

# ToxCast™

## One Step in the NRC Vision of 21<sup>st</sup> Century Toxicology

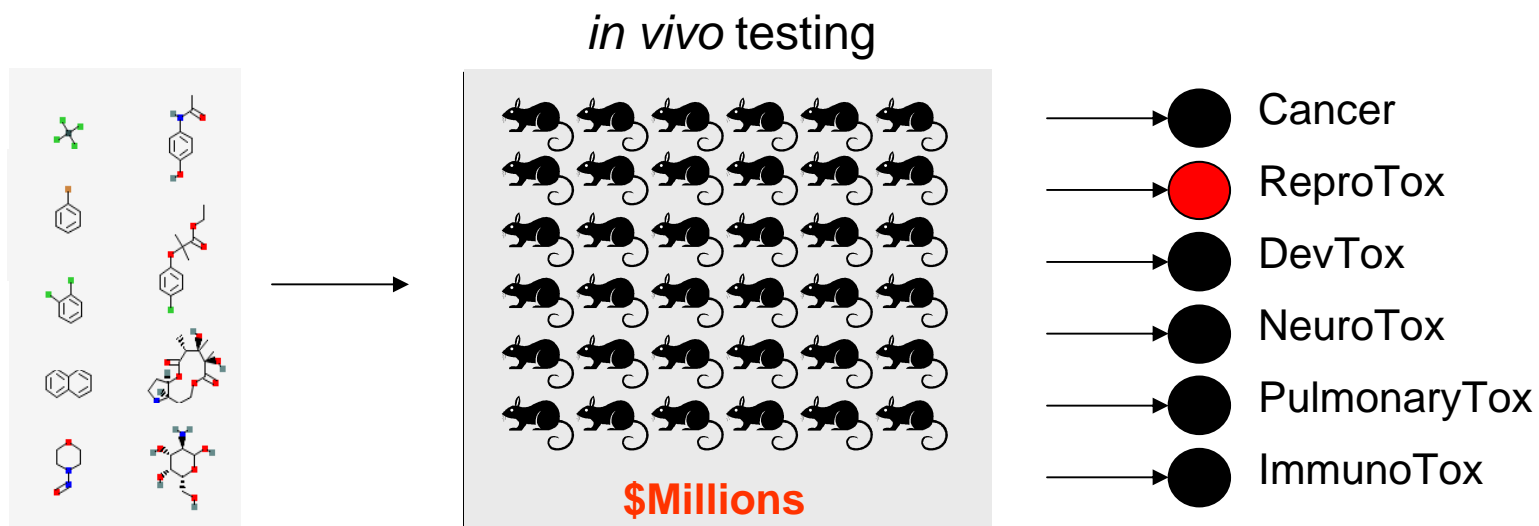
*Michigan Regional SOT  
Midland, MI*

UNITED STATES ENVIRONMENTAL PROTECTION AGENCY



**COMPUTATIONAL  
TOXICOLOGY**

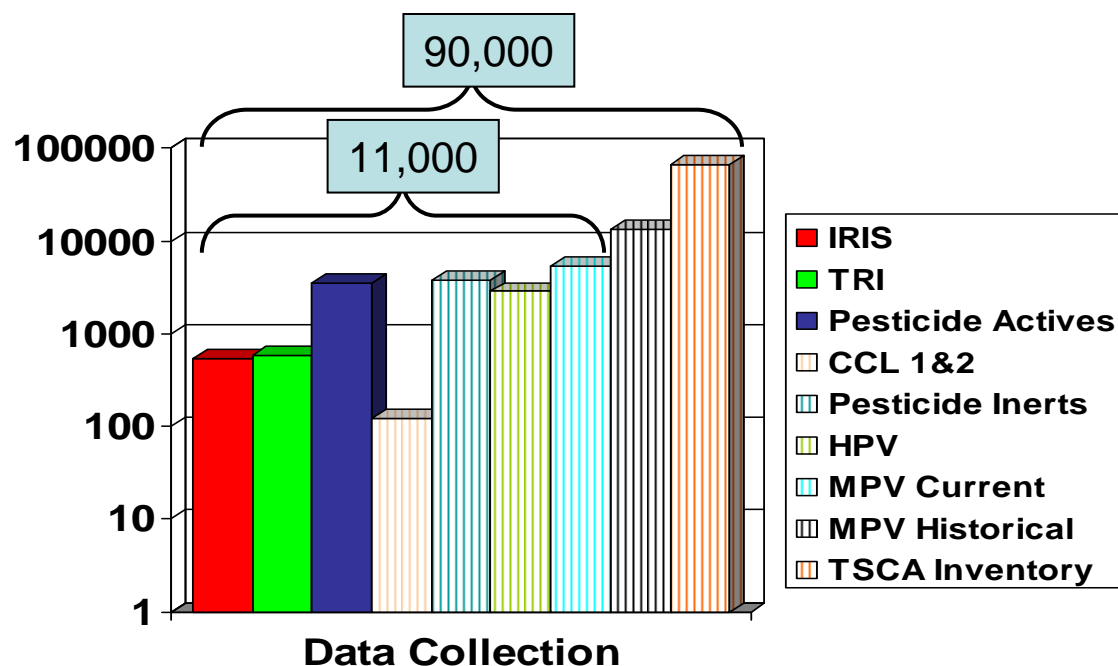
# Current Approach for Toxicity Testing



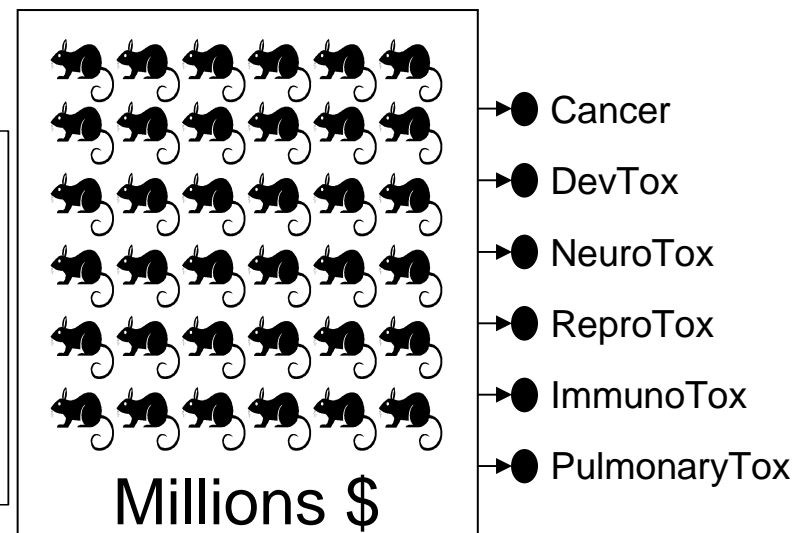


# Putting Numbers on the Problem

## Too Many Chemicals

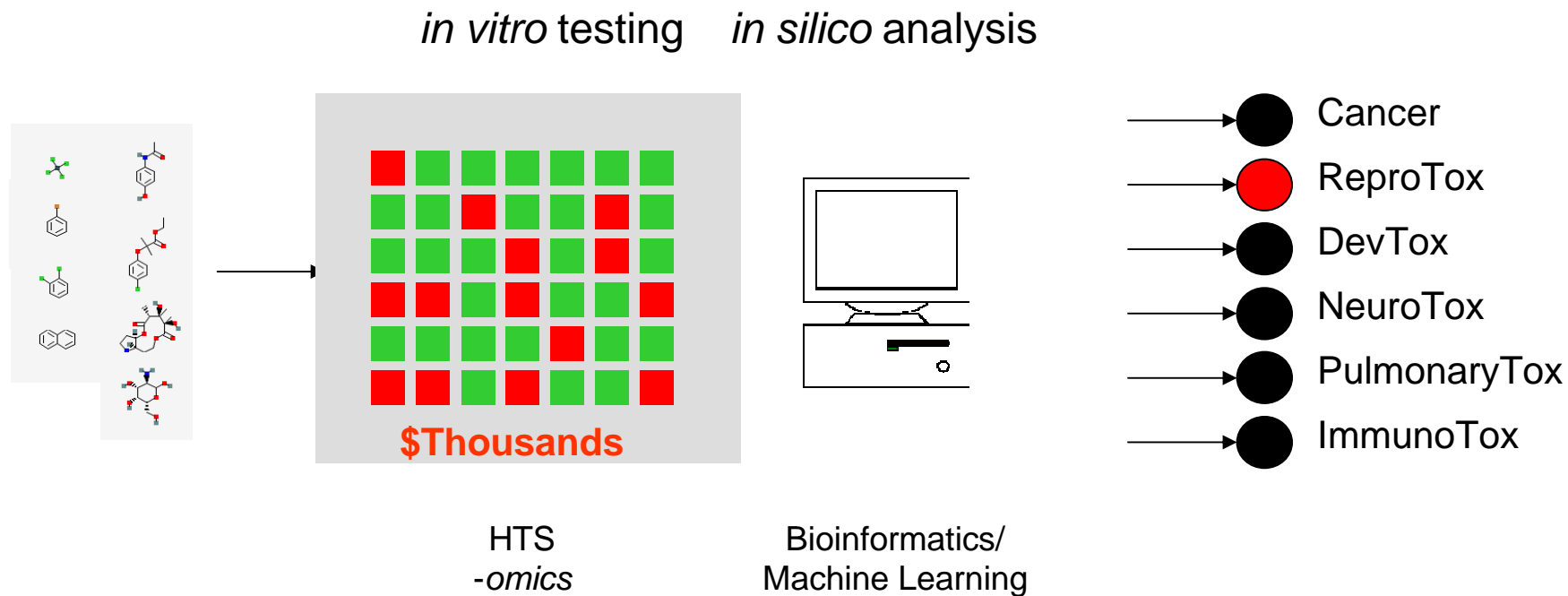


## Too High a Cost



...and not enough data.

# Future of Toxicity Testing



EPAs Approach: The ToxCast Research Program



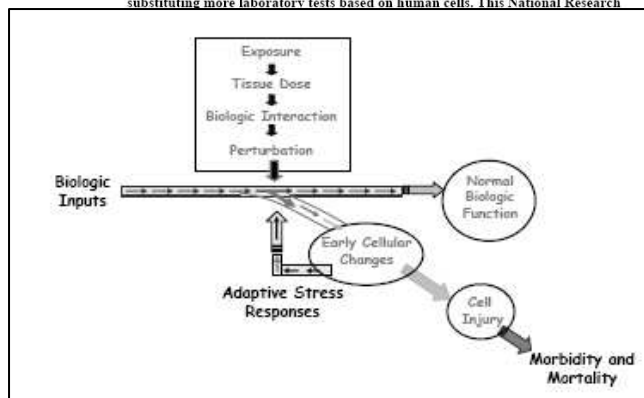
# Transforming Toxicology

July 2007

## Toxicity Testing in the 21st Century: A Vision and a Strategy

Advances in molecular biology, biotechnology, and other fields are paving the way for major improvements in how scientists evaluate the health risks posed by potentially toxic chemicals found at low levels in the environment. These advances would make toxicity testing quicker, less expensive, and more directly relevant to human exposures. They could also reduce the need for animal testing by substituting more laboratory tests based on human cells. This National Research

REPORT  
IN BRIEF  
THE NATIONAL  
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and extrapolation that remain controversial. Test animals are often exposed to higher doses than would be expected for typical human exposures, requiring assumptions about



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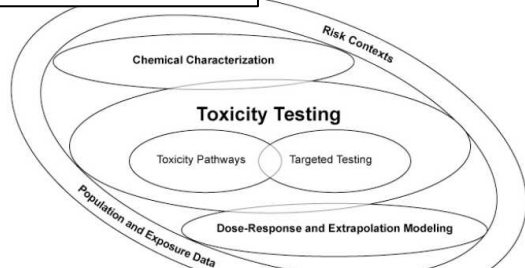


Figure 1. The committee's vision for toxicity testing is a process that can include chemical characterization, toxicity testing, and dose-response and extrapolation modeling as part of broader agency decision-making.

Office of Research  
National Center for Computational Toxicology

## POLICY FORUM

TOXICOLOGY

## Transforming Environmental Health Protection

Francis S. Collins,<sup>1\*</sup> George M. Gray,<sup>2\*</sup> John R. Bucher<sup>3\*</sup>

In 2005, the U.S. Environmental Protection Agency (EPA), with support from the U.S. National Toxicology Program (NTP), funded a project at the National Research Council (NRC) to develop a long-range vision for toxicity testing and a strategic plan for implementing that vision. Both agencies wanted future toxicity testing and assessment paradigms to meet evolving regulatory needs. Challenges include the large numbers of substances that need to be tested and how to incorporate recent advances in molecular toxicology, computational sciences, and information technology, to rely increasingly on human as opposed to animal data; and to offer increased efficiency in design and costs (1–5). In response, the NRC Committee on Toxicity Testing and Assessment of Environmental Agents produced two reports that reviewed current toxicity testing, identified key issues, and developed a vision and implementation strategy to create a major shift in the assessment of chemical hazard and risk (6, 7). Although the NRC reports have laid out a solid theoretical rationale, comprehensive and rigorously gathered data (and comparisons with historical animal data) will determine whether the hypothesized improvements will be realized in practice. For this purpose, NTP, EPA, and the National Institutes of Health Chemical Genomics Center (NCGC) (organizations with expertise in experimental toxicology, computational toxicology, and high-throughput technologies, respectively) have established a collaborative research program.

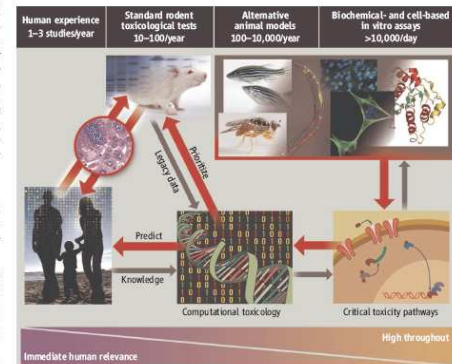
**EPA, NCGC, and NTP Joint Activities**  
In 2004, the NTP released its vision and roadmap for the 21st century (1), which established initiatives to integrate high-

throughput screening (HTS) and other automated screening assays into its testing program. In 2005, the EPA established the National Center for Computational Toxicology (NCCT). Through these initiatives, NTP and EPA, with the NCGC, are promoting the evolution of toxicology from a predominantly observational science at the level of disease-specific models in vivo to a predominantly predictive science focused on broad inclusion of target-specific, mechanism-based, biological observations in vitro (1, 4) (see figure, below).

