

USING HIGH THROUGHPUT SCREENING FOR PREDICTIVE MODELING OF REPRODUCTIVE TOXICITY

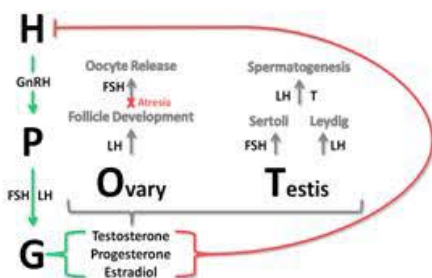


Matt Martin
August 22nd, 2011

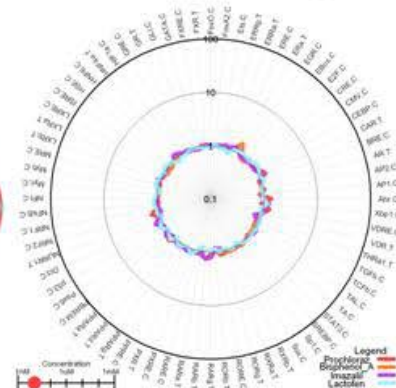
USEPA – Office of Research & Development
National Center for Computational Toxicology

Overview

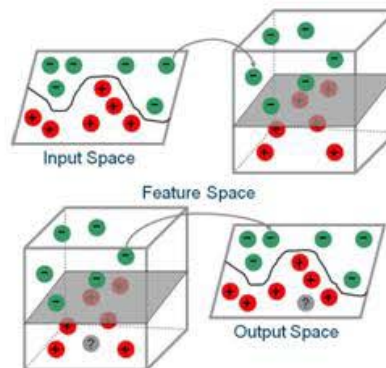
Reproductive Physiology & Impairment



Transcription Factor Profiling



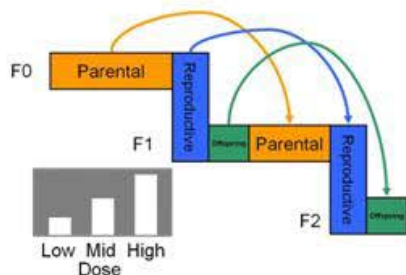
Classification Modeling of Toxicity



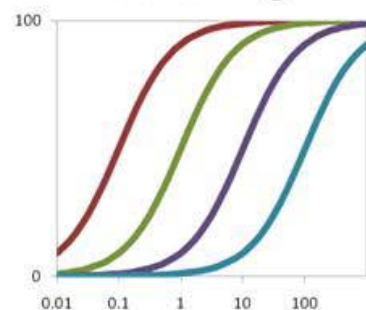
Forward Validation Study



Reproductive Toxicity Testing



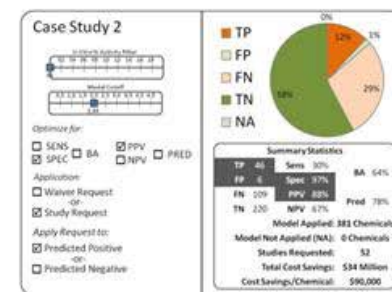
Quantitative High Throughput Screening



Predictive Model of Reproductive Toxicity



Case Studies in Testing Prioritization



Overview of ToxRefDB & ToxCast

- ToxRefDB

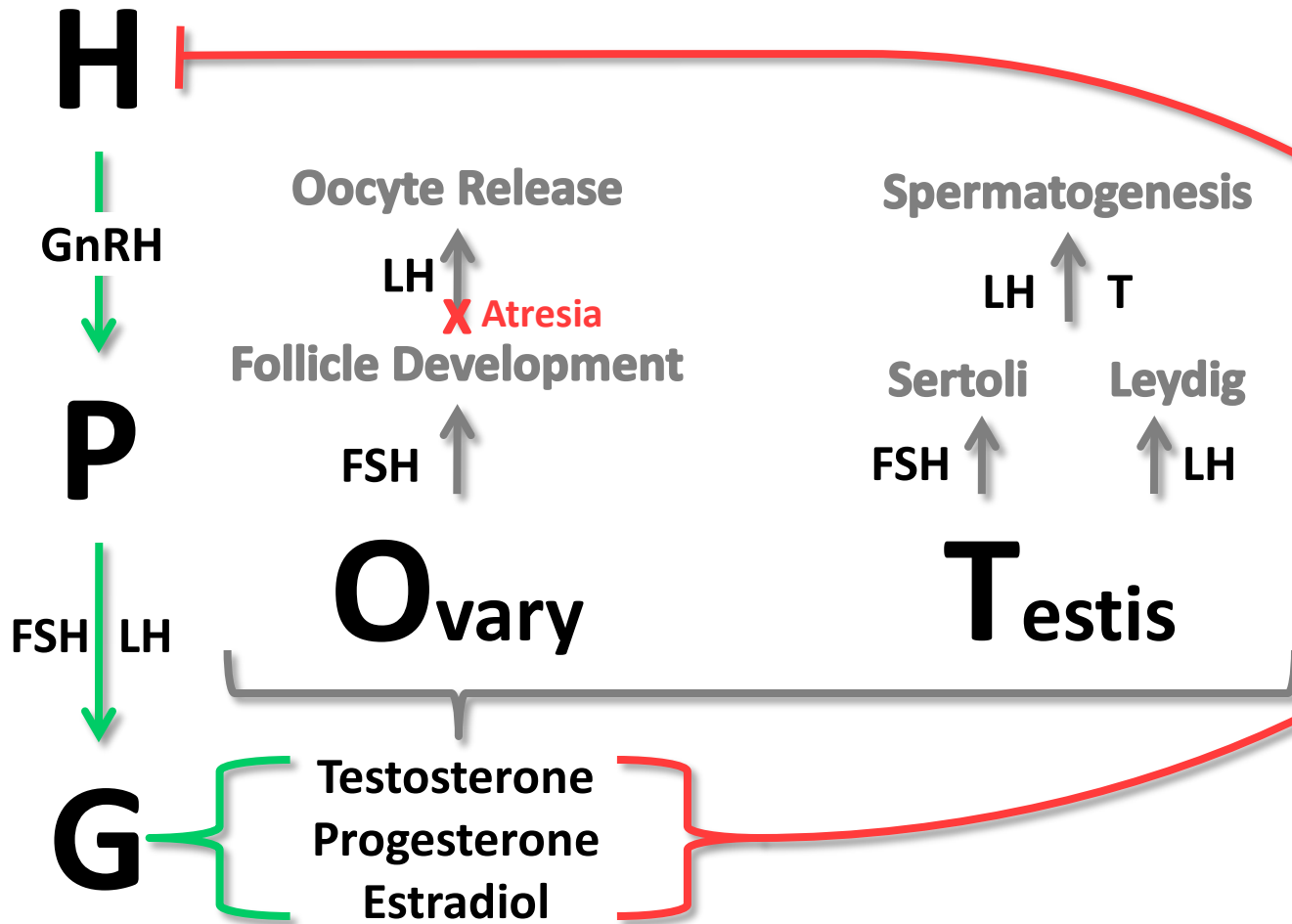
- Publically available relational database developed to capture and store traditional animal toxicity studies
- Cancer, Reproductive and Developmental studies on 820 chemicals captured to date
- Primary anchor for predictive modeling research

- ToxCast

- High throughput data generating research program aimed at using bioactivity fingerprints to predict toxicity and apply to chemical testing decision making
- Phase I: 309 data rich chemicals tested in over 600 assays and used to develop predictive models and serve as a proof-of-concept
- Phase II: 700-1000 additional chemicals tested in potentially 1000 assays and used to validate, expand, and apply predictive models of toxicity

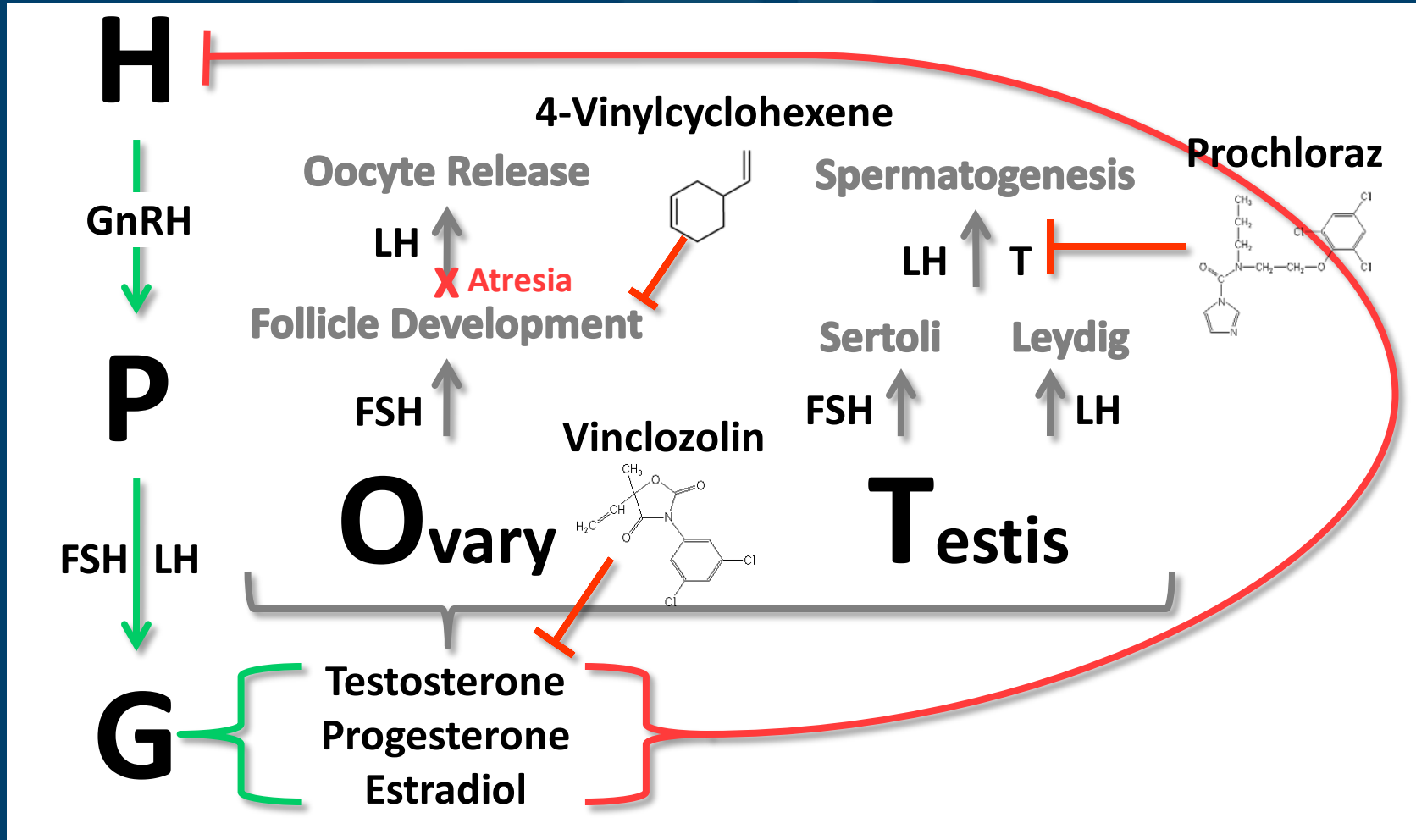
Reproductive Physiology & Impairment

Reproductive Physiology



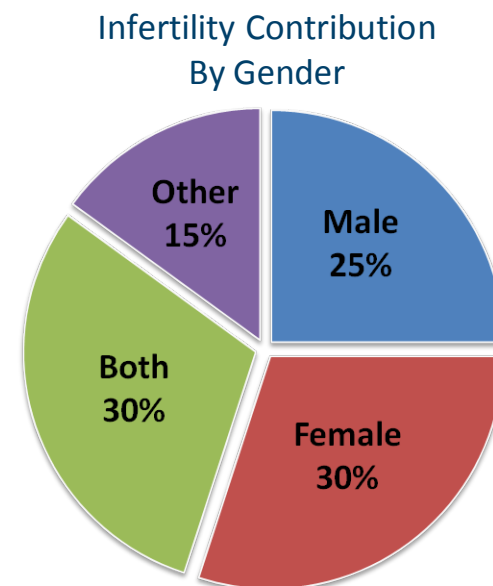
Toxicology

Reproductive Physiology



Human Reproductive Impairment & Toxicity

- 10-20% of couples sub-fertile or infertile
- Reproductive impairment
 - Male infertility
 - Generally sperm-related, but can be neuromuscular
 - Female infertility
 - Complex; many etiologies
 - Impaired fecundity (e.g., miscarriage)
 - 25% of pregnancies lost prior to clinical recognition
- Environmental contribution
 - Unknown, but links exist
 - Strong: Smoking, obesity, traffic exhaust, dioxins, combustion products
 - Suggestive: Pesticides, food additives, persistent pollutants, PCBs, PFAAs
 - Many study confounders
 - Age, lifestyle, disease background, sample size, privacy, endpoint selection



