for information about apical outcomes
- Using linked data to better discriminate between adaptation vs adversity – predicting which molecular markers lead to apical outcomes
- Quantifying and incorporating uncertainty and variability in predictions
- Legal defensibility of new methods and assessment products

# Acknowledgements and Questions

## Tox21 Colleagues:

NTP Crew

FDA Collaborators

NCATS Collaborators

## EPA Colleagues:

NERL

NHEERL

NCEA



EPA's National Center for Computational Toxicology