**Toxicity pathways.** In vitro and in vivo tools are being used to identify cellular responses after chemical exposure expected to result in adverse health effects (7). HTS methods are a primary means of discovery for drug development, and screening of >100,000 compounds per day is routine (8). However, drug-discovery HTS methods traditionally test compounds at one concentra-

We propose a shift from primarily in vivo animal studies to in vitro assays, in vivo assays with lower organisms, and computational modeling for toxicity assessments.

tion, usually between 2 and 10  $\mu$ M, and tolerate high false-negative rates. In contrast, in the EPA, NCGC, and NTP combined effort, all compounds are tested at as many as 15 concentrations, generally ranging from ~5 nM to ~100  $\mu$ M, to generate a concentration-response curve (9). This approach is highly reproducible, produces significantly lower false-positive and false-negative rates than the traditional HTS methods (9), and facilitates multiassay comparisons. Finally, an informatics platform has been built to compare results among HTS screens; this is being expanded to allow comparisons with historical toxicologic NTP and EPA data (<http://ncgc.nih.gov/pub/openhts>). HTS data collected by EPA and NTP, as well as by the NCGC and other Molecular Libraries Initiative centers (<http://mln.nih.gov/>), are being made publicly available through Web-based databases [e.g., PubChem (<http://pubchem.ncbi.nlm.nih.gov/>)]. In addition,



**Transforming toxicology.** The studies we propose will test whether high-throughput and computational toxicology approaches can yield data predictive of results from animal toxicity studies, will allow prioritization of chemicals for further testing, and can assist in prediction of risk to humans.

906

15 FEBRUARY 2008 VOL 319 SCIENCE www.sciencemag.org

Science, Feb 15, 2008



### Robots could reduce animal tests

U.S. scientists are taking the first step towards testing potentially hazardous chemicals on cells grown in a laboratory, without using live animals.

Two government agencies are looking into the merits of using high-speed automated robots to carry out tests.

The long-term goal is to reduce the cost, time and number of animals used in screening everything from pesticides to household chemicals.

The move follows calls for scientists to rely less on animal studies.

Robots would be able to carry out hundreds of thousands of chemical tests a day to identify chemicals with toxic effects.

Details were published in the journal Science and discussed at the annual meeting of the American Association for the Advancement of Science (AAAS) in Boston.

Faster and cheaper

Speaking in a live link-up, Dr. Francis Collins, Director of the National Human Genome Research Institute at the National Institute of Health (NIH), said



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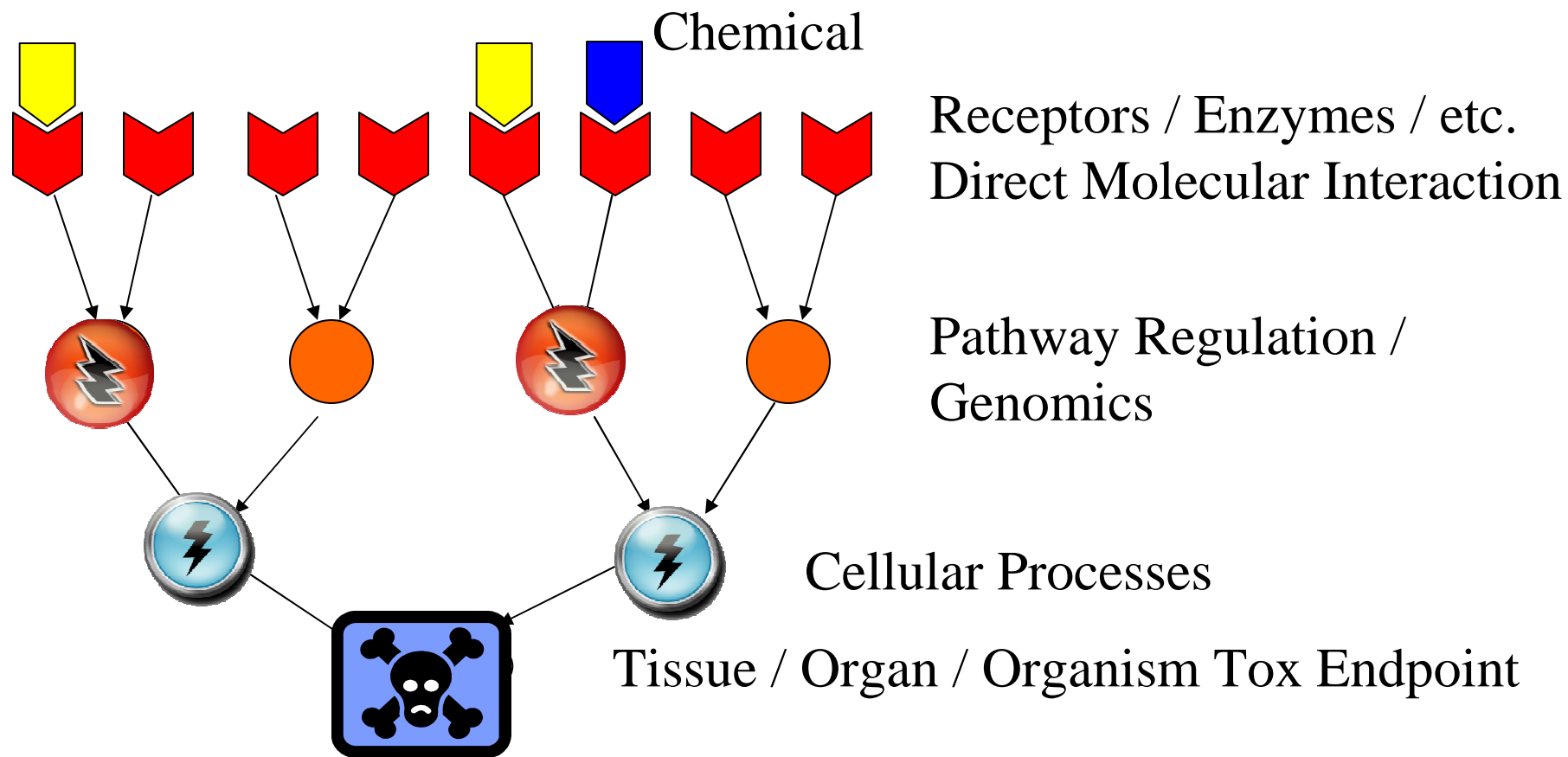
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# Toxicity Pathways

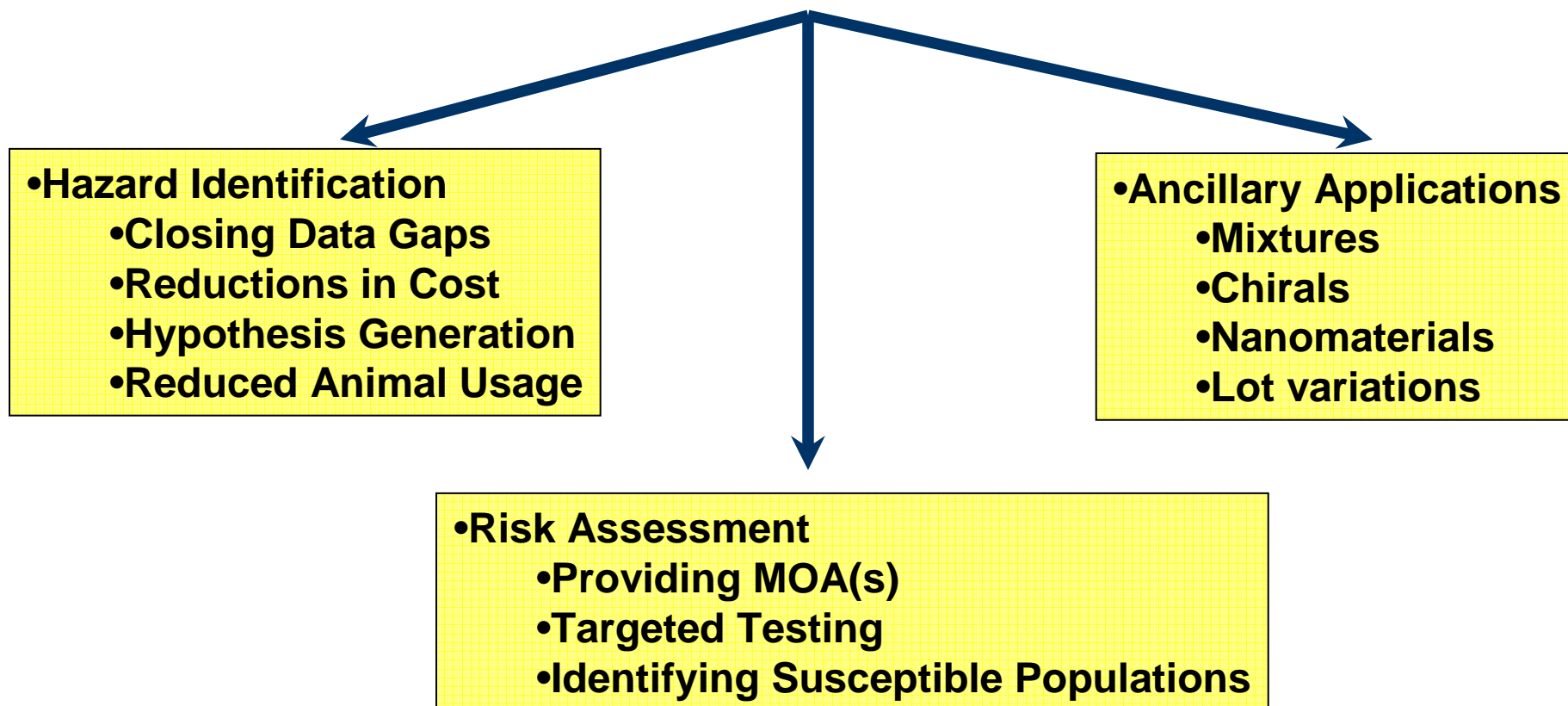




# Key Challenges

- Find the Toxicity Pathways
  - Hepato vs developmental
- Obtain HTS Assays for Them
  - Including metabolic capability
- Screen Chemical Libraries
  - Coverage of p-chem properties
- Link Results to in vivo Effects
  - Gold standard and dosimetry

# Implications for Success



# Phased Development of ToxCast

Phase	Number of Chemicals	Chemical Criteria	Purpose	Number of Assays	Cost per Chemical	Target Date
I	320	Data Rich (pesticides)	Signature Development	>400	\$20k	FY07-08
Ila	>300	Data Rich Chemicals	Validation	>400	\$15-20k	FY09
Ilb	>100	Known Human Toxicants	Extrapolation	>400	\$15-20k	FY09
Ilc	>300	Expanded Structure and Use Diversity	Extension	>400	\$15-20k	FY10
III	Thousands	Data poor	Prediction and Prioritization	???	\$10-15k	FY11-12

- Affordable science-based system for categorizing chemicals
- Increasing confidence as database grows
- Identifies potential mechanisms of action
- Refines and reduces animal use for hazard ID and risk assessment

## Key Components of a Proof of Concept

- Chemicals
- Assays covering Toxicity Pathways
- Linkage to Traditional Phenotype Findings
- Data Analysis and Interpretation

# The ToxCast\_320

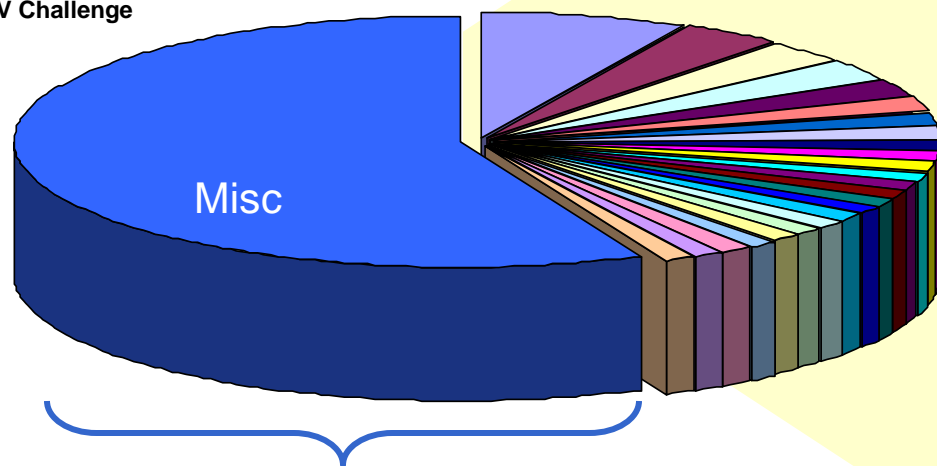
309 Unique Structures

Replicates for QC

291 Pesticide Actives  
9 Industrial Chemicals  
8 Metabolites

56/73 Proposed Tier 1  
EDSP

14 HPV  
11 HPV Challenge



Classes with  
> 3 chemicals

Misc MOA classes with  
3 or fewer representatives

- Acetylcholine esterase inhibitors
- conazole fungicides
- Sodium channel modulators
- pyrethroid ester insecticides
- organothiophosphate acaricides
- dinitroaniline herbicides
- pyridine herbicides
- thiocarbamate herbicides
- imidazolinone herbicides
- organophosphate insecticides
- phenyl organothiophosphate insecticides
- aliphatic organothiophosphate insecticides
- amide herbicides
- aromatic fungicides
- chloroacetanilide herbicides
- chlorotriazine herbicides
- growth inhibitors
- organophosphate acaricides
- oxime carbamate insecticides
- phenylurea herbicides
- pyrethroid ester acaricides
- strobilurin fungicides
- unclassified acaricides
- unclassified herbicides





# Extraction of DER information

Chemical Info
Study Design
Treatment Group Info
Treatment-related Effects
Endpoint/ Critical Effects

**STUDY TYPE:** Combined chronic toxicity/oncogenicity feeding – Rat  
OPPTS 870.4300 [§83-5]

**DP BARCODE:** D257223  
**P.C. CODE:** 111901

**SUBMISSION CODE:** S564270  
**TOX. CHEM. NO.:** 497AB

**TEST MATERIAL (PURITY):** Imazalil (purity ≥97.4%)  
**SYNONYMS:** R023979

**CITATION:** Van Deun, K. 1999. Combined oral chronic toxicity/carcinogenicity study with Imazalil in the SPF Wistar rat. Dept. Toxicology, Janssen Res 2340 Beerse, Belgium. Laboratory report number, 3817, June 8, 1999. MRID 44858001. Unpublished.