NO ROUTINE WAY TO ASSESS CHEMICALLY-INDUCED HUMAN REPRODUCTIVE TOXICITY POTENTIAL

Reproductive Toxicity Testing

Reproductive Toxicity Testing

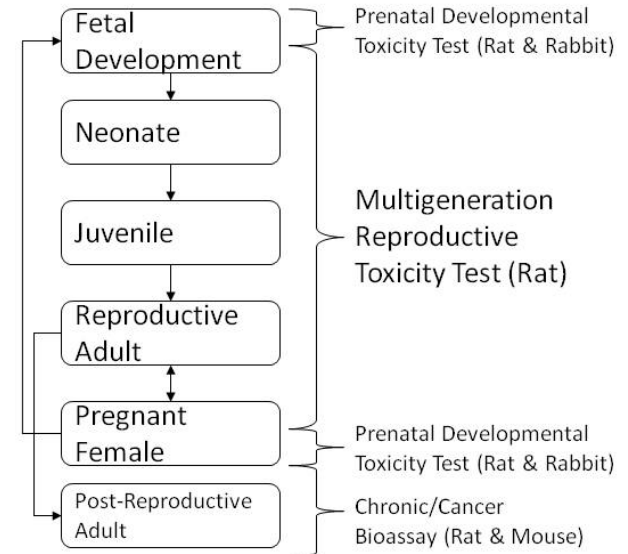
- Regulatory testing protocols

- Environmental chemicals

- **Multigeneration reproductive toxicity study (MGR)**
- Continuous-breeding protocol
- Extended one-generation reproductive toxicity study (EOGRTS)

- Pharmaceuticals

- Fertility study (Segment I)
- Peri- & Post-natal toxicity study (Segment III)

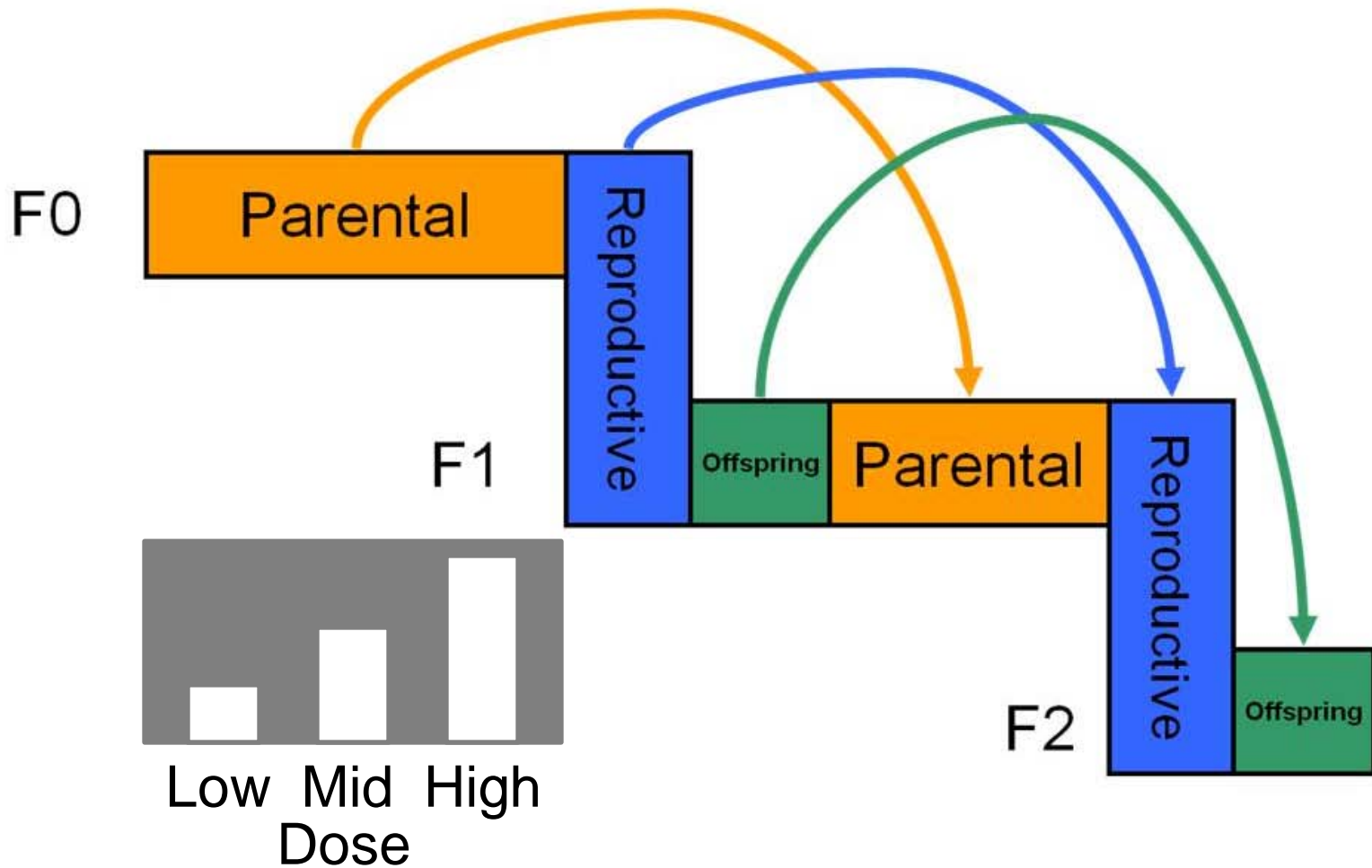


- Additional reproductive test methods

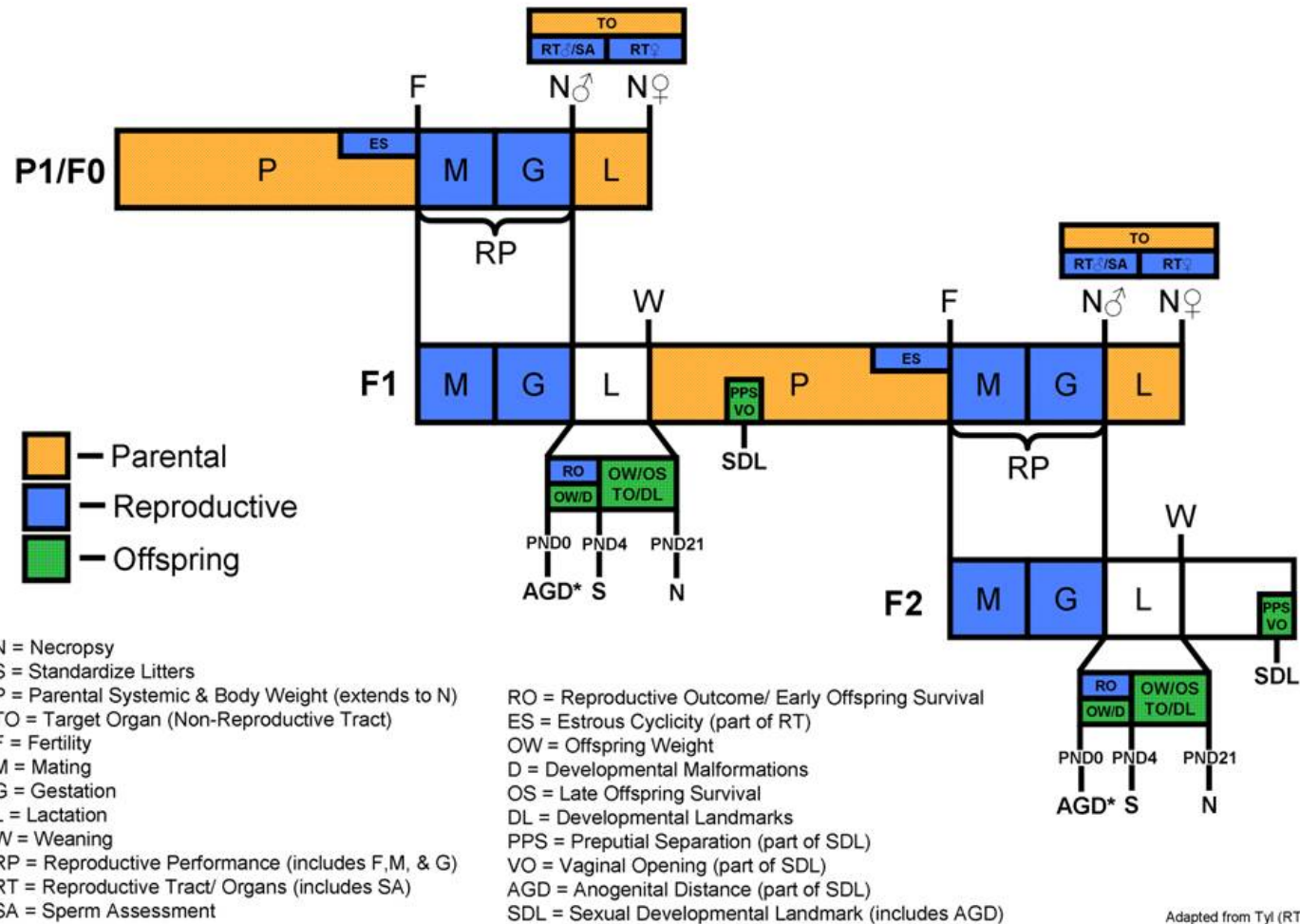
- Reproductive tissue culture systems
- In vivo endocrine assays: Hershberger, uterotrophic, pubertal
- In vitro endocrine assays: Receptor binding, transcriptional or steroidal
- Computational models: Structure-based or In Vitro batteries

– **NO CURRENT VIABLE ALTERNATIVE TO PRIORITIZE OR REPLACE MGR STUDY**

Multigeneration Reproductive Toxicity Study (MGR)



Multigeneration Reproductive Toxicity Study (MGR)



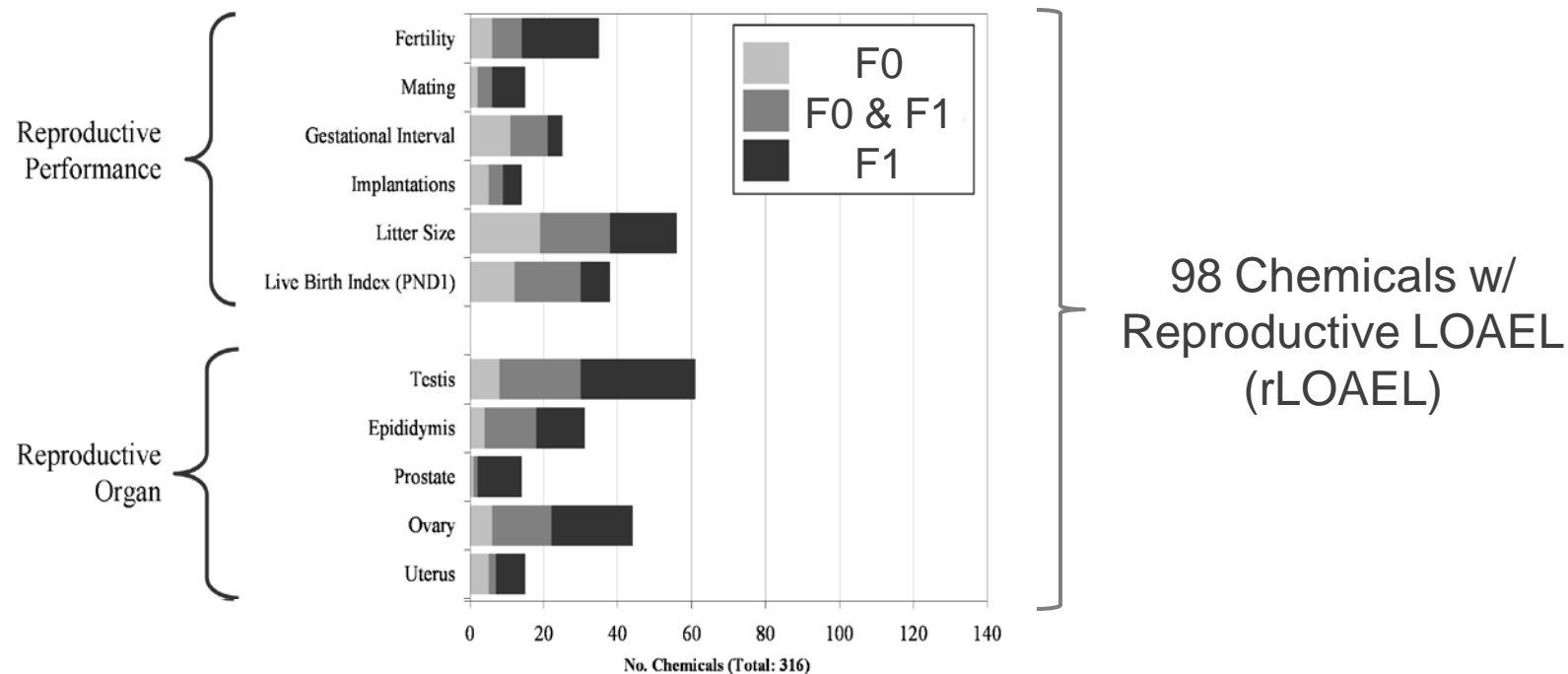
Adapted from Tyl (RTI)