**SPONSOR:** Janssen Pharmaceutica N.V., 2340 Beerse, Belgium

## EXECUTIVE SUMMARY:

In a chronic toxicity/oncogenicity study (MRID 44858001), Imazalil (≥97.4% a.i.) was administered in the diet to groups of 50 male and 50 female Hannover substrain (SPF) Wistar-derived rats at concentrations of 0, 50, 200, 1200, or 2400 ppm (equivalent to 0.0, 2.7, 10.8, 65.8, and 134.8 mg/kg/day for males and 0.0, 3.6, 14.6, 85.2, and 168.8 mg/kg/day for females) for two years. All rats were observed daily for clinical signs of toxicity and morbidity, weighed weekly, and food consumption monitored biweekly. Blood and urine samples were collected after 6, 12, and 18 months of treatment and at study end. Surviving rats were sacrificed after 104 weeks of treatment. All rats were necropsied and the tissues and organs inspected grossly and microscopically for toxicity-related effects and the carcinogenic potential of Imazalil.

The absolute weights of most organs were decreased while their weights relative to body weight increased for male and female rats in the 1200 and 2400 ppm treatment groups. These effects are considered related to inanition and inappetence and not a direct result of Imazalil treatment. However, effects found in the liver and thyroid was considered directly related to treatment. The absolute liver weight of male rats in the 2400 ppm group was increased while it was decreased in female rats. The associated relative liver weights of male and female rats in the 1200 and 2400 ppm groups were significantly increased 9-26%. In addition, the absolute and relative thyroid weights of male but not female rats in the 1200 and 2400 ppm groups were increased.

The effect of treatment on the liver (males and females) and thyroid (males only) were confirmed microscopically, but had distinct sex-related etiologies. The incidence of clear cell and basophilic foci was equivocal while eosinophilic foci were significantly increased for male rats in the 2400 ppm group. In female rats of the 2400 ppm group, the incidences of clear cell and basophilic foci were significantly decreased but the incidence of eosinophilic foci was unaffected. Also, the incidence of hepatocyte fatty vacuolation was increased only in male rats of the 1200 ppm and 2400 ppm groups while the incidence of pigmentation was increased only in females of the 200, 1200, and 2400 ppm groups. In addition, the location of hepatocellular hypertrophy was distinctly different. Female rats in the 1200 and 2400 ppm groups had significant increases in centriacinar and periacinar hypertrophy while male rats only had centriacinar hypertrophy. Finally, the incidence of thyroid follicular cell hyperplasia was increased only in male rats of the 1200 and 2400 ppm groups.

The lowest observed adverse effect level (LOAEL) for male and female rats was 1200 ppm (65.8 and 85.2 mg/kg/day, respectively) with a corresponding no observed adverse effect level (NOAEL) of 200 ppm (10.8 mg/kg/day males, 14.6 mg/kg/day females). These are based on the effects found on body weight, weight gain, and the macro- and microscopic effects noted in the liver of all rats and the thyroid of male rats.

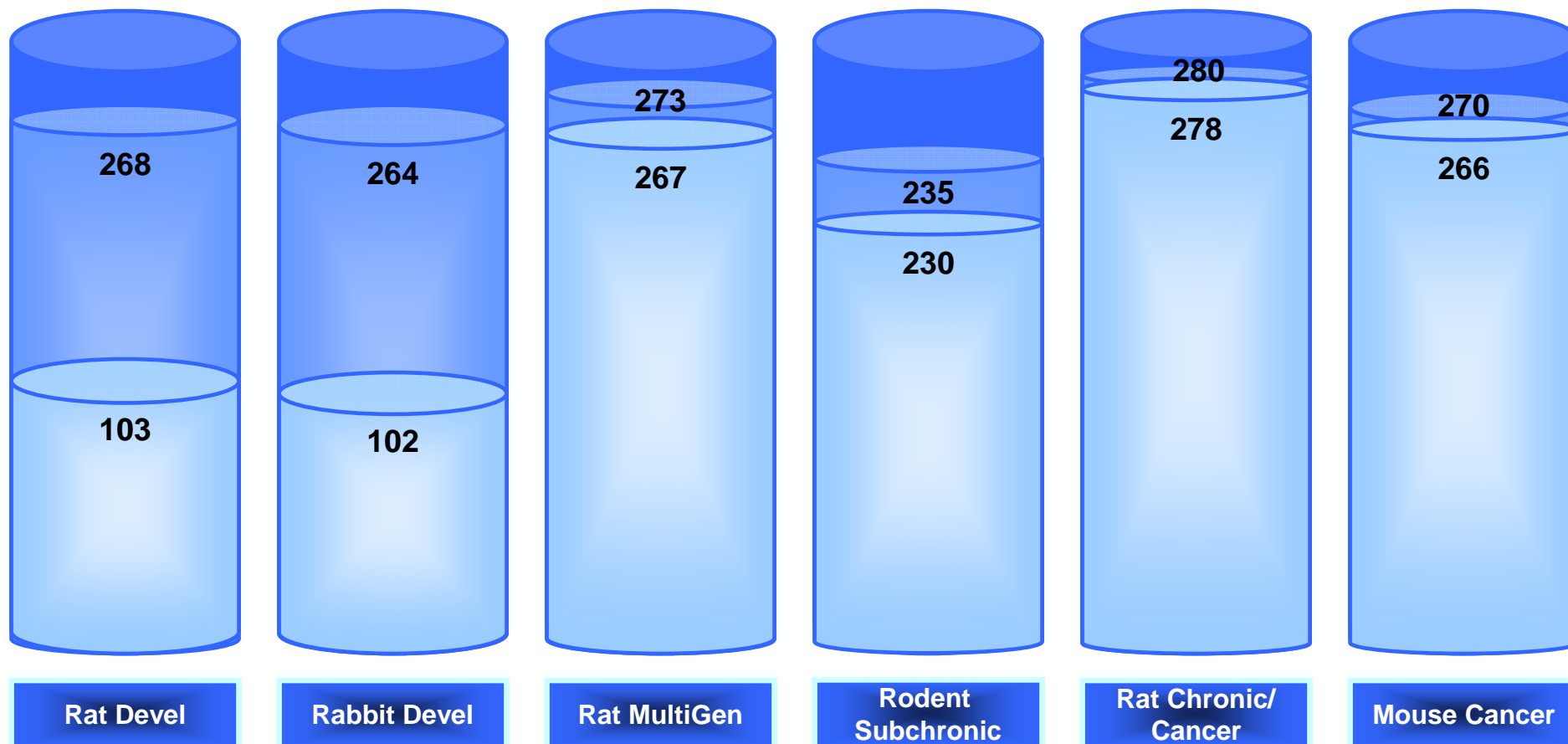
Male rats had a significant increase in the incidence of hepatocellular adenomas and thyroid follicular neoplasia while no increase was found for female rats. These results indicate a difference in the disposition of Imazalil between the sexes increases hepatic and thyroid neoplasia in male rats, likely through differences in metabolic activation of the test material.

This chronic toxicity/oncogenicity study in the rat is Acceptable/guideline and satisfies the guideline requirement for a combined chronic toxicity/oncogenicity study in rats [§3-5]. No deficiencies were noted for this study.