ToxRefDB

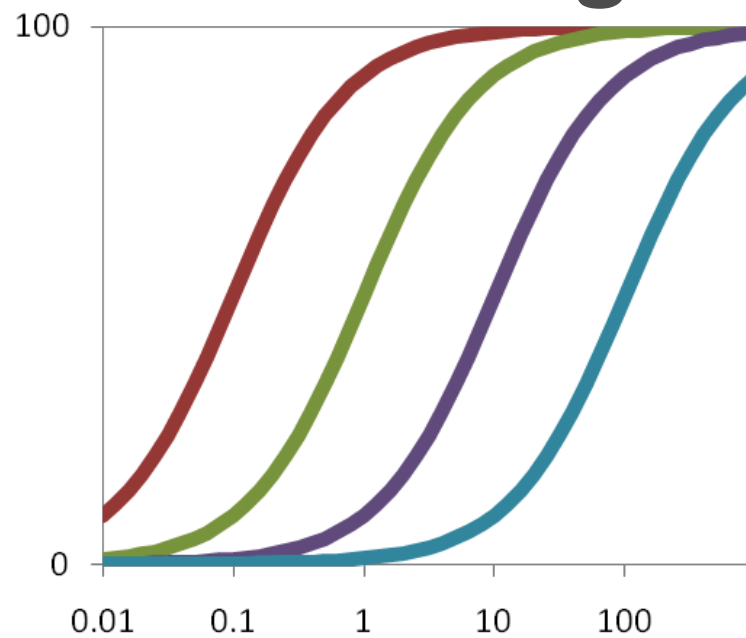
Capturing 30 years of Animal Toxicity Data

Multigeneration reproductive toxicity studies (MGR) in ToxRefDB

- As published in Martin et al. (2009): 316 chemicals entered
- As of July 2011: 416 chemicals entered
- 650 unique effects observed, 120 being unique reproductive effects
- Quantified universe of MGR studies & the inherent inefficiencies/deficiencies
- **Identified aggregated endpoint for predictive modeling**



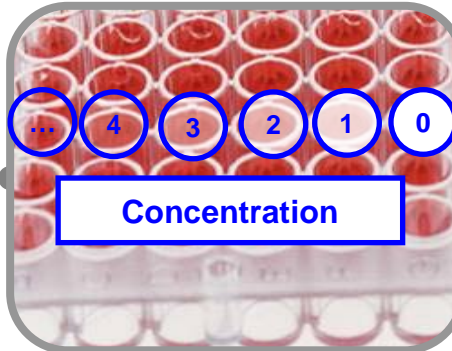
Quantitative High Throughput Screening



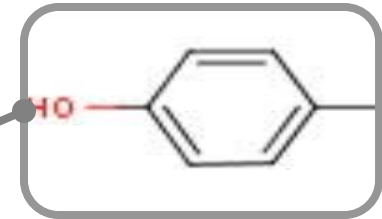
Quantitative High Throughput Screening



HTS Robotic Platform



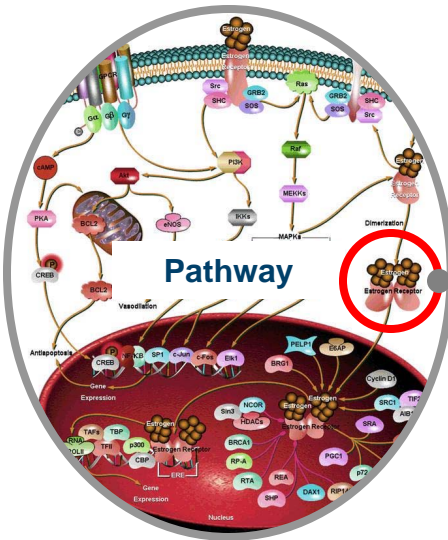
Concentration



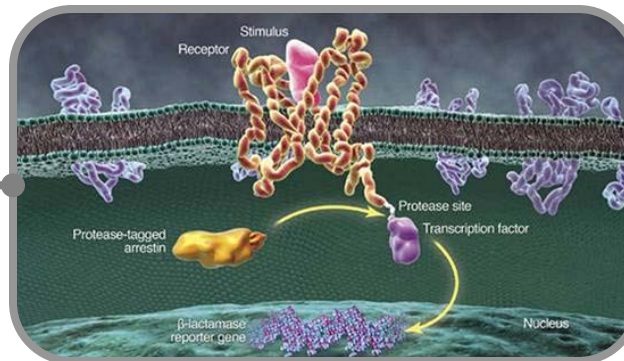
Chemical Exposure



Cell Population

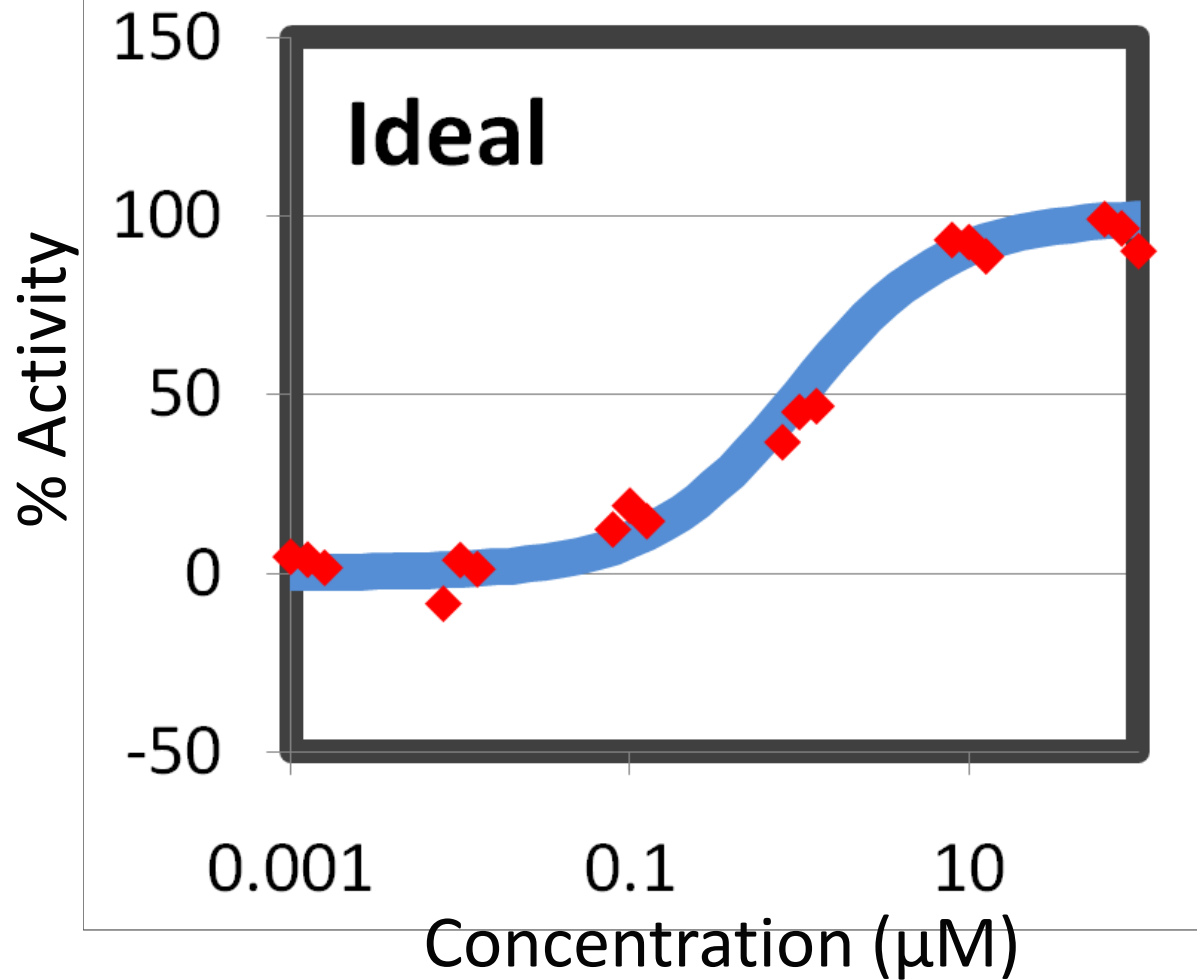


Pathway

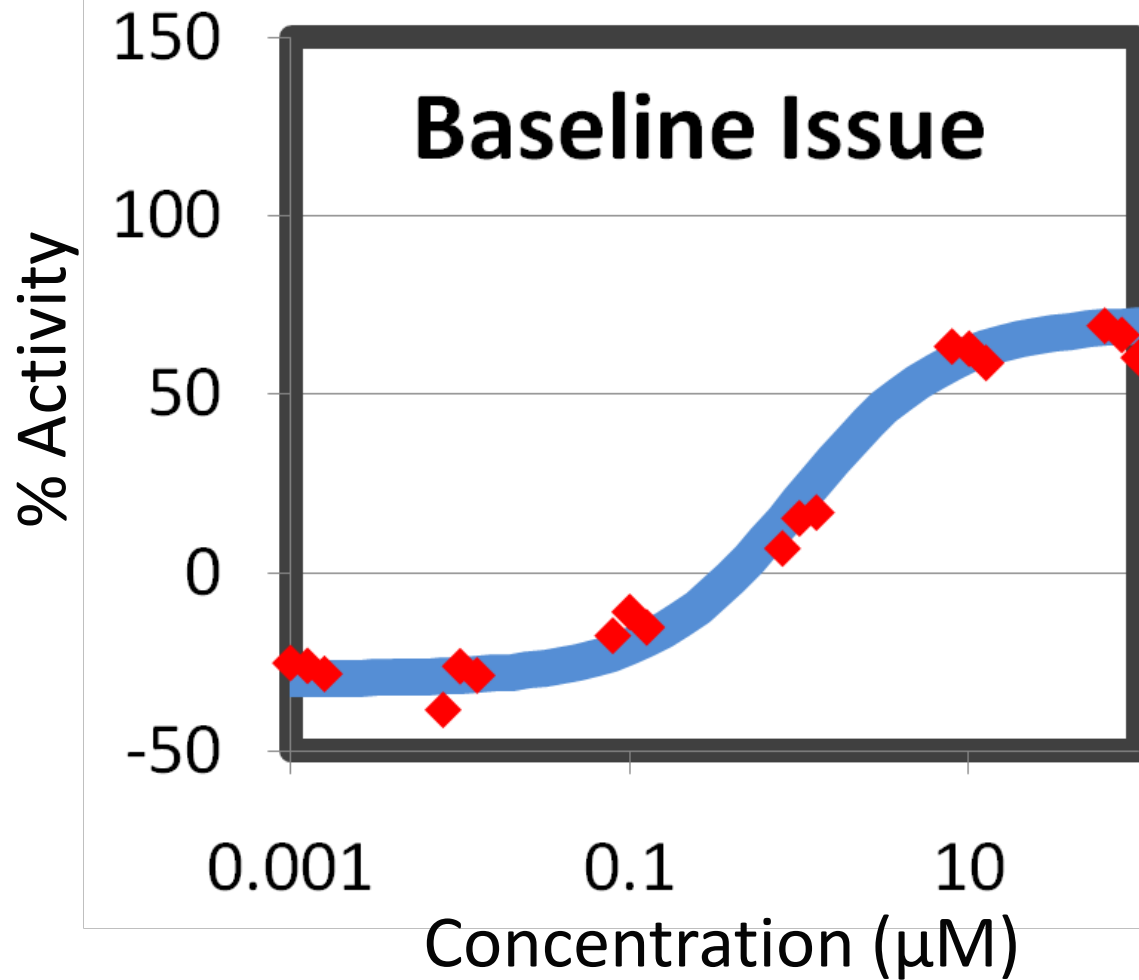


Assay Target Biology
(e.g., Estrogen Receptor)