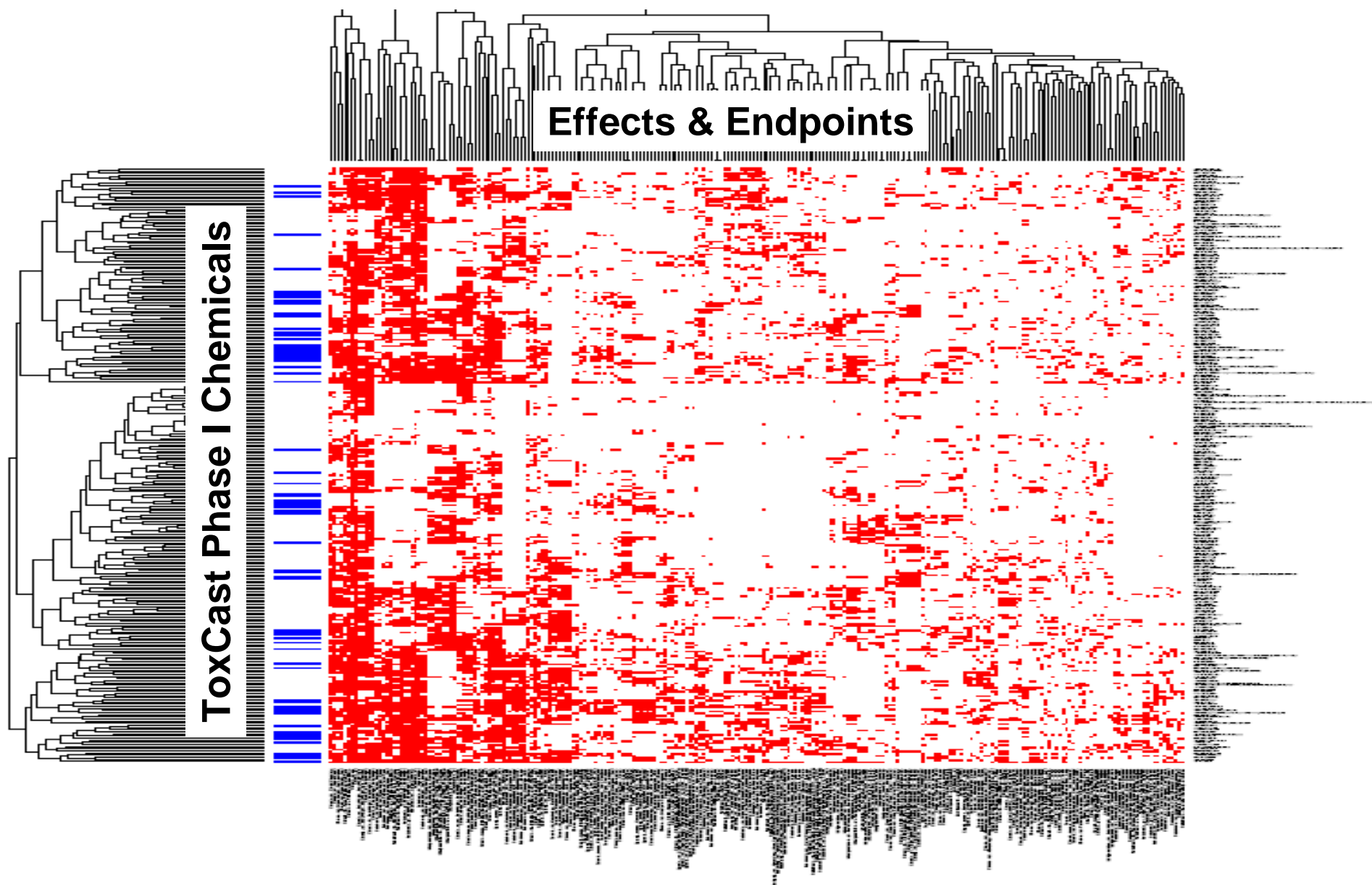
# ToxRefDB Data Entry Status

ToxCast Phase I Chemicals Only

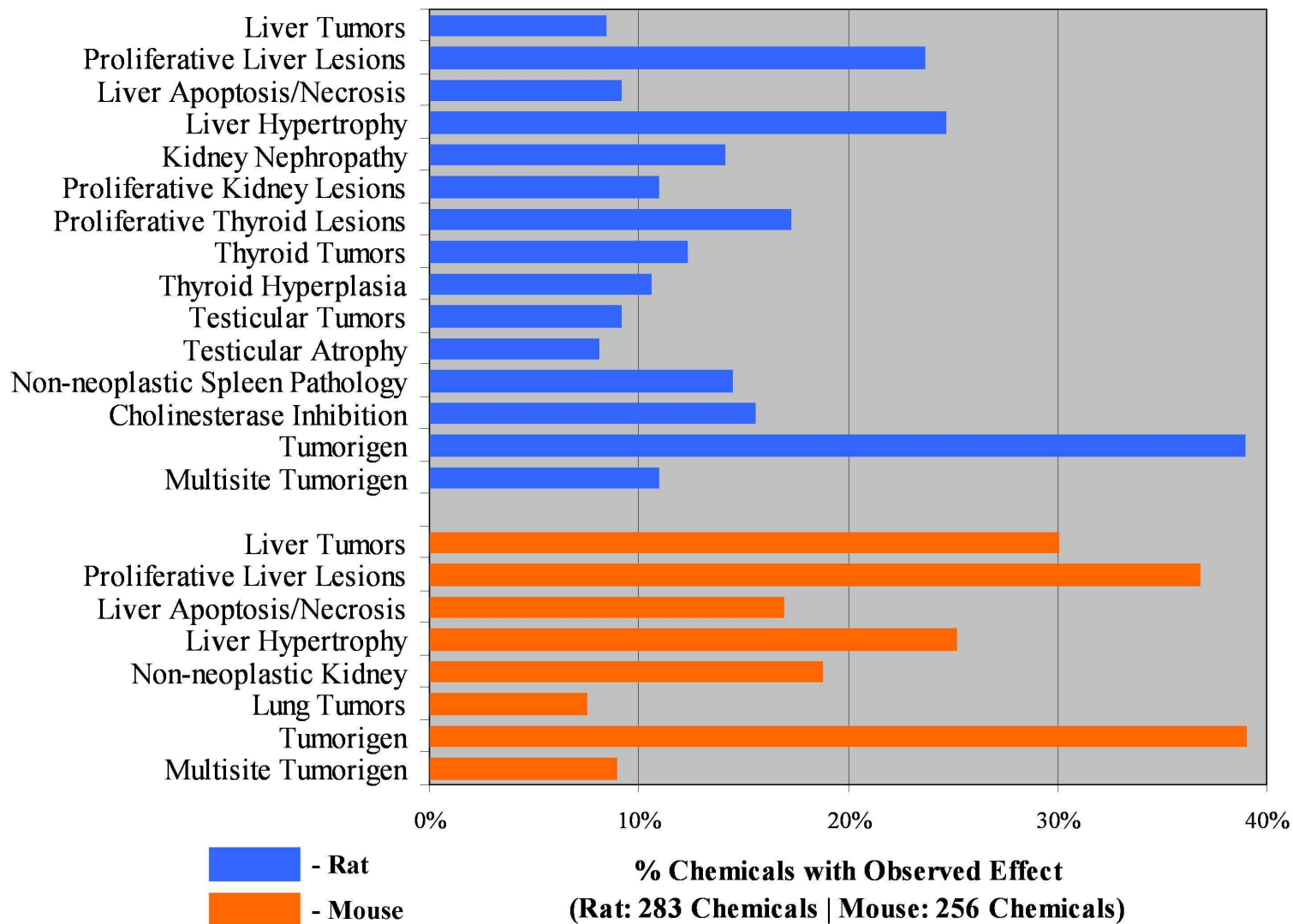
Total: 291 Pesticides



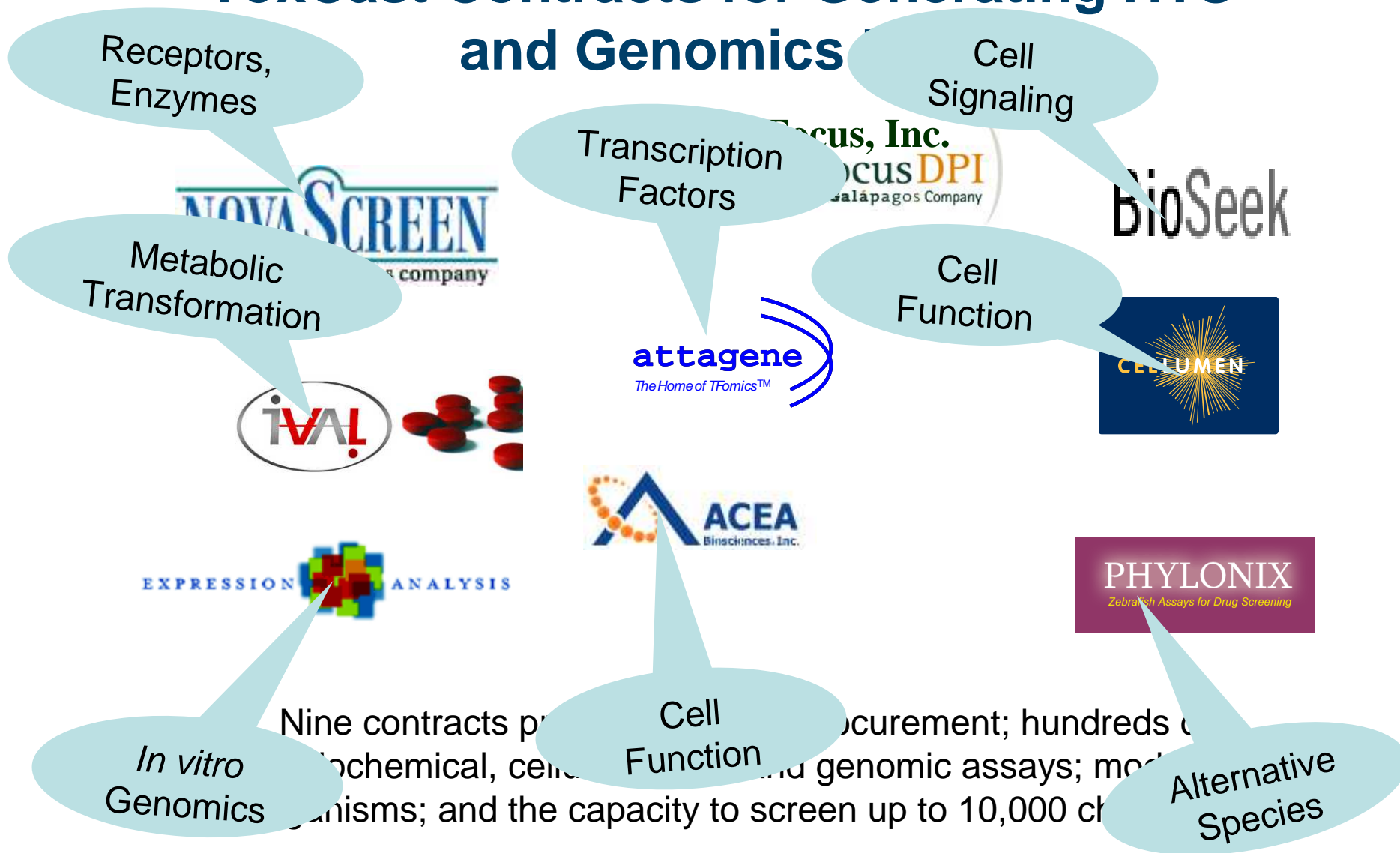
# \$400 Million Dollars Worth of *In Vivo* Chronic/Cancer Bioassay Effects and Endpoints



# Common Phenotypes in Chronic Rodent Studies



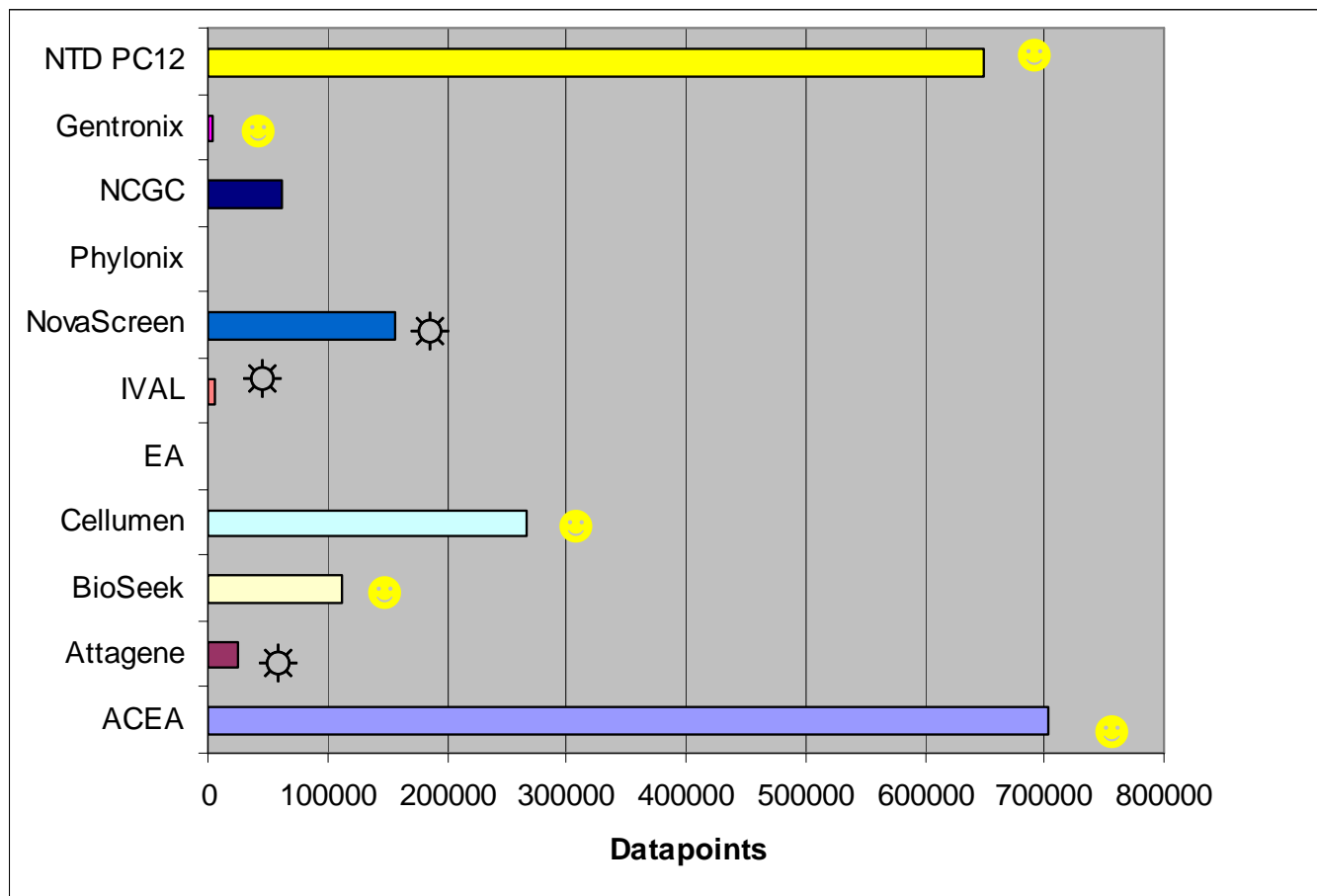
# ToxCast Contracts for Generating HTS and Genomics



Nine contracts provide for the procurement; hundreds of assays; and genomic assays; model organisms; and the capacity to screen up to 10,000 chemicals.



# A Deluge of Data .....



● Data acquisition completed; ⚙ Concentration response follow up underway



320 Chemicals



Transporter

GPCR

Enzyme, other

Ion channel

NR

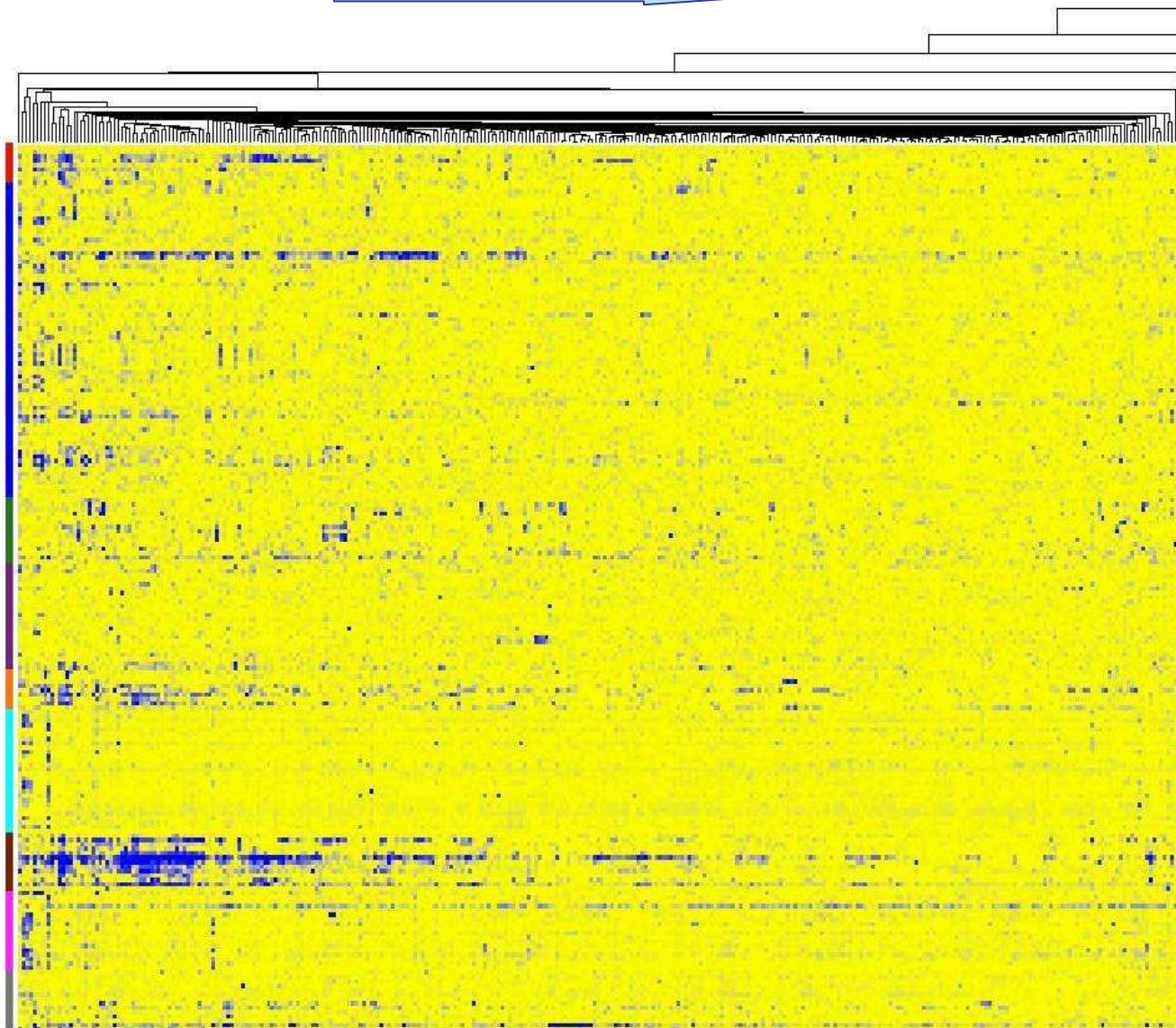
Kinase

CYP450

Phosphatase

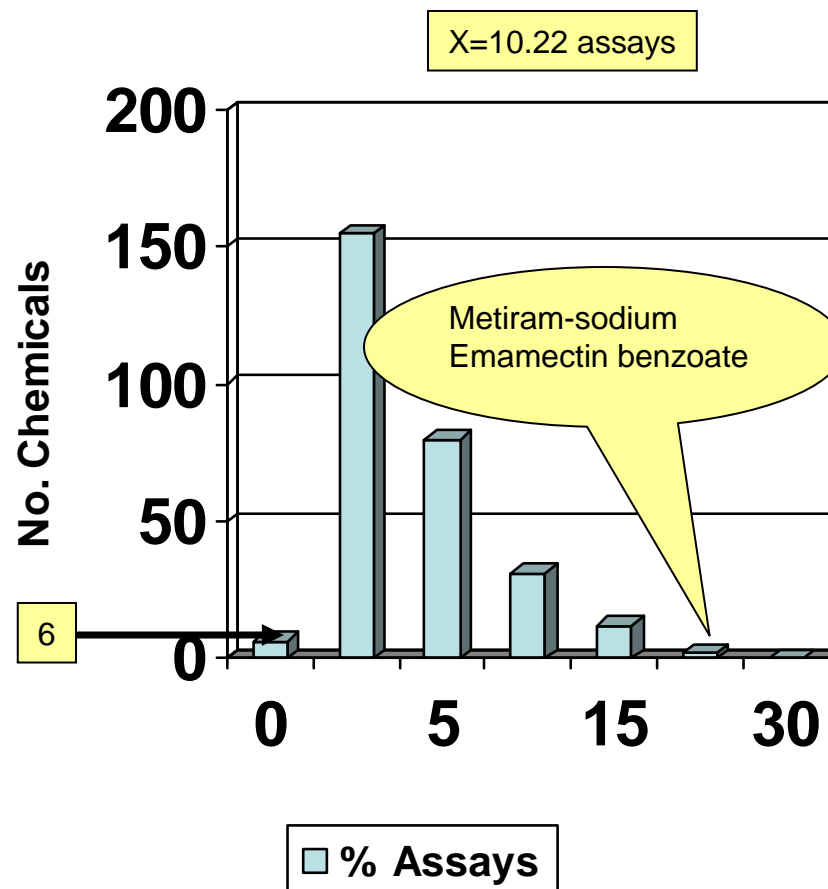
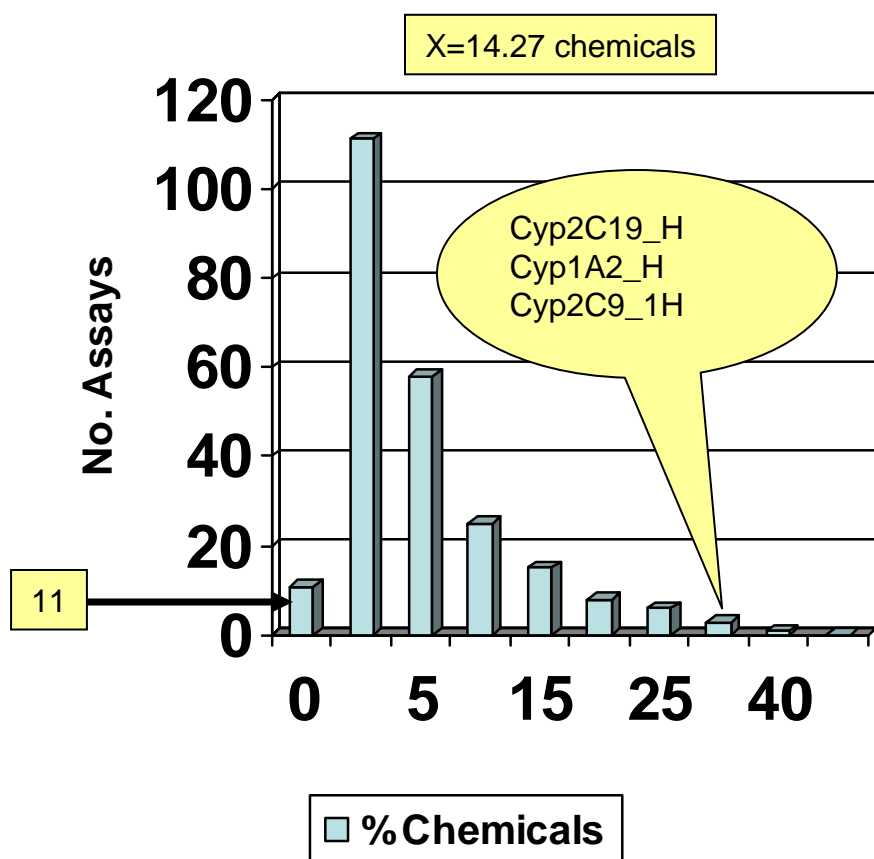
Protease

201 Assays



Activity (% of Control)

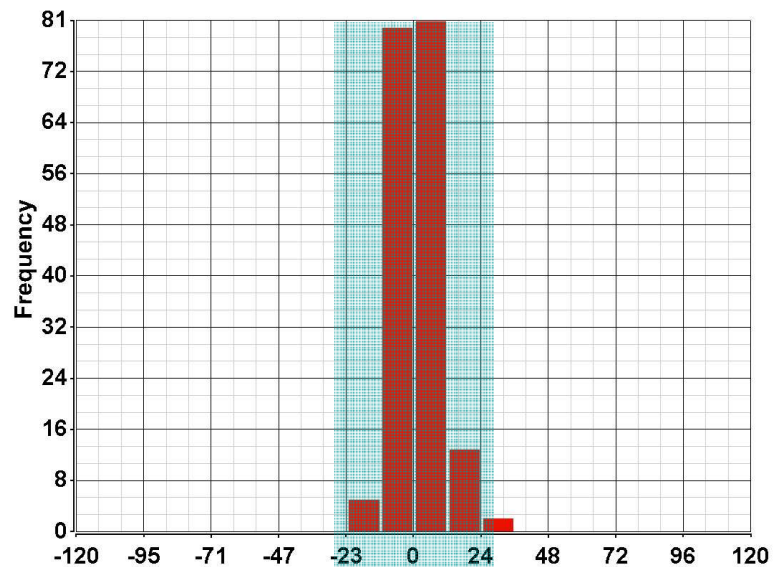
# NovaScreen Descriptive Statistics (30% Cutoff)



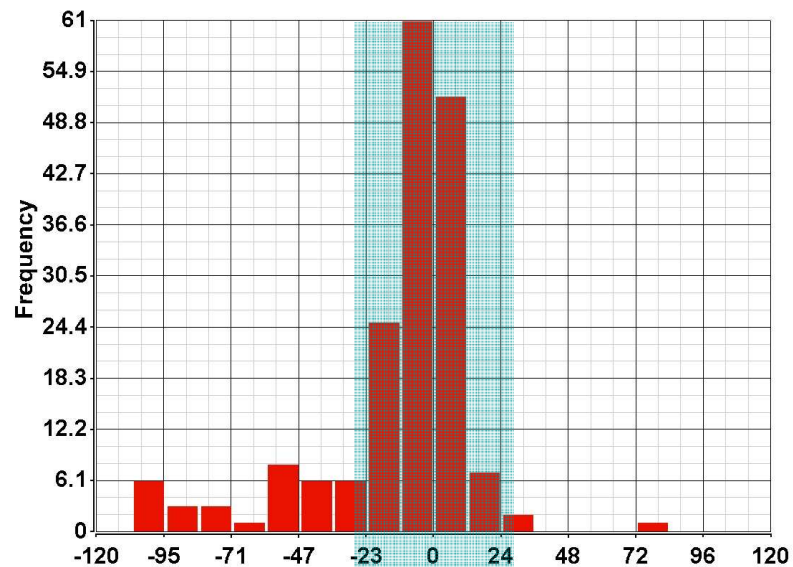


# Examples of Chemical Activity Patterns

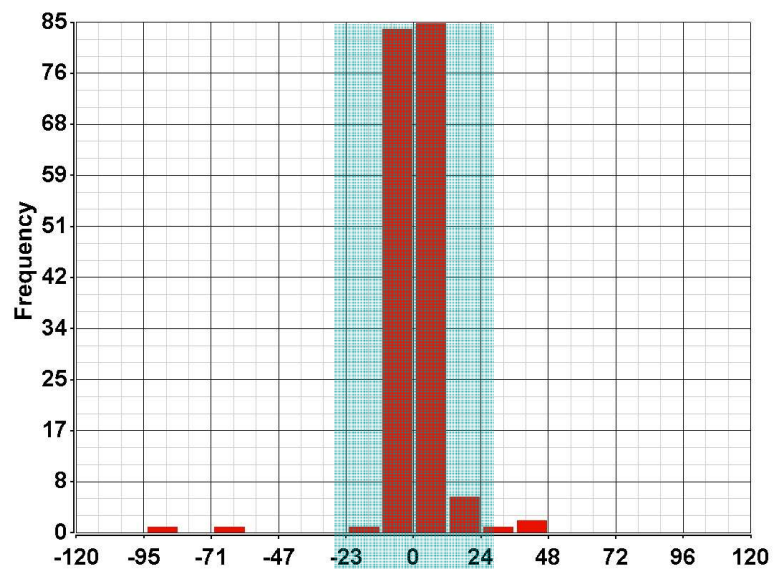
Boric acid



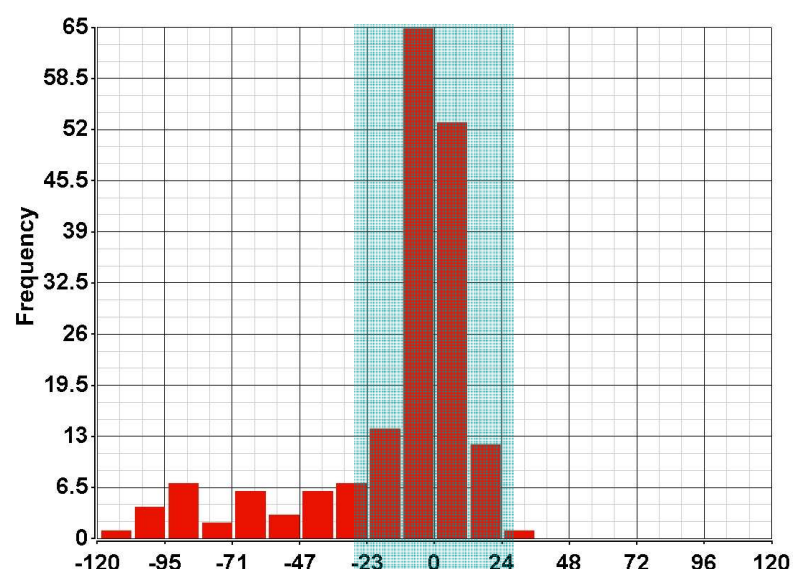
HPTE



Imazapic



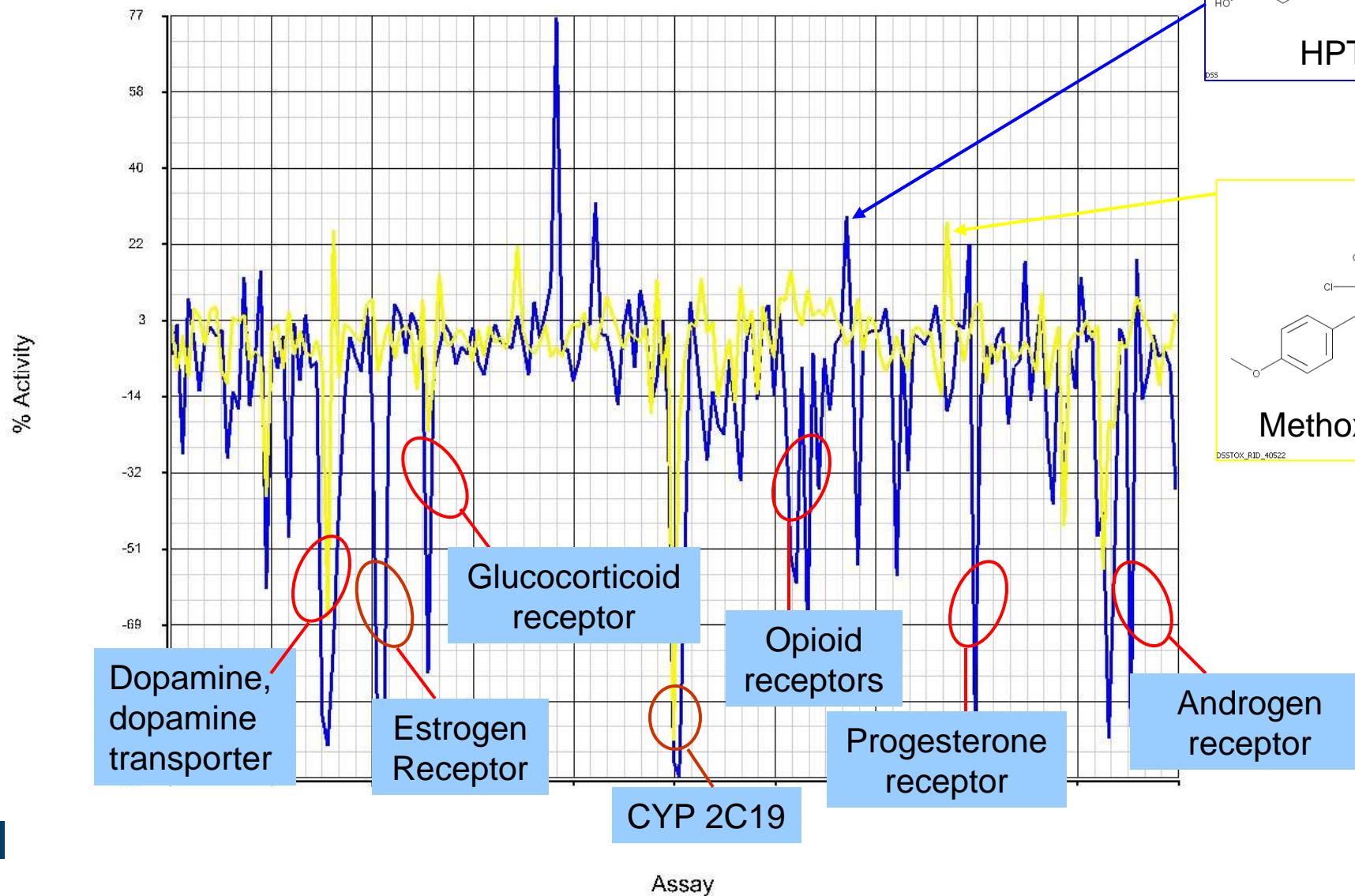
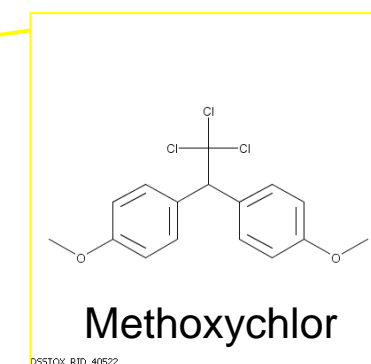
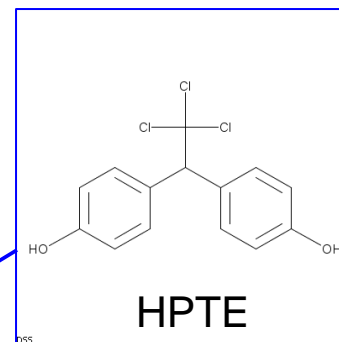
Fentin