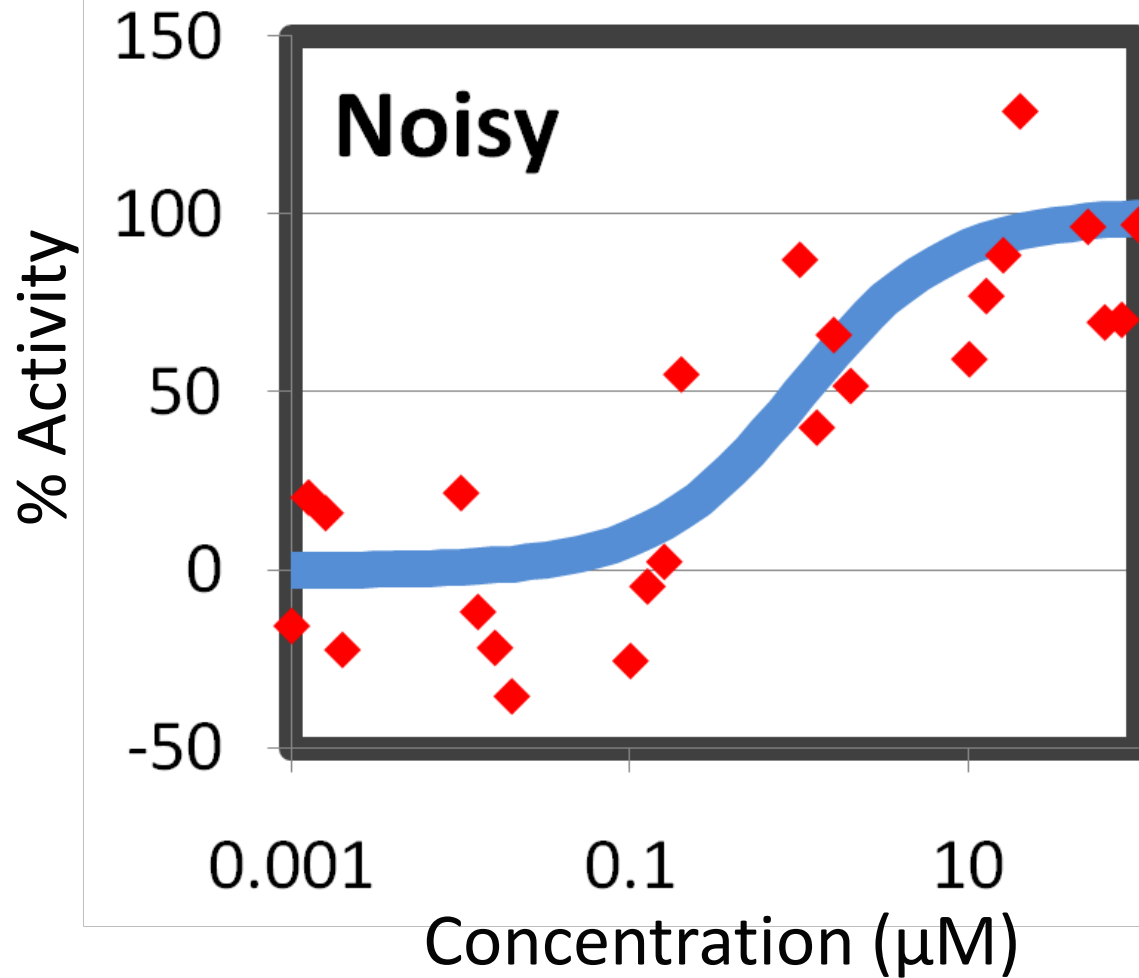
ToxCast Data Analysis



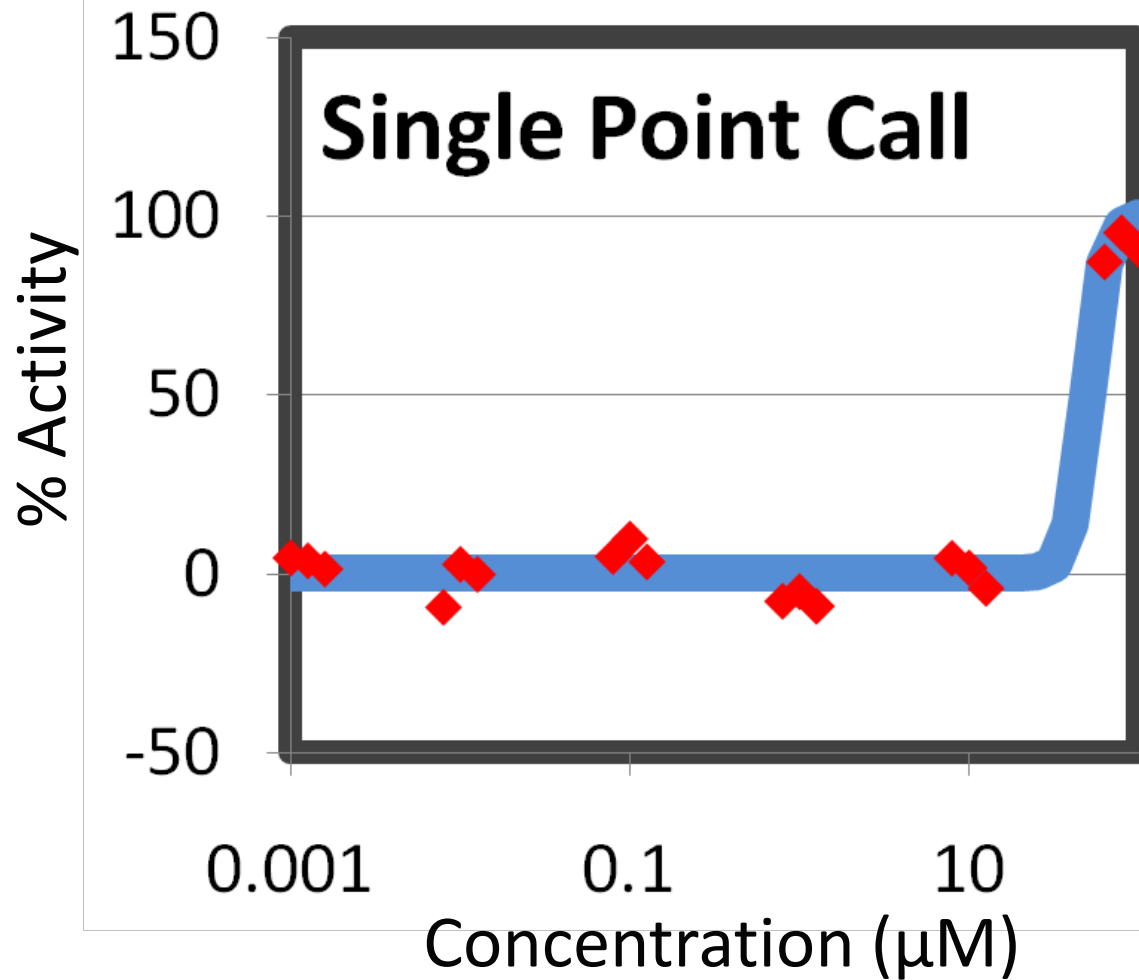
ToxCast Data Analysis



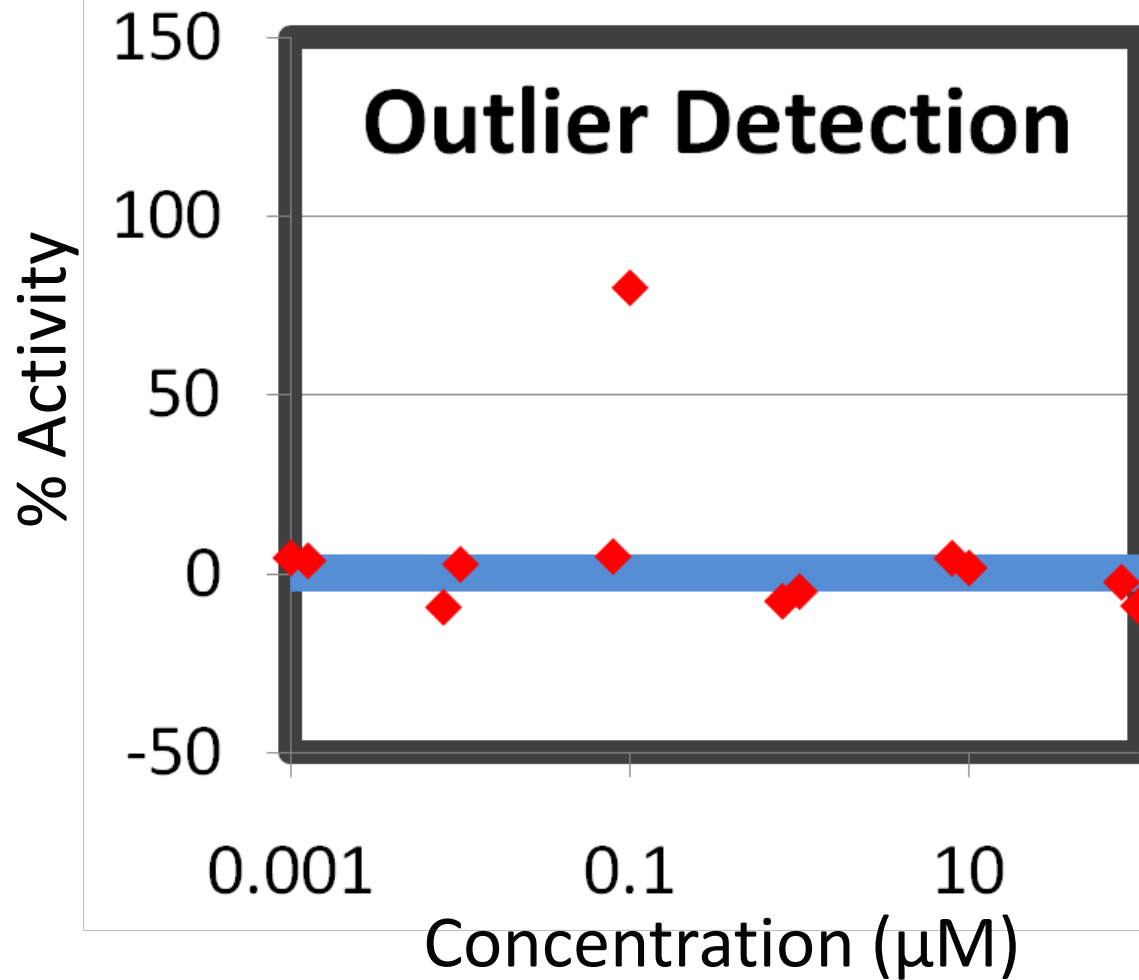
ToxCast Data Analysis



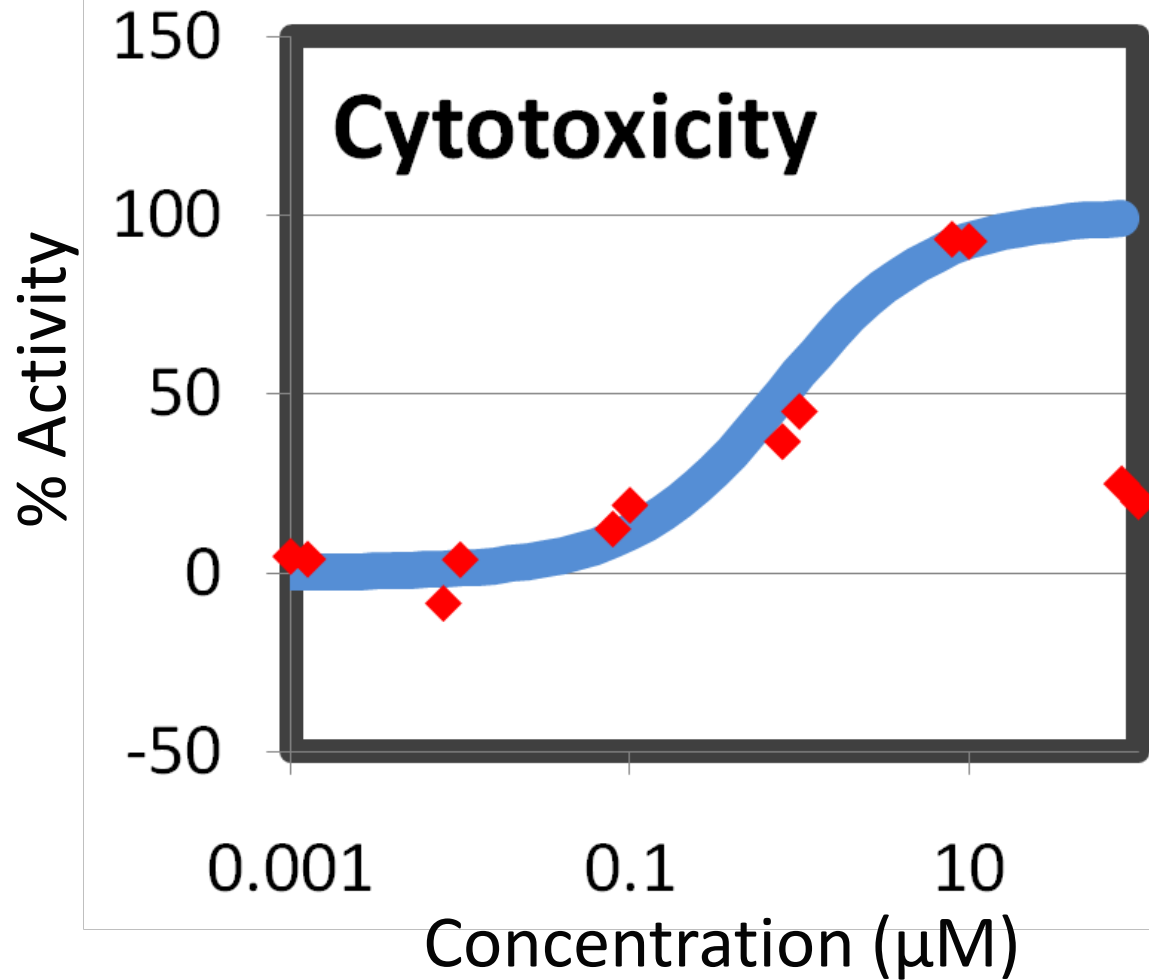
ToxCast Data Analysis



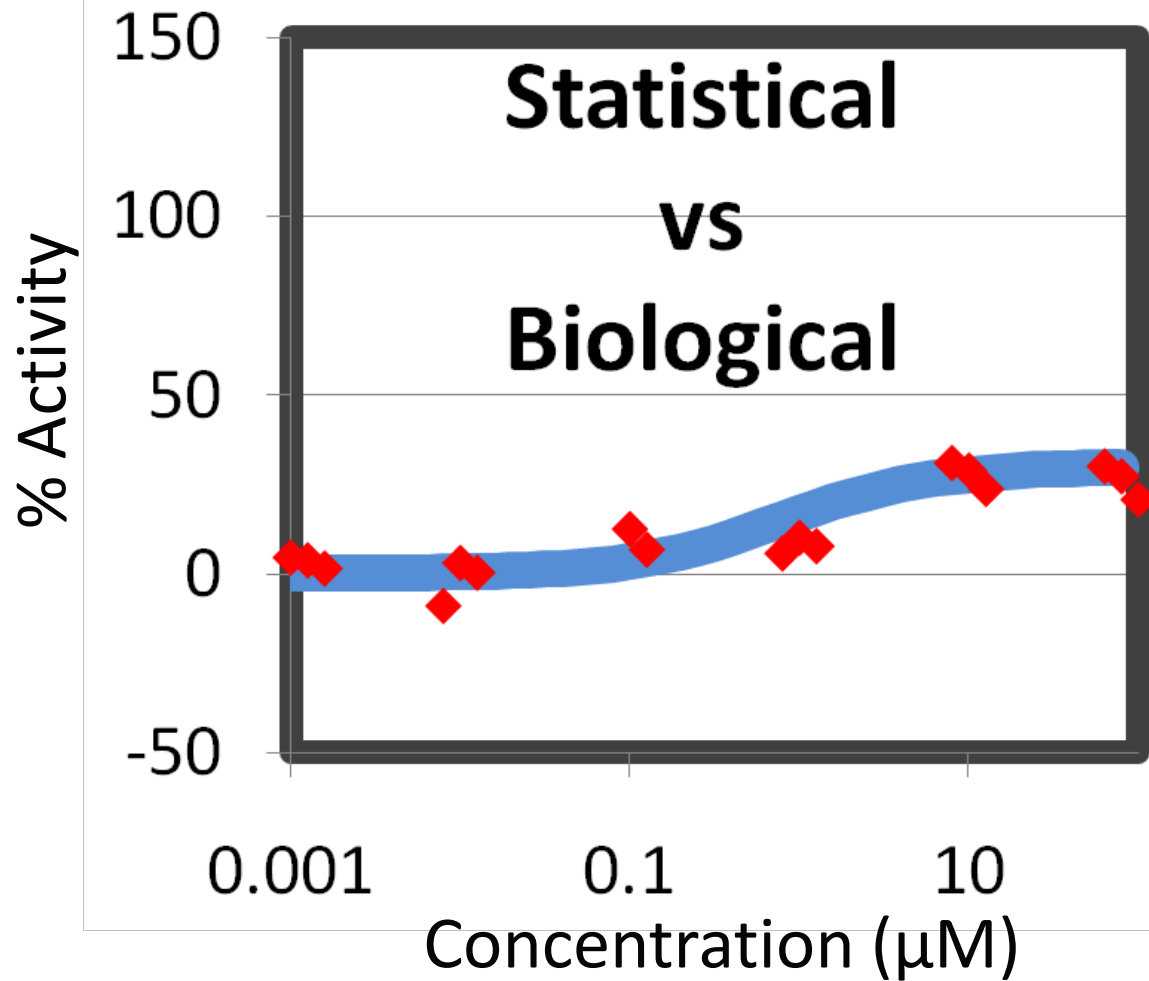
ToxCast Data Analysis



ToxCast Data Analysis

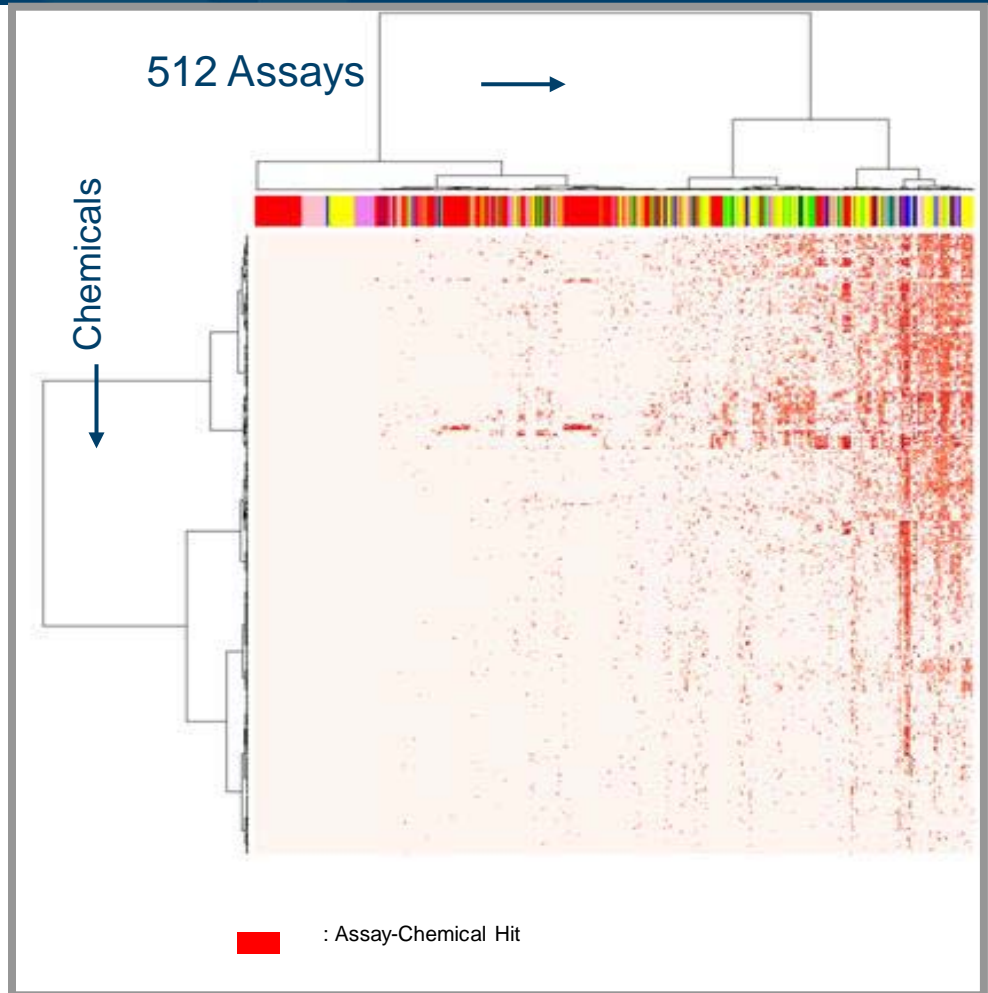