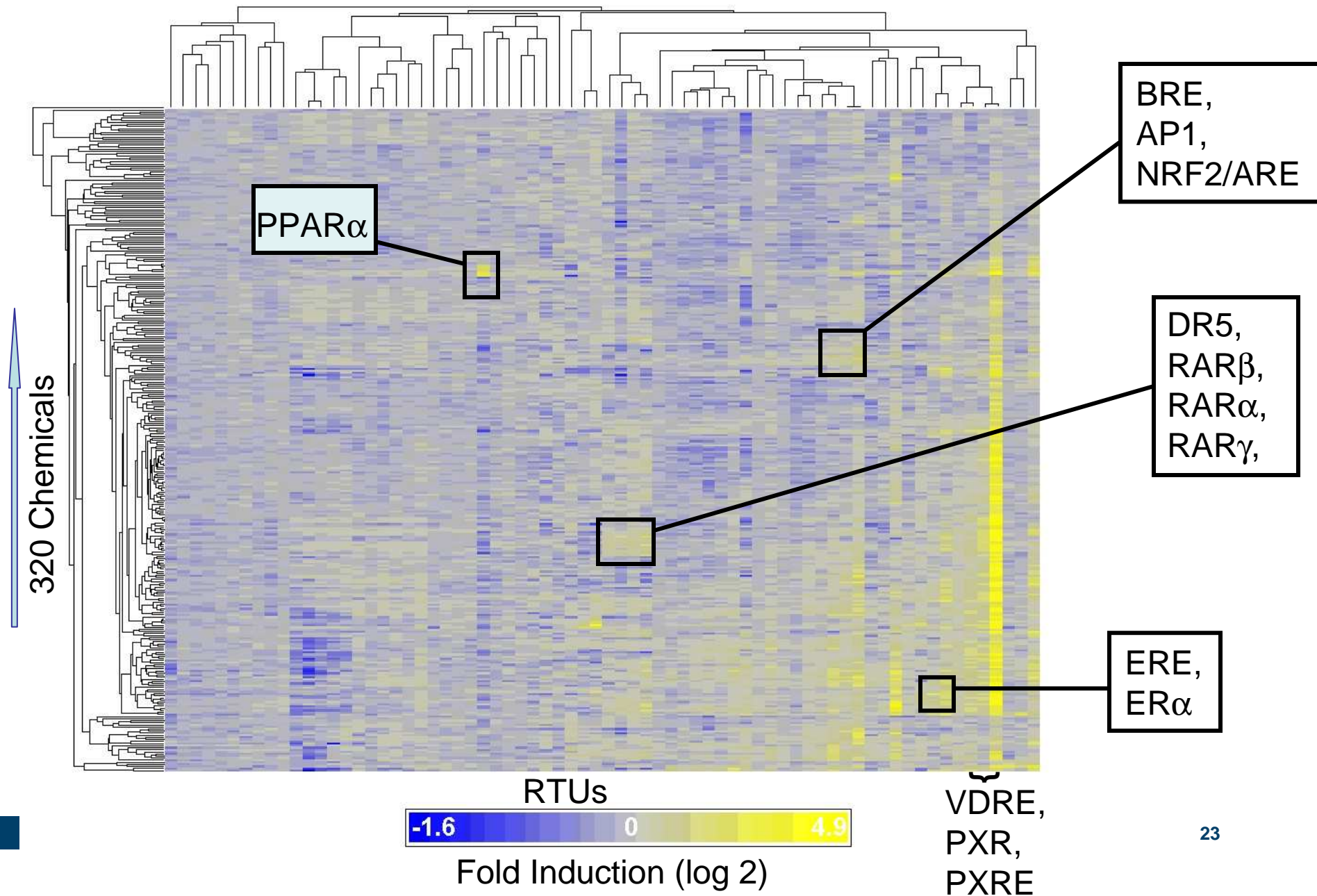
# Activity Spectrum (Novascreen)















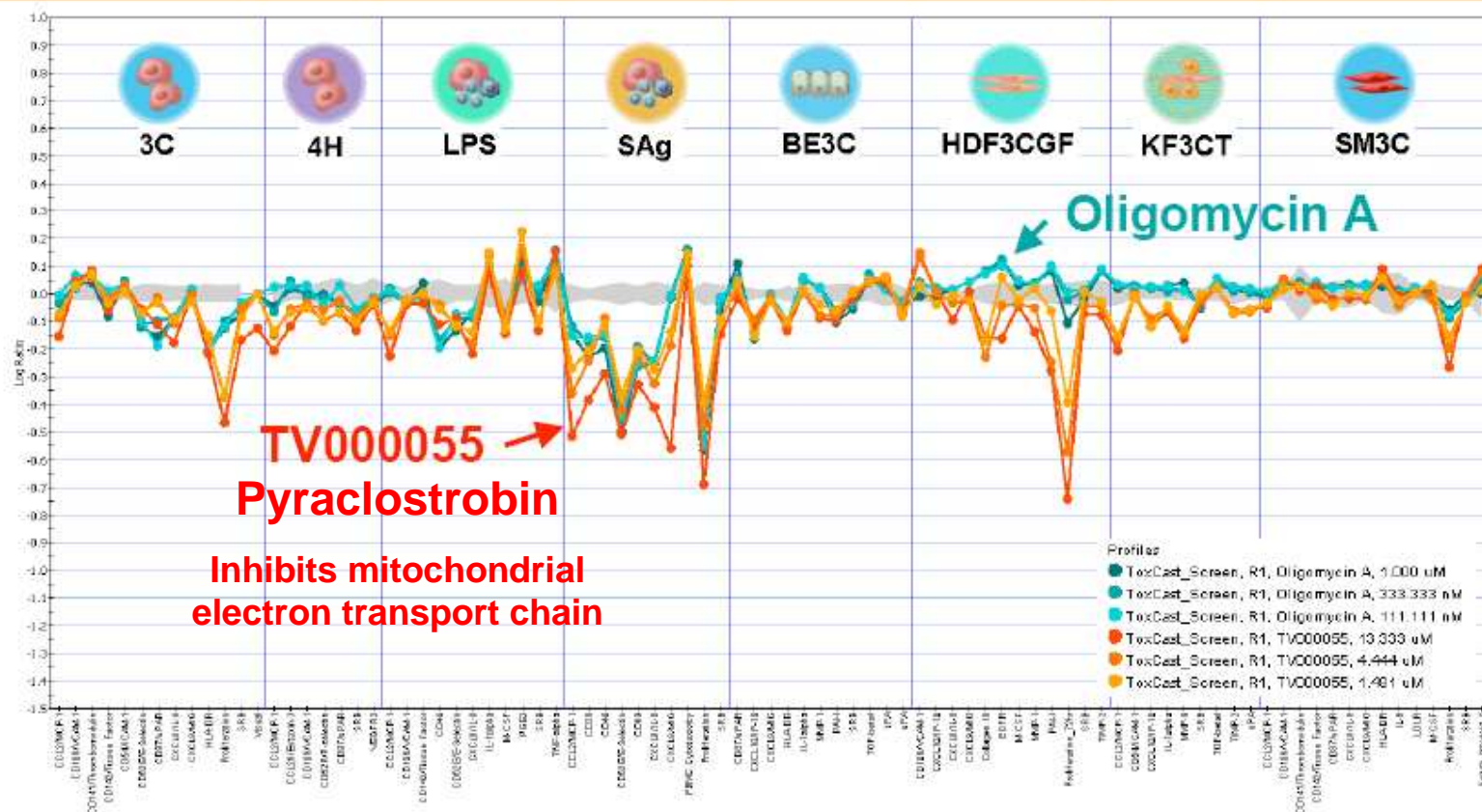
# Hierarchical Cluster Attagene Results



# BioMAP Systems for EPA ToxCast

► System	► Cell Types	Environment	Readout Parameters
3C 	Endothelial cells	IL-1 $\beta$ +TNF- $\alpha$ +IFN- $\gamma$	MCP-1, VCAM-1, ICAM-1, Thrombomodulin, Tissue Factor, E-selectin, uPAR, IL-8, MIG, HLA-DR, Proliferation, Vis., SRB (13)
4H 	Endothelial cells	IL-4+histamine	VEGFR11, P-selectin, VCAM-1, uPAR, Eotaxin-3, MCP-1, SRB (7)
LPS 	Peripheral blood mononuclear cells + Endothelial cells	TLR4	CD40, VCAM-1, Tissue Factor, MCP-1, E-selectin, IL-1 $\alpha$ , IL-8, M-CSF, TNF- $\alpha$ , PGE2, SRB (11)
SAg 	Peripheral blood mononuclear cells + Endothelial cells	TCR	MCP-1, CD38, CD40, CD69, E-selectin, IL-8, MIG, PBMC Cytotox., SRB, Proliferation (10)
BE3C 	Bronchial epithelial cells	IL-1 $\beta$ +TNF- $\alpha$ +IFN- $\gamma$	uPAR, IP-10, MIG, HLA-DR, IL-1 $\alpha$ , MMP-1, PAI-1, SRB, TGF-b1, tPA, uPA (11)
HDF3CGF 	Fibroblasts	IL-1 $\beta$ +TNF- $\alpha$ +IFN- $\gamma$ +bFGF+EGF+PDGF-BB	VCAM-1, IP-10, IL-8, MIG, Collagen III, M-CSF, MMP-1, PAI-1, Proliferation, TIMP-1, EGFR, SRB (12)
KF3CT 	Keratinocytes + Fibroblasts	IL-1 $\beta$ +TNF- $\alpha$ +IFN- $\gamma$ +TGF- $\beta$	MCP-1, ICAM-1, IP-10, IL-1 $\alpha$ , MMP-9, TGF- $\beta$ 1, TIMP-2, uPA, SRB (9)
SM3C 	Vascular smooth muscle cells	IL-1 $\beta$ +TNF- $\alpha$ +IFN- $\gamma$	MCP-1, VCAM-1, Thrombomodulin, Tissue Factor, IL-6, LDLR, SAA, uPAR, IL-8, MIG, HLA-DR, M-CSF, Prolif., SRB (14)

# BioMAP Profiles of Oligomycin A and TV000055

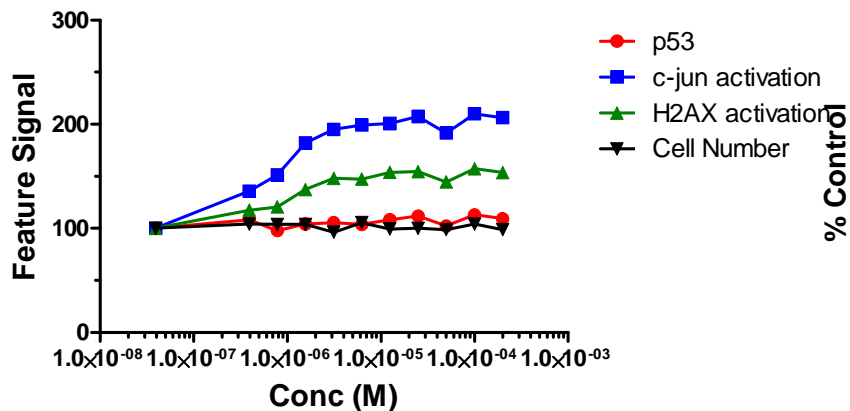


- Oligomycin A is an inhibitor of mitochondrial ATPase
- Similarity suggests inhibition of mitochondrial function by TV000055
  - (TV000055 is most similar to Complex I inhibitors)

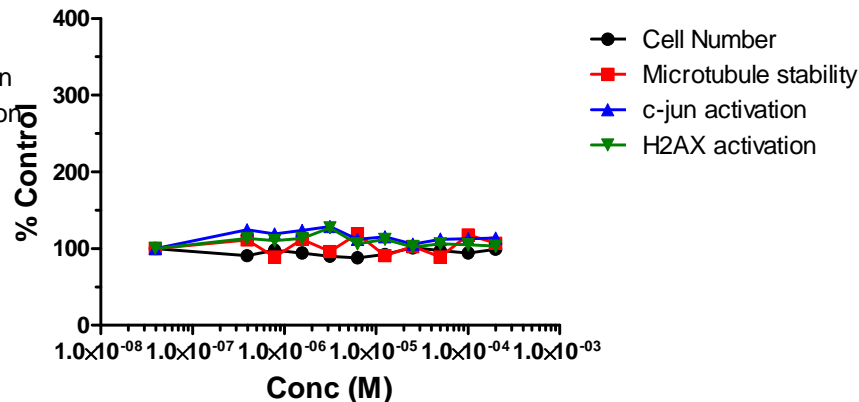


# Examples of Chemical Responses in HepG2 High Content Screening Assays

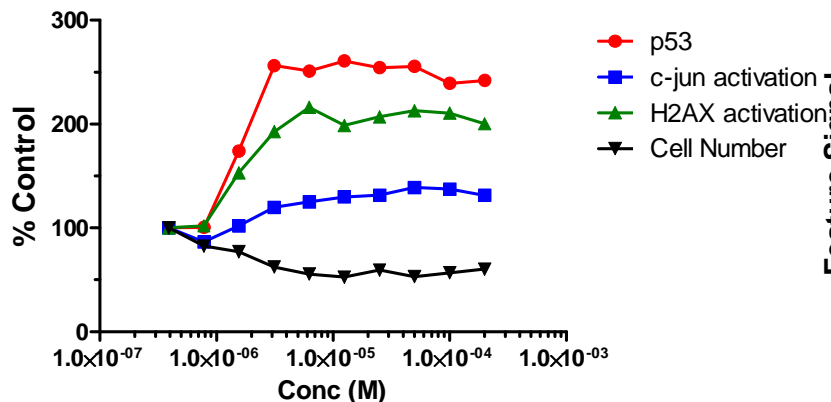
TV000126  
1 Hr



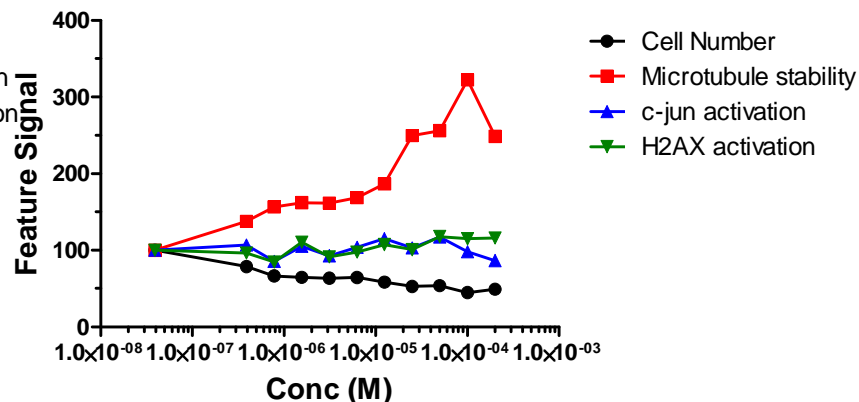
TV000303  
1 Hr



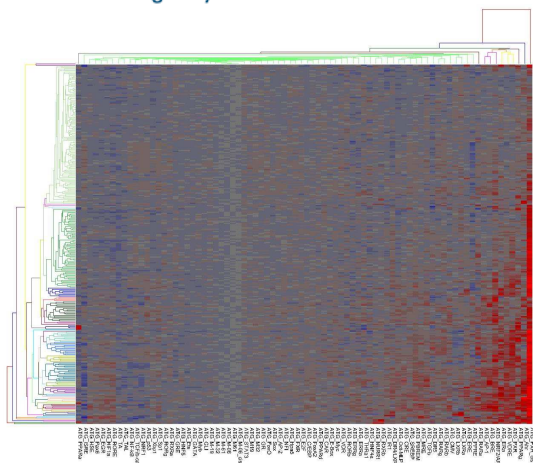
TV000126  
24 Hr



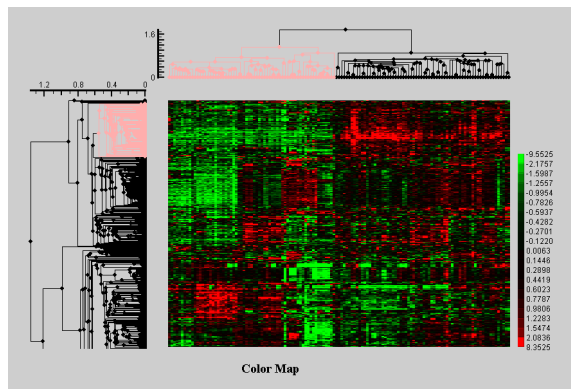
TV000303  
24 Hr



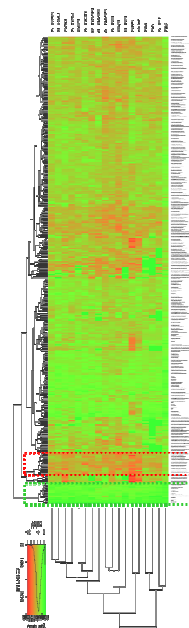
# Correlating HTS to Toxicity



Cellular Assays

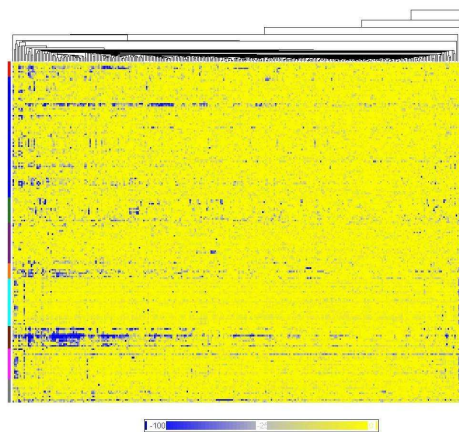


Physical chemical properties

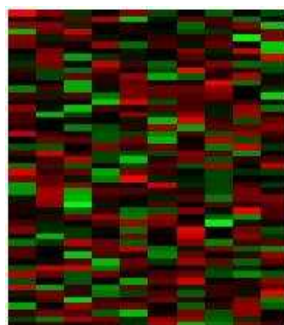


In silico Predictions

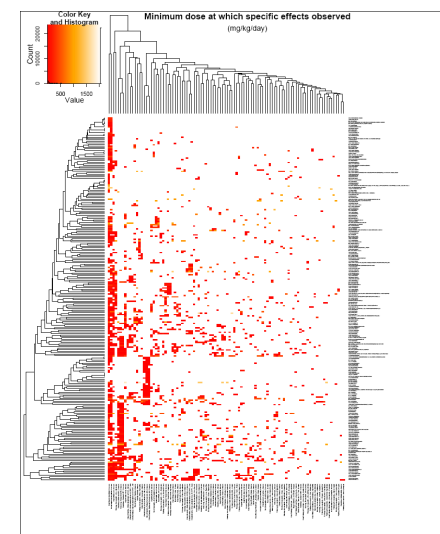
**Profile Matching**



Biochemical Assays



Genomic Signatures



Toxicology Endpoints



## ACToR: Aggregated Computational Toxicology Resource

# ACToR: Aggregated Computational Toxicology Resource

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## Data Collection: ToxCast\_320

SCID	GCID	CASRN	Name	Hazard	AcuteTox	SubchronicTox	ChronicTox	GenTox	DevTox	ReproTox	NeuroTox	ImmuTox	DermTox	RespiTox	HepatoTox	Endocrine	CardioTox	Ecotox	ToxOther
<a href="#">12622</a>	<a href="#">447</a>	94-75-7	2,4-D	11	6	1	7	16	23	8	8	4	3	2	1	2	1	2	1
<a href="#">12623</a>	<a href="#">6424</a>	94-82-6	2,4-DB	8	4	1	4	8	7	6	5	2	2		1	2	1	2	
<a href="#">12624</a>	<a href="#">7712</a>	136-45-8	2,5-Pyrimedicarboxylic acid, dipropyl ester	3	1	1	1	5	2	1									
<a href="#">12625</a>	<a href="#">1174</a>	90-43-7	2-Phenylphenol	6	2	1	2	10	1	3	2	1	1				1	1	
<a href="#">12626</a>	<a href="#">4555</a>	55406-53-6	3-Iodo-2-propynylbutylcarbamate	6	2	1	2	3	3	2	2								
<a href="#">12627</a>	<a href="#">4555</a>	55406-53-6	3-Iodo-2-propynylbutylcarbamate	6	2	1	2	3	3	2	2								

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ToxRefDB

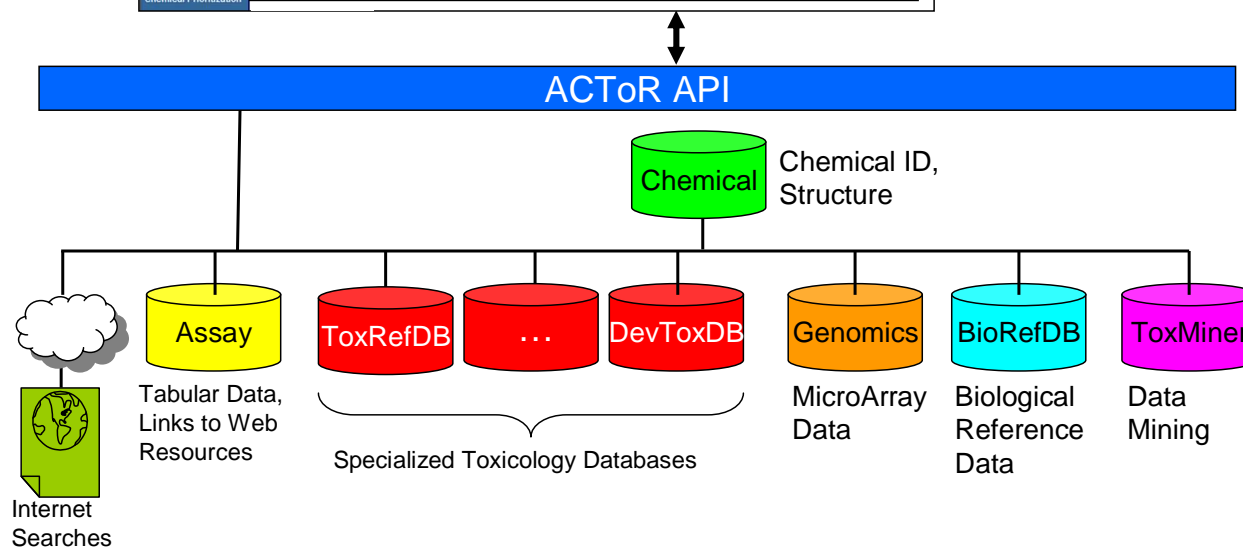
DSSTox

ToxMiner

ToxCast

Chemical Prioritization

ACToR Web  
Browser



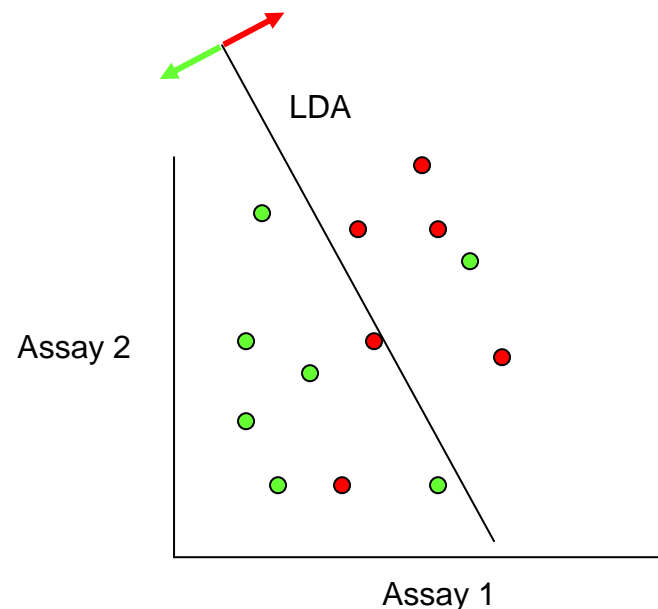
# Comparing Activities by Chemical Class

Conazole  
Fungicides vs.  
NovaScreen  
Assays

NAME	CYP2C19	CYP2C9	CYP3A1	Dopamine Transporter (Human)	CYP2D2	Androgen Receptor	Dopamine Transporter (Rat)	CYP2B6	CYP2D1	CYP3A4	Progesterone Receptor	Benzodiazepine Receptor
Cyproconazole	1	1	1	1	1	0	1	0	0	1	0	0
Difenoconazole	1	1	1	1	1	0	0	1	1	0	0	0
Diniconazole	1	1	1	0	1	0	0	0	1	1	1	0
Fenbuconazole	1	1	0	0	0	0	0	0	0	1	0	0
Flusilazole	1	1	1	0	1	1	0	1	1	NA	1	1
Hexaconazole	1	1	1	1	1	0	1	1	1	NA	1	0
Imazalil	1	1	1	1	1	1	1	1	1	1	1	1
Myclobutanil	1	1	1	1	0	0	0	0	0	NA	0	0
Paclobutrazol	1	0	1	1	0	1	1	0	1	1	0	0
Prochloraz	1	1	1	1	1	1	1	1	1	NA	1	1
Propiconazole	1	1	1	0	0	0	0	1	0	NA	0	1
Tetraconazole	1	1	1	0	1	1	0	1	0	1	1	0
Triadimefon	1	1	0	1	1	1	1	0	0	1	0	1
Triadimenol	1	0	0	1	0	1	1	0	0	0	0	0
Triflumizole	1	1	1	1	1	1	0	1	1	1	1	1
Triticonazole	1	1	1	1	0	1	1	0	0	NA	0	0
Totals	16	14	13	11	10	9	8	8	8	8	7	6

# Association Analysis / Signatures

- Use Machine Learning methods
  - SLR: Stepwise Logistic Regression
  - LDA: Linear Discriminant Analysis
  - SVM: Support Vector Machines
  - Many others
- For each binary endpoint, build models of form
  - $Predictor = F(\text{assay values})$
  - If
    - $Predictor$  for a chemical meets criteria
  - Then
    - Predict endpoint to be positive for the chemical



	Truth	
	+	-
Test	+	TP
	-	FN

# Example of Signature Development: Pesticide MOA

MOA	Chemicals	Positives	Sensitivity	Specificity	PPV	NPV	Accuracy
thiocarbamate herbicides	303	6	1.00	0.99	0.70	1.00	0.99
dinitroaniline herbicides	303	7	1.00	0.98	0.61	1.00	0.99
Sodium channel modulators	303	11	0.90	0.97	0.51	1.00	0.93
pyrethroid ester insecticides	303	10	0.65	0.98	0.62	0.99	0.81
conazole fungicides	303	13	0.65	0.97	0.52	0.99	0.81
pyridine herbicides	303	6	0.60	0.99	0.67	0.99	0.79
Sodium channel modulators	303	11	0.60	0.98	0.53	0.99	0.79
conazole fungicides	303	13	0.50	0.98	0.53	0.98	0.74
Acetylcholine esterase inhibitors	303	27	0.50	0.97	0.66	0.96	0.74
Acetylcholine esterase inhibitors	303	27	0.52	0.95	0.57	0.96	0.73
pyrethroid ester insecticides	303	10	0.50	0.95	0.32	0.98	0.73
organothiophosphate acaricides	303	9	0.00	1.00	0.00	0.98	0.50
organothiophosphate acaricides	303	9	0.00	1.00	0.00	0.98	0.50
pyridine herbicides	303	6	0.00	1.00	0.00	0.98	0.50
thiocarbamate herbicides	303	6	0.00	1.00	0.00	0.98	0.50
dinitroaniline herbicides	303	7	0.00	0.99	0.00	0.98	0.49

Input variables: NovaScreen, Attagene, Bioseek and physical chemical properties

# Evolution of Phase I

- **ToxCast 1.0 (April, 2007)**

- Enzyme inhibition/receptor binding HTS (Novascreen)
- NR/transcription factors (Attagene, NCGC)
- Cellular impedance (ACEA)
- Complex cell interactions (BioSeek)
- Hepatocellular HCS (Cellumen)
- Hepatic, renal and airway cytotoxicity (IVAL)
- In vitro hepatogenomics (IVAL, Expression Analysis)
- Zebrafish developmental toxicity (Phylonix)

**8 Assay Sources  
& 412 Endpoints**

- **ToxCast 1.1 (January, 2008)**

- Neurite outgrowth HCS (NHEERL)
- Cell proliferation (NHEERL)
- Zebrafish developmental toxicity (NHEERL)