ToxCast Data Analysis



ToxCast Dataset Used In Modeling

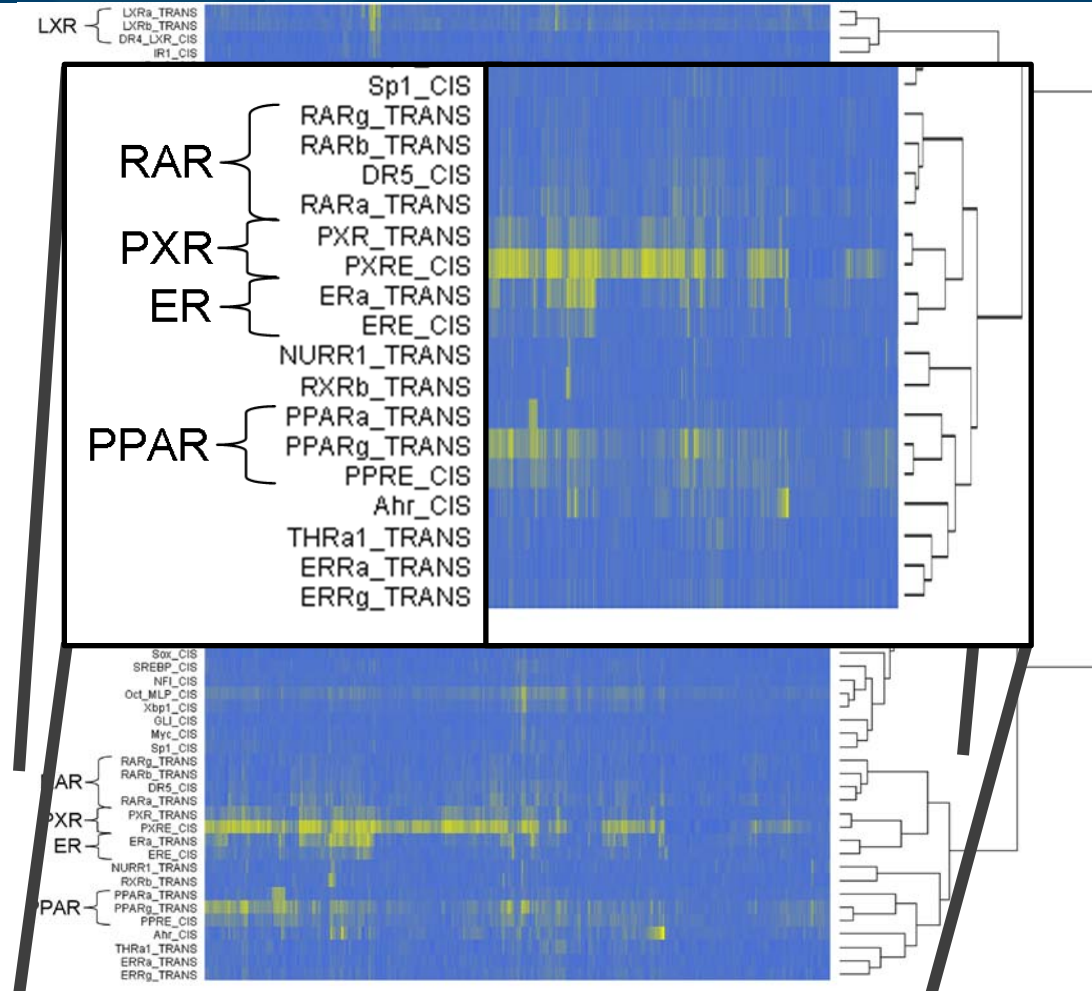
- 8 Assay Technologies
- Commercially Available Cell-free & Cell-based Assays
- 5 Technologies Used In Modeling
- 512 Total Assays Used In Modeling
- Typically Run with Negative & Positive Control
- Run in Concentration Response Format
- >3 Million Data Points



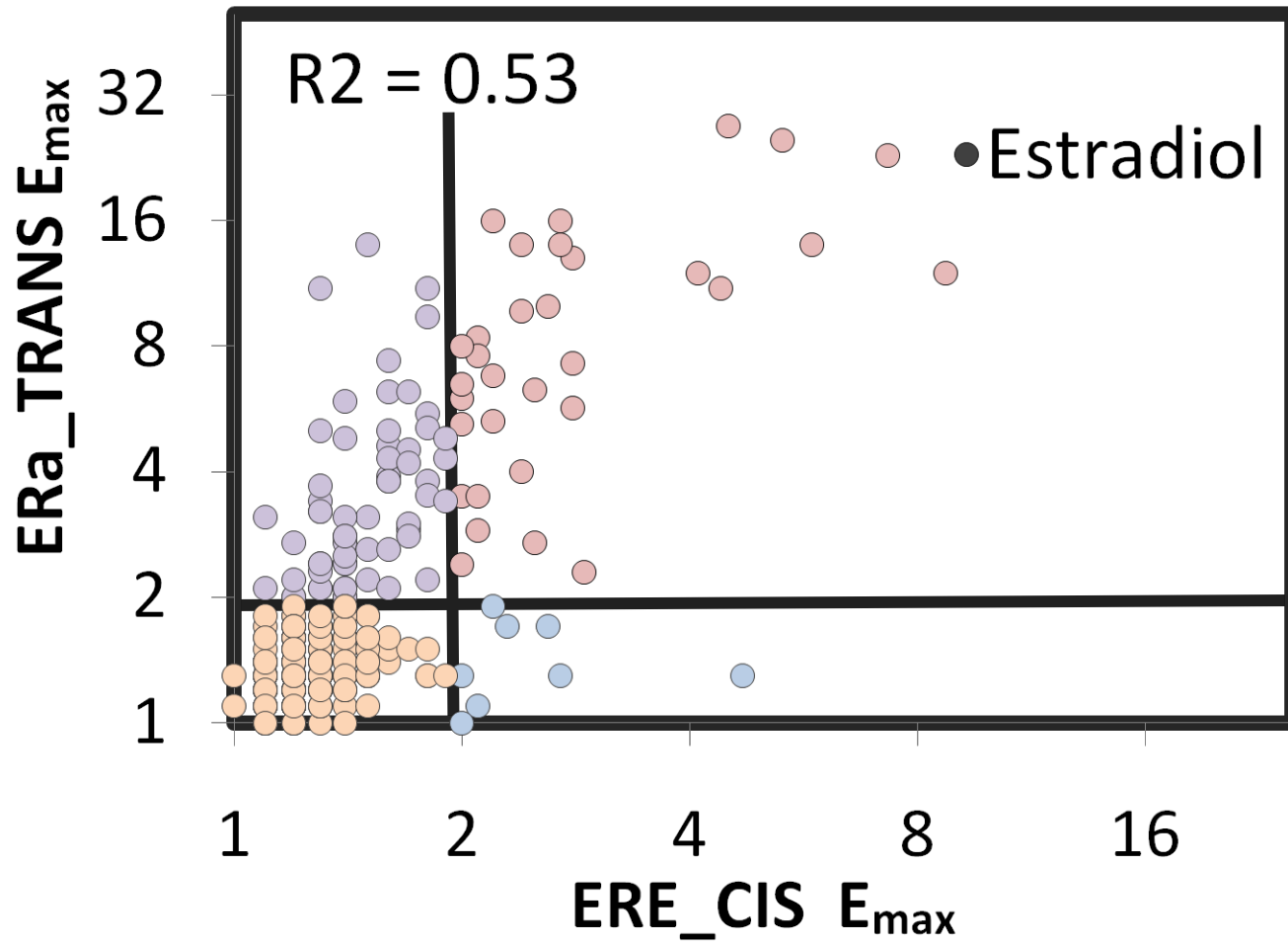
Transcription Factor Profiling

Transcription Factor (TF) Profiling

- Complimentary readouts among 73 endpoints in 2 systems
- CIS: Endogenous TF
- TRANS: Exogenous GAL4 reporter gene system
- Focus on nuclear receptors & oxidative stress pathways
- Reproducible data & highly sensitive
- Run on initial 309 chemicals and completed an additional ~700 chemicals

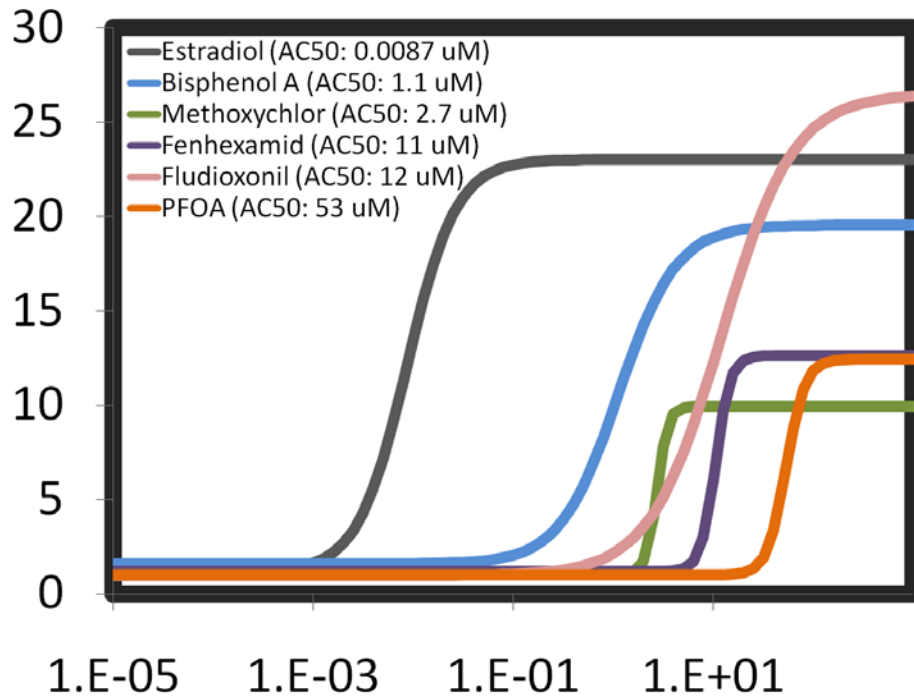


Transcription Factor Profiling

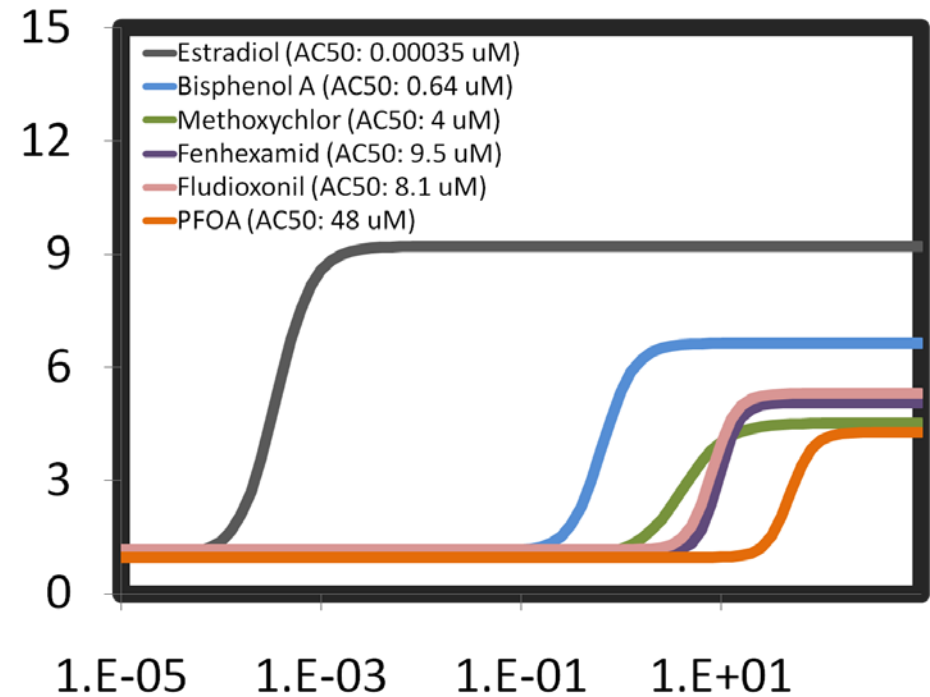


Transcription Factor Profiling

ERa_TRANS



ERE_CIS

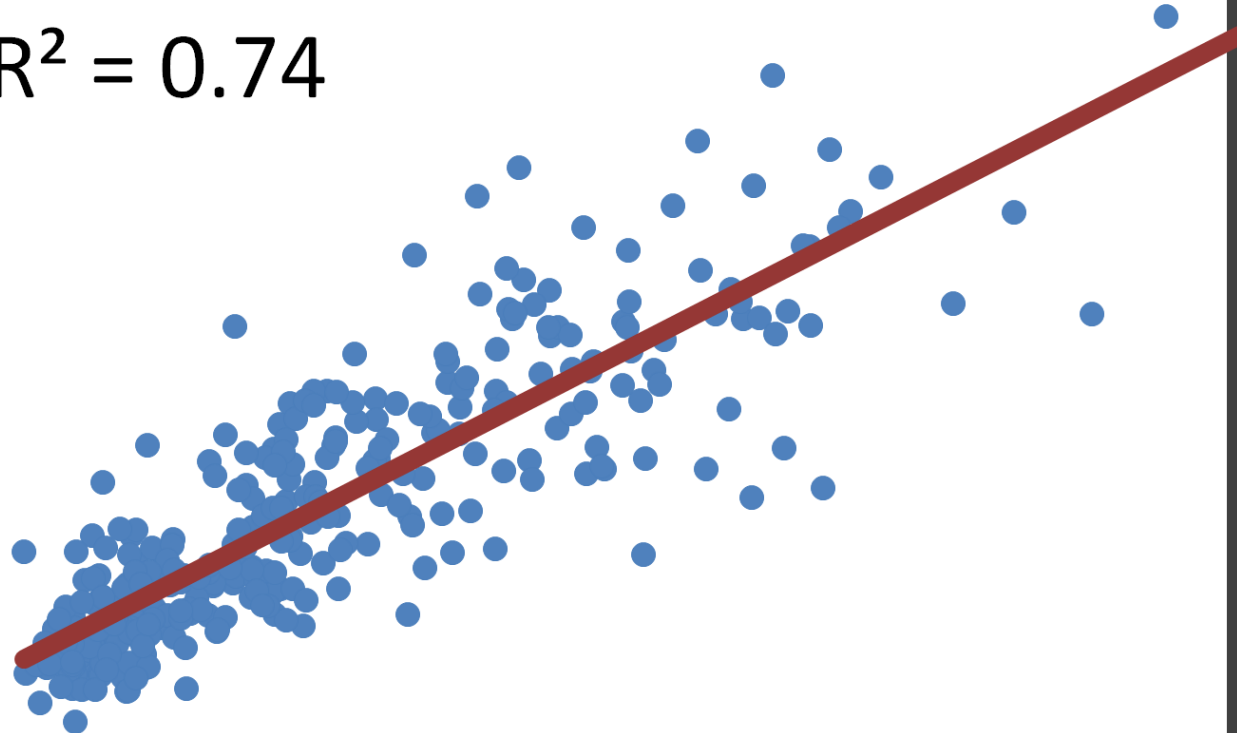


Concentration (μM)

Transcription Factor Profiling

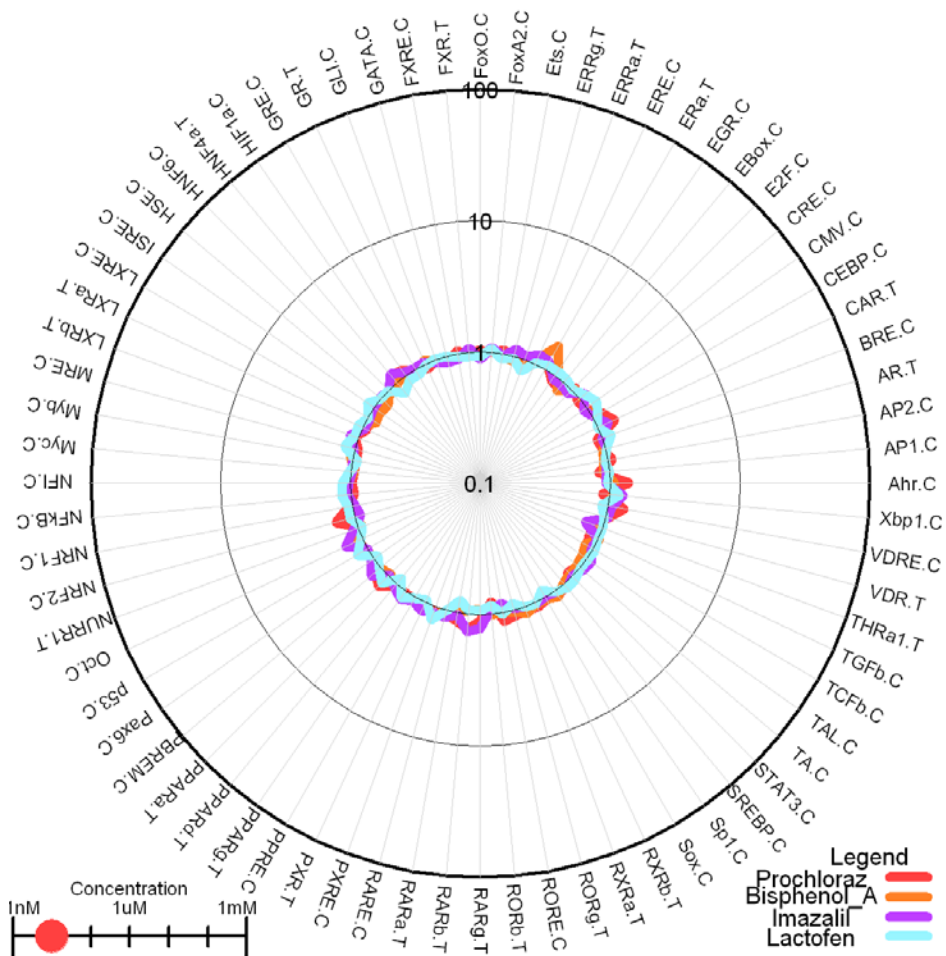
Global Activity (Avg[E_{max}])

$$R^2 = 0.74$$

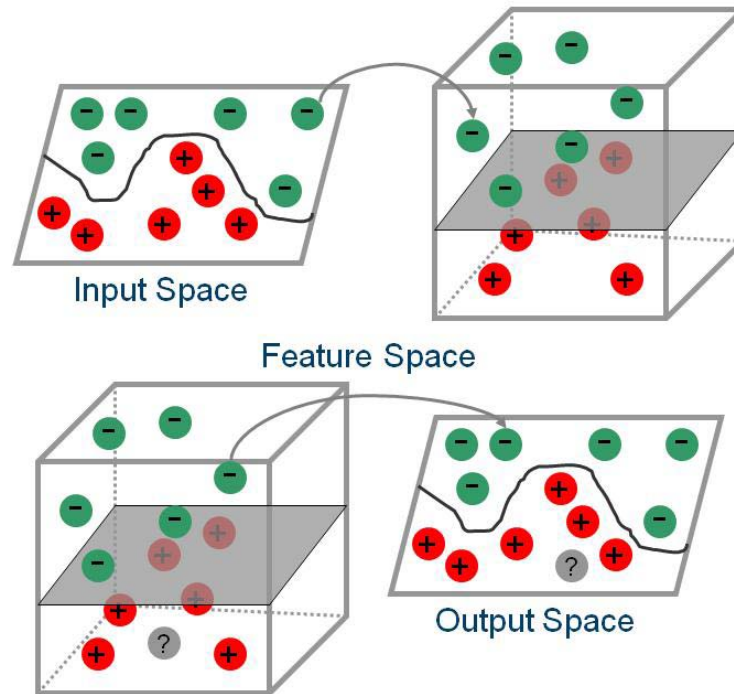


Efficacy of Markers of Oxidative Stress
Nrf2, MRE, CRE (Avg[E_{max}])