**+3 Assay Sources  
& 16 Endpoints**

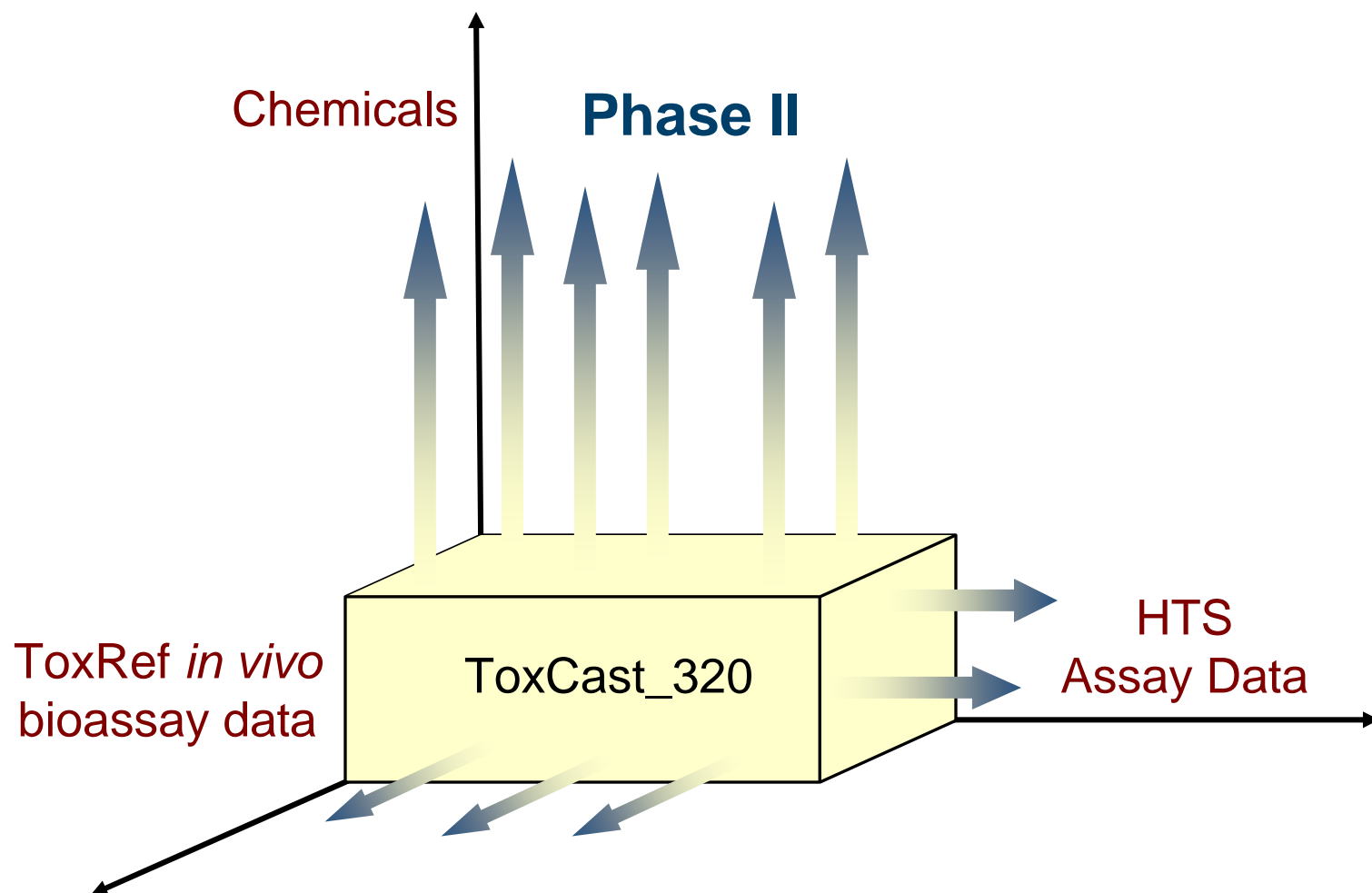
- **ToxCast 1.2 (March, 2008)**

- Organ culture: liver, kidney, lung (Hamner Institutes)
- HTS Genotoxicity (Gentronix)
- Toxicity and signaling pathways (Invitrogen)
- NR Activation and translocation (CellzDirect)
- 3D Cellular microarray with metabolism (Solidus)

**+5 Assay Sources  
& 32 Endpoints**

**16 Assay Sources, 460 Endpoints**

# Beyond the Proof of Concept



# Moving Forward

- Completion of Data Acquisition and Data Mining for Phase I
- Publication and Public Release of all Data
  - 8 core papers by mid Summer (including 2 on ToxRefDB)
  - Predictive Signatures by September
  - Additional partner papers by Fall
  - ToxRefDB and ACToR public release
- OECD Molecular Screening Initiative (June, Bilthoven)
- Data Summit, Fall 2008
- MOU partnership with NTP/NIEHS and NCGC/NHGRI
  - Workings Groups in Pathways, Chemicals, Informatics, [Targeted Testing]
  - Minimum of 2816 additional chemicals to be placed at NCGC
    - Subset to feed Phase II of Toxcast
- EPA Research Strategy and FY10 Research Initiative



## The ToxCast Team





**ToxCast Website: [www.epa.gov/ncct/toxcast](http://www.epa.gov/ncct/toxcast)**

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The EPA Web site will be unavailable on Sunday, March 2, 2008 from 8:00 pm until 10:00 pm ET.

## ToxCast™ Program

**Predicting Hazard, Characterizing Toxicity Pathways, and Prioritizing the Toxicity Testing of Environmental Chemicals**

### Introduction

In 2007, EPA launched ToxCast™ in order to develop a cost-effective approach for prioritizing the toxicity testing of large numbers of chemicals in a short period of time. Using data from state-of-the-art high throughput screening (HTS) bioassays developed in the pharmaceutical industry, ToxCast™ is building computational models to forecast the potential human toxicity of chemicals. These hazard predictions will provide EPA regulatory programs with science-based information helpful in prioritizing chemicals for more detailed toxicological evaluations, and lead to more efficient use of animal testing.

In its first phase, ToxCast™ is profiling over 300 well-characterized chemicals (primarily pesticides) in over 400 HTS endpoints. These endpoints include biochemical assays of protein function, cell-based transcriptional reporter assays, multi-cell interaction assays, transcriptomics on primary cell cultures, and developmental assays in zebrafish embryos. Almost all of the compounds being examined in Phase 1 of ToxCast™ have been tested in traditional toxicology tests, including developmental toxicity, multi-generation studies, and sub-chronic and chronic rodent bioassays. ToxRefDB, a relational database being created to house this information, will contain nearly \$1B worth of toxicity studies in animals when completed. ToxRefDB is integrated into a more comprehensive data management system developed by NCCT called ACToR (Aggregated Computational Toxicology Resource), that manages the large-scale datasets of ToxCast™.

ACToR is comprised of several independent data repositories linked to a common database of chemical structures and properties, and to tools for development of predictive HTS and genomic bioactivity signatures that strongly correlate with specific toxicity endpoints from ToxRefDB. These ToxCast™ signatures will be defined and evaluated by their ability to predict outcomes from existing mammalian toxicity testing, and identify toxicity pathways that are relevant to human health effects.

The second phase of ToxCast™ will screen additional compounds representing broader chemical structure and use classes, in order to evaluate the predictive bioactivity signatures developed in Phase I. Following successful conclusion of Phases I and II, ToxCast™ will provide EPA regulatory programs an efficient tool for rapidly and efficiently screening compounds and prioritizing further toxicity testing.

#### ToxCast™ Navigation

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[ToxCast™ Assays](#)

[ToxCast™ Information Management](#)

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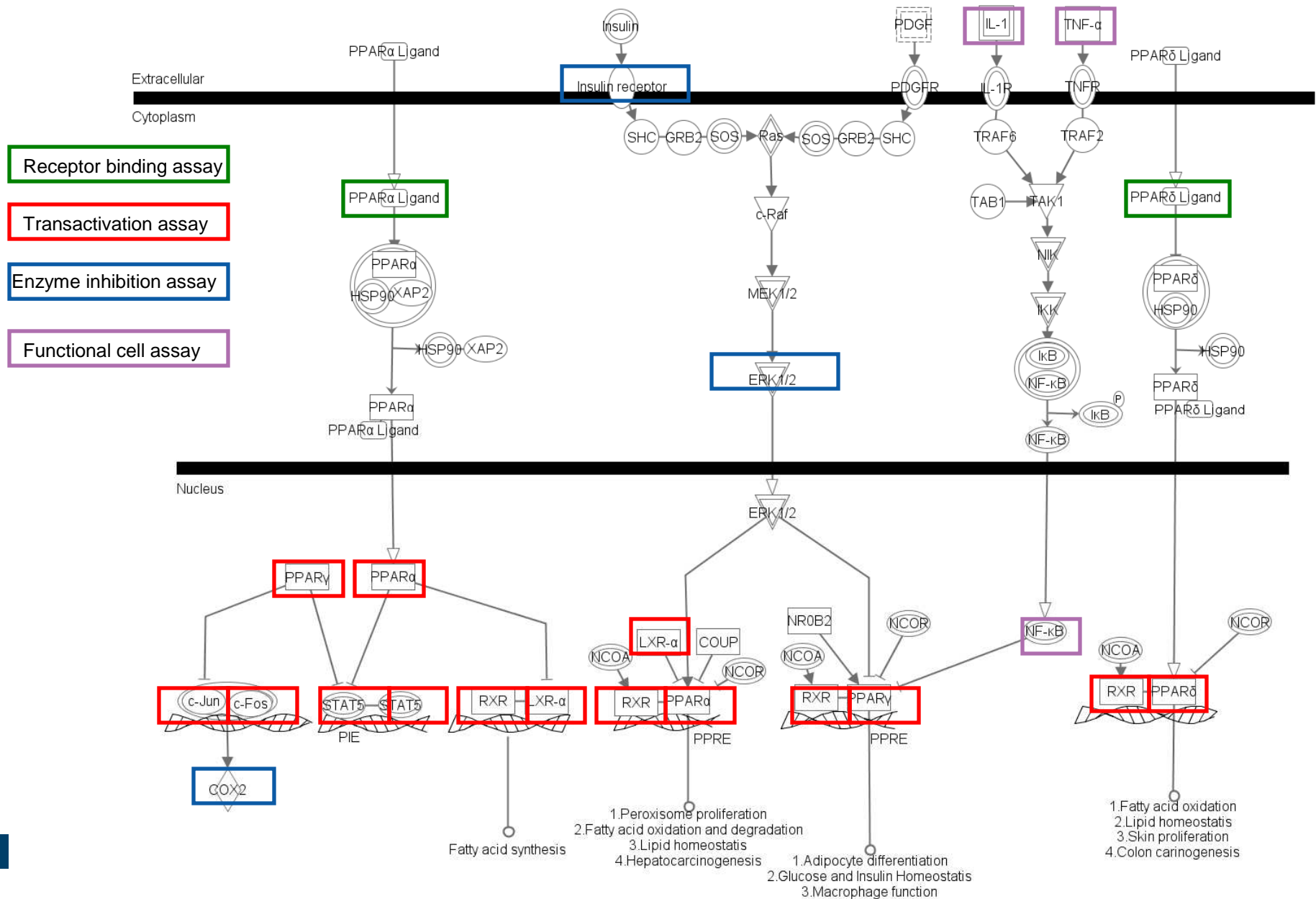
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[ToxCast™ Presentations](#)

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# PPAR SIGNALING PATHWAY





# Virtual Tissues: From Pathways to Dose-Response

Environmental  
Chemicals

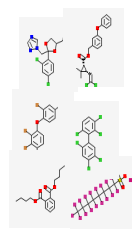
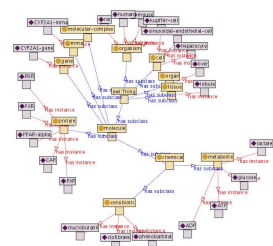
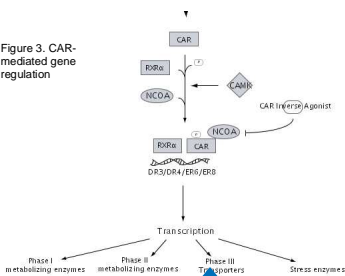
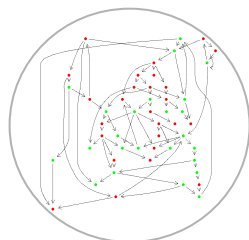


Figure 3. CAR-mediated gene regulation



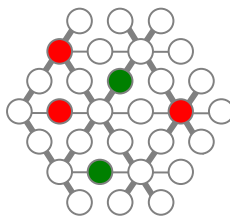
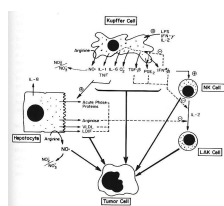
Knowledgebase  
Toxicity  
Pathways

Molecular  
Pathways



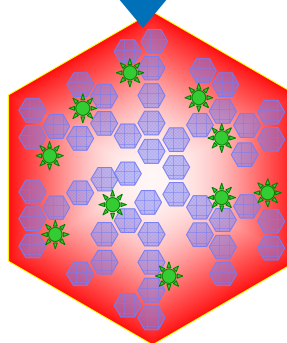
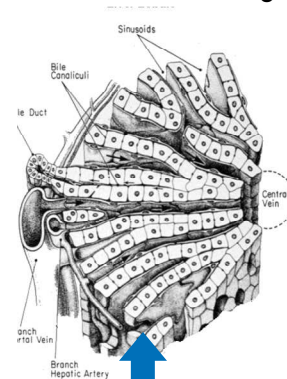
Intracellular  
Pathways

Cellular  
Alterations



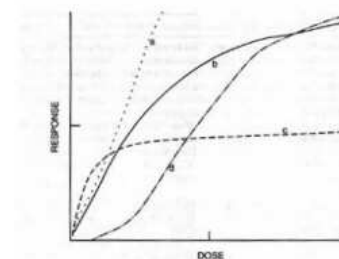
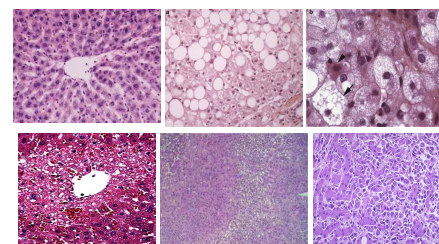
Cellular  
Networks

Tissue  
Changes



Virtual  
Tissues

Tissue  
Injury



Dose-Response