Transcription Factor Profiling

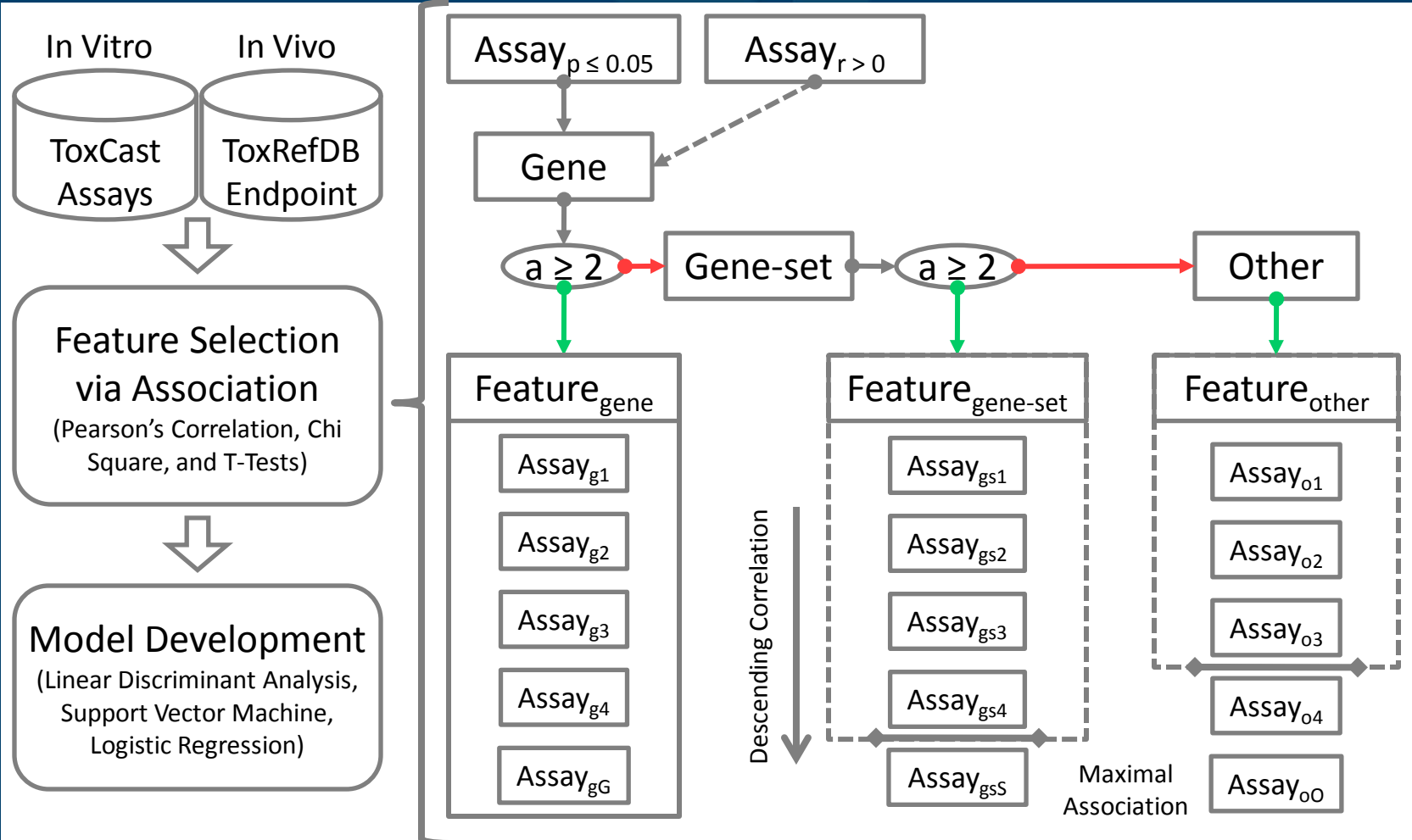


Classification Modeling Of Toxicity

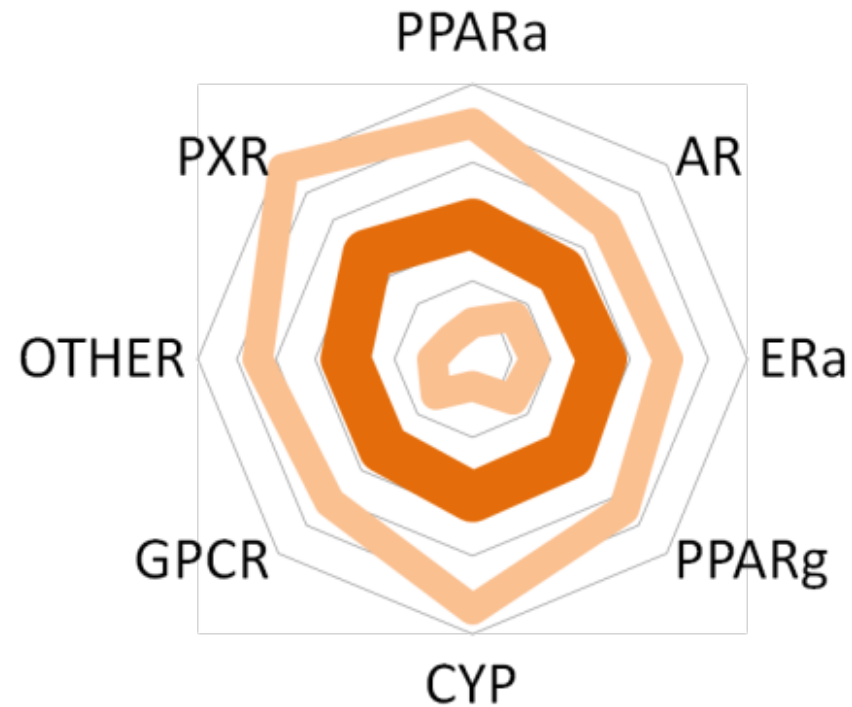


Feature Selection, Aggregation & Reduction

1 1
1 0
0 1
0 0
1 1
1 0
0 1
0 0
1 1
1 0
0 1
0 0

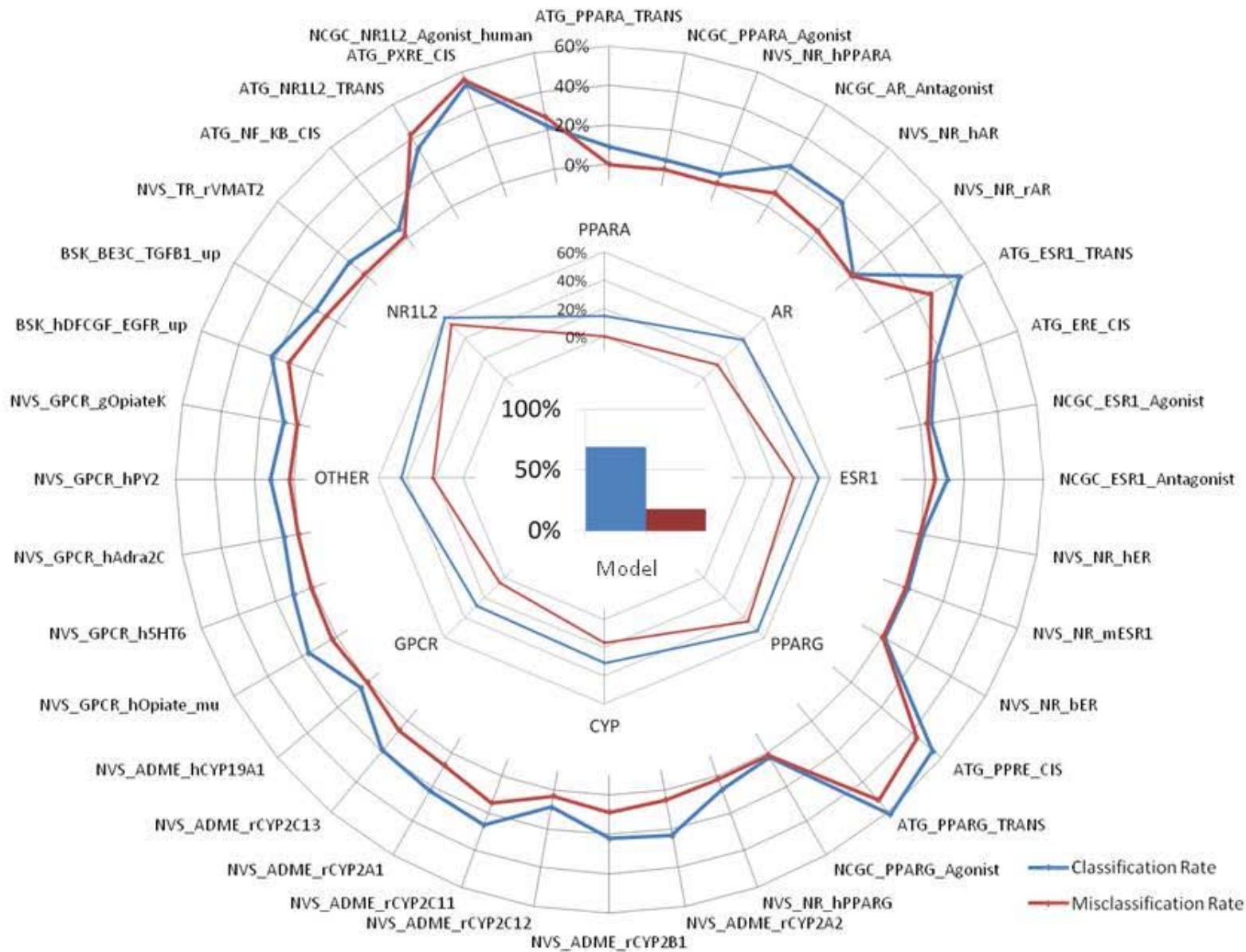


Predictive Model of Reproductive Toxicity



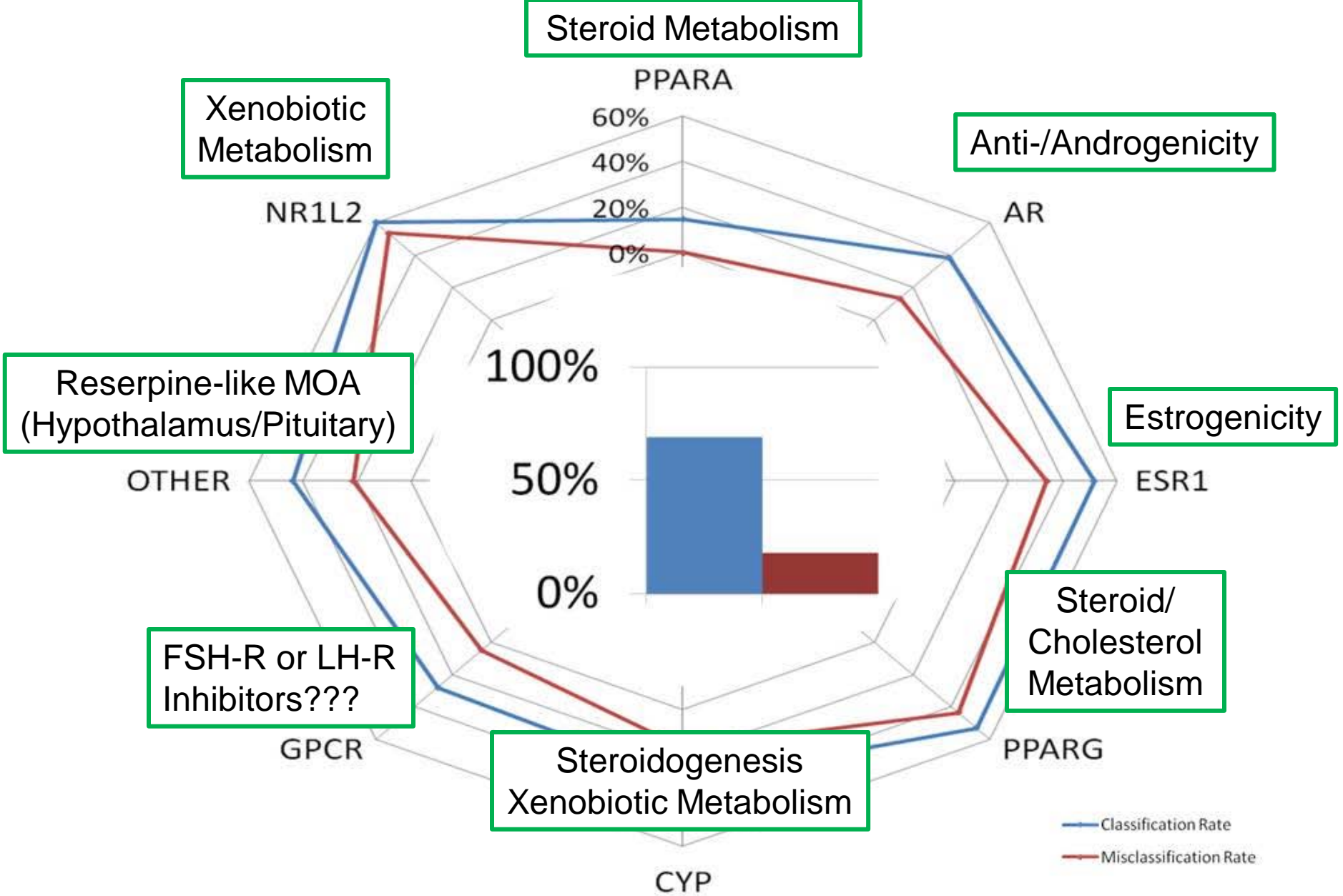
Predictive Model of Reproductive Toxicity

	<i>In vitro</i> Activity	Little to No <i>In vitro</i> Activity (<2% Active)	Total <i>In vivo</i> Chemical Counts
Acceptable Reproductive Study	206 (A)	50 (B)	256
Unacceptable Reproductive Study	31 (C)	8 (D)	39
No Reproductive Study Available	10 (E)	4 (F)	14
Total <i>In vitro</i> Chemical Counts	247	62	309



Classification Rate = TP / Total # of Positives

Misclassification Rate = FP / Total # of Negatives

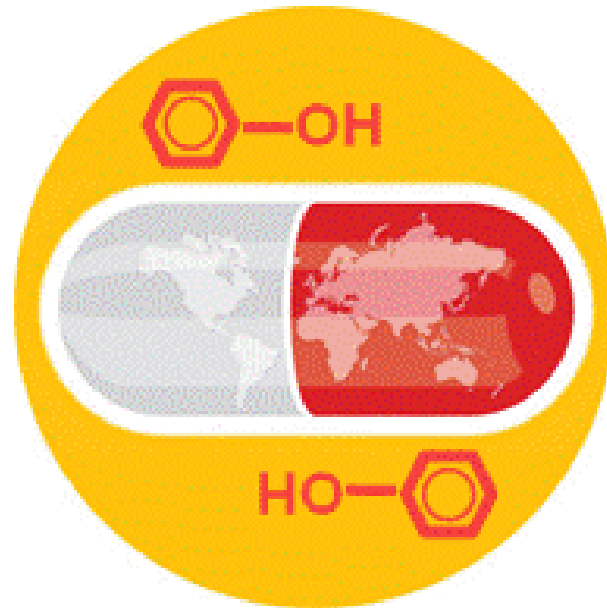


Classification Rate = TP / Total # of Positives
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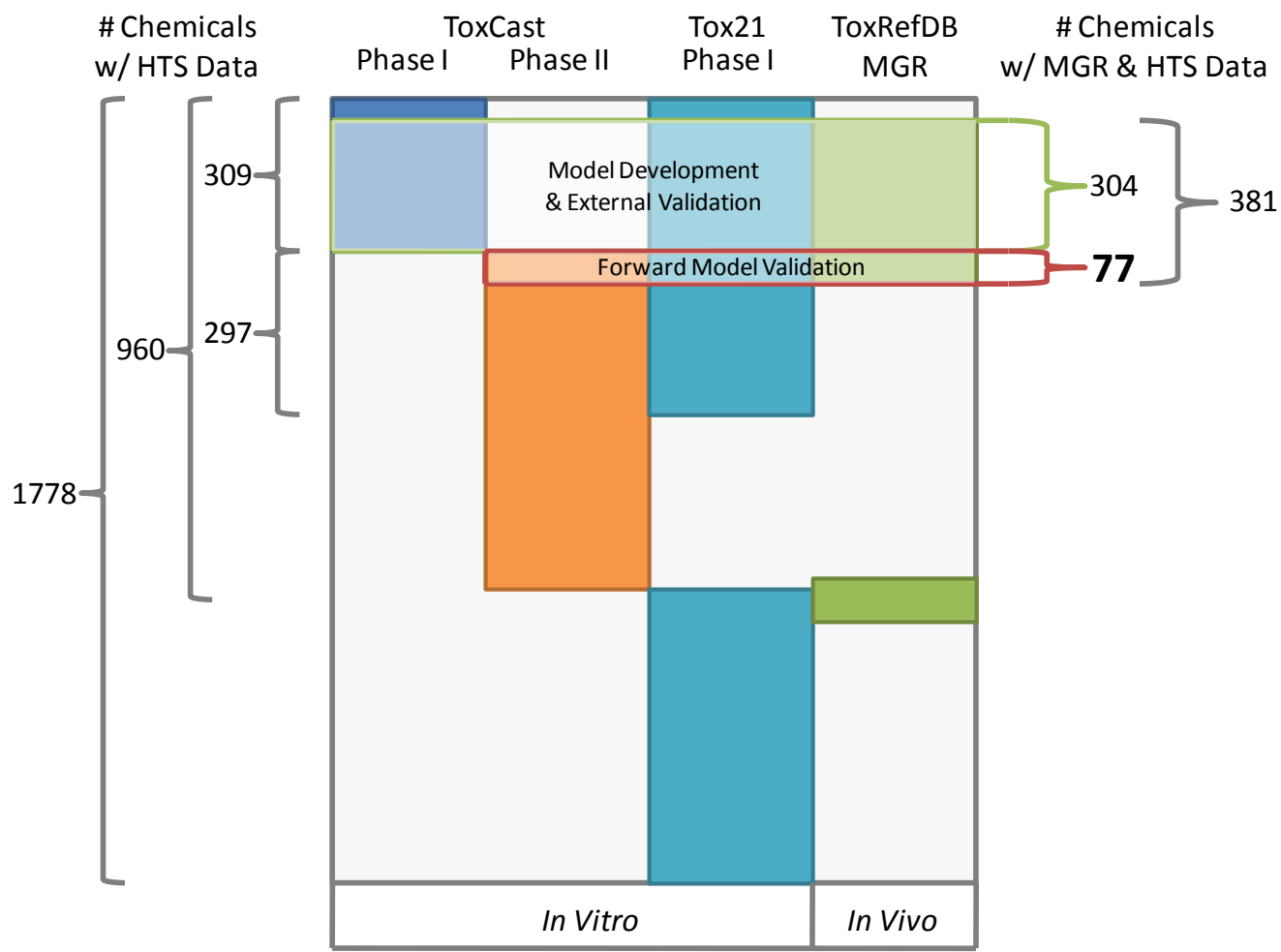
Model Summary Statistics

Cross-Validation Statistics		Full Model Statistics				Parameter Coefficients	
Learner	LDA	TP	55	F1	73%	PPARα	1.37
CV	5-fold	FP	28	RR	6.3	AR	0.98
No. F	8	FN	13	OR	17	ERα	0.45
Assays	36	TN	110	PPV	66%	PPARγ	0.23
BA Train	77%	SENS	81%	NPV	90%	CYP	0.28
SD Train	2%	SPEC	80%	Pred	78%	GPCR	0.5
BA Test	74%	BA	80%	P-Value	4.2E-17	OTHER	0.45
SD Test	5%	A	80%	Cutoff	0.6	PXR	-0.21

Forward Validation Study



Forward Validation Study



Forward Validation Study

	Model Development Chemical Set	Forward Validation Chemical Set
TP	55	37
FP	28	8
FN	13	6
TN	110	11
NA	50	15
SENSIVITY	81%	86%
SPECIFICITY	80%	58%
BALANCED ACCURACY	80%	72%
ACCURACY	80%	77%

77

Forward Validation Study

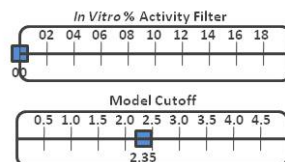
- Maintained predictivity in light of real-world confounders
 - Increased chemical diversity
 - Assay attrition & replacement
 - High positive prevalence
- High sensitivity and overall accuracy
 - **FP**: Sodium dodecyl sulfate is detergent most likely interfering with assays
 - **FN**: Acrylamide and benzene containing compounds acting through germ-cell mutagenicity mode-of-action (MOA)
 - **TP vs TN**: Structurally-related chemicals with different reproductive outcomes (Benzyl Butyl Phthalate vs Octyl Phthalate)
- Impact
 - Capable of predicting reproductive toxicity across a diverse chemical set
 - MOA : Bronopol - CYP Inhibition – Disruption of Steroidogenesis
 - Immediate impact on chemical testing decision making

Future Directions

- Assay development to fill biological gaps in model
 - Steroidogenesis
 - Germ cell mutagenicity
 - Incorporation of metabolism
- Systems modeling of the neuroendocrine system
 - Moving from a classification model to a systems model
 - Account for timing, dose/concentration, life-stage
- Developing an Integrated Testing Strategy
 - Combine model with alternative test methods
 - Incorporate systems models for mechanistic guidance
 - Combine model with other HTS-derived models
 - **Prioritize chemicals for further testing**

Case Study in Testing Prioritization

Case Study 2



Optimize for:

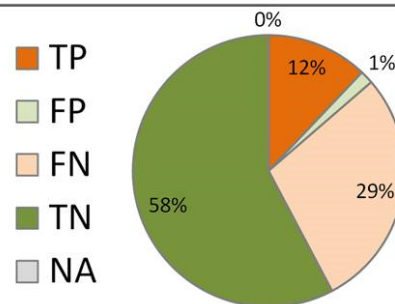
- SENS BA PPV PRED
 SPEC NPV

Application:

- Waiver Request
 -or-
 Study Request

Apply Request to:

- Predicted Positive
 -or-
 Predicted Negative



Summary Statistics

TP	46	Sens	30%	BA	64%
FP	6	Spec	97%		
FN	109	PPV	88%	Pred	78%
TN	220	NPV	67%		

Model Applied: 381 Chemicals

Model Not Applied (NA): 0 Chemicals

Studies Requested: 52

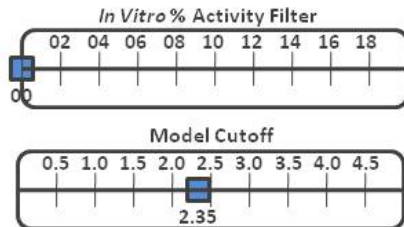
Total Cost Savings: \$34 Million

Cost Savings/Chemical: \$90,000

Case Study in Testing Prioritization

- Applied to 381 chemicals (155 chemicals positive)
 - 78% overall model balanced accuracy using default parameters
 - Adjustable parameters
- Case Study Context
 - Authority to request MGR study
 - Must prioritize a portion of chemical set
- Performed with no additional information
- Cost savings estimates = Increased efficiency of using the model to select which chemicals to test vs. selecting chemicals randomly

Case Study in Testing Prioritization



Optimize for:

- SENS BA PPV PRED
 SPEC NPV

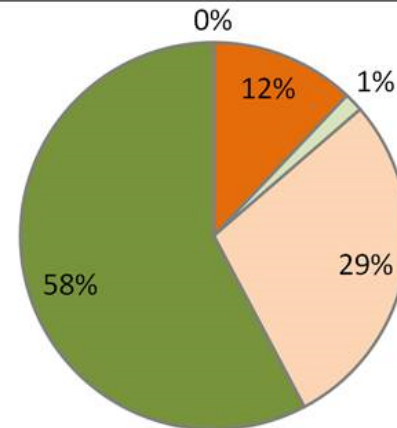
Application:

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Apply Request to:

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 -or-
 Predicted Negative

- TP
- FP
- FN
- TN
- NA



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Conclusions

- Captured 30 years of traditional reproductive toxicity data using a standardized vocabulary enabling consistent endpoint definitions
- Analyzing and interpreting HTS data requires bioinformatic workflows and understanding of assay confounders and other considerations
- Used HTS to develop forward validated predictive model of reproductive toxicity with biological plausability
- Demonstrated ability of model to impact chemical testing decision making
- Dashboard concept enables user interaction with model and direct input into decision points

