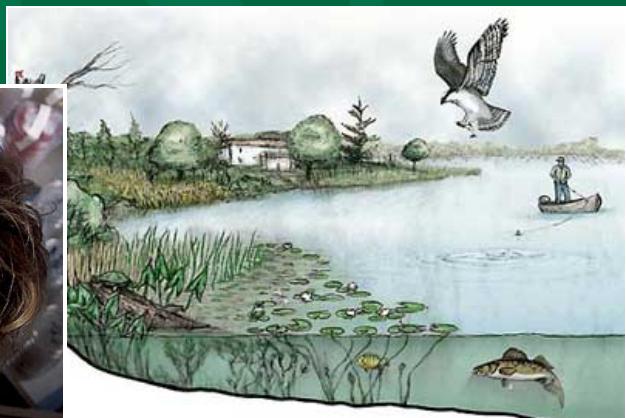


Application of Computational Toxicology to Prospective and Diagnostic Ecological Risk Assessment

**Daniel L. Villeneuve, US EPA, Duluth, MN*

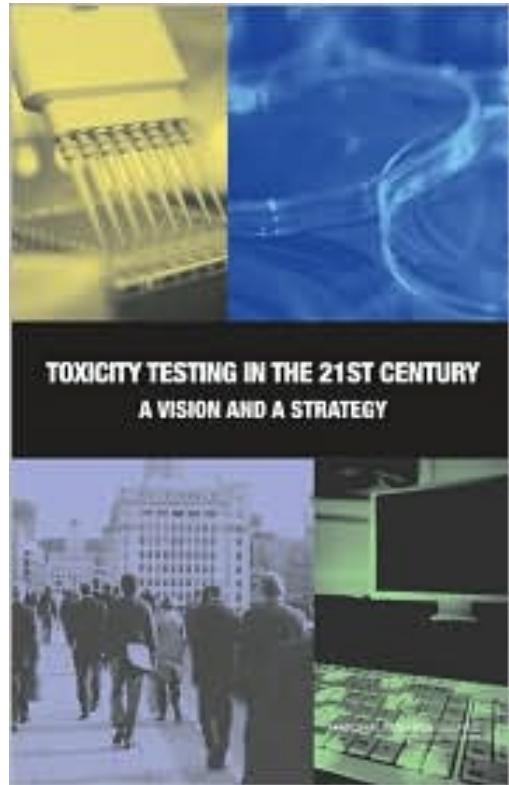


Toxicity Testing in the 20th Century



- Expensive
- Time-consuming
- Animal intensive
- Empirical
 - observation > understanding

Toxicity Testing in the 21st Century



Four competing objectives

- **Depth** – providing the most accurate, detailed, characterization possible.
- **Breadth** – providing data on the broadest universe of chemicals, endpoints, species, life-stages, etc.
- **Animal welfare** – using the fewest animals possible and minimizing suffering.
- **Conservation** – minimizing expenditure of money and time on testing and review.

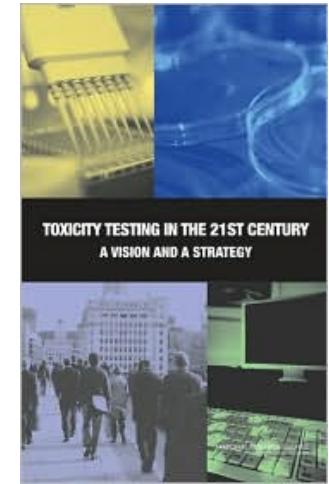
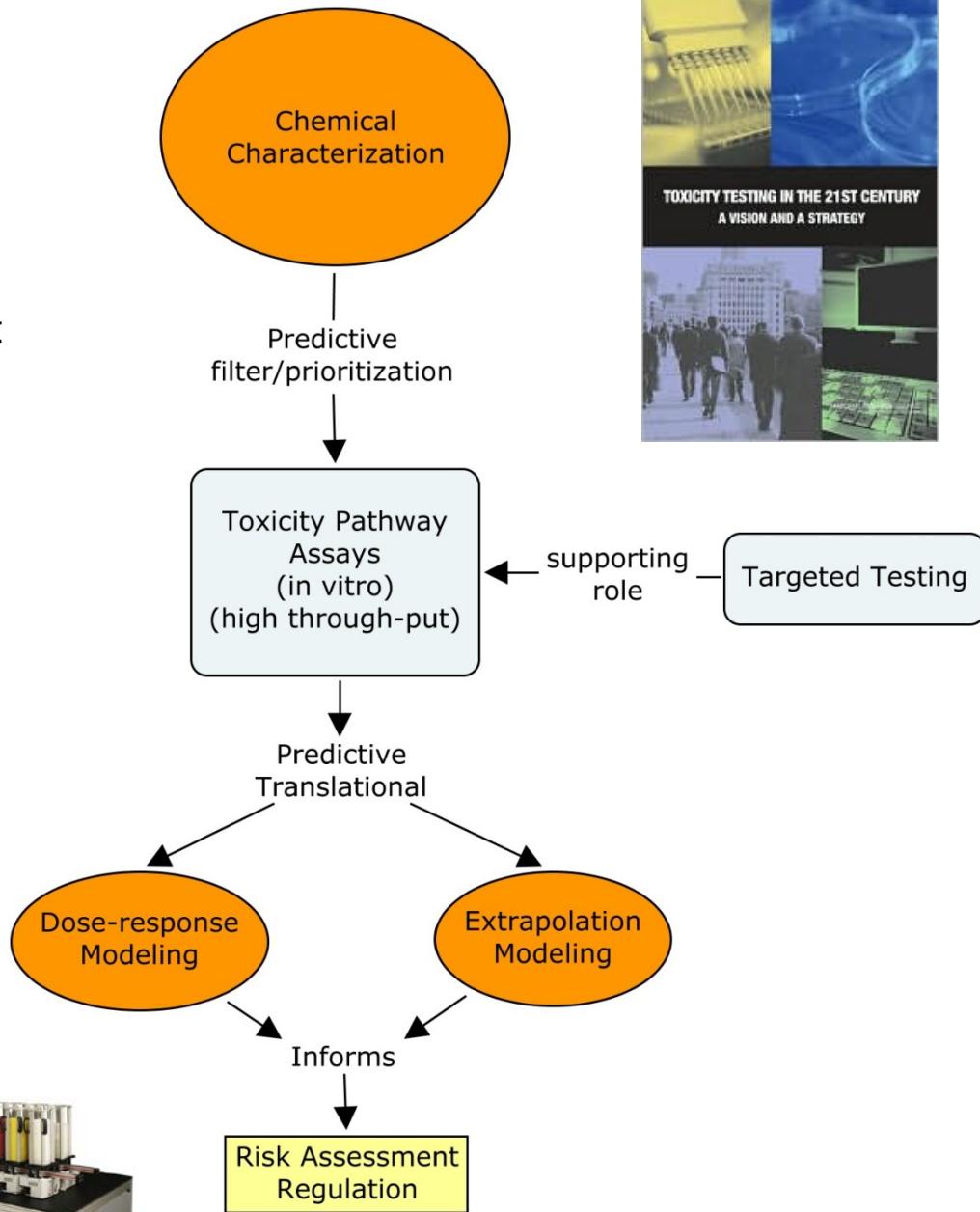
Components of the Vision



“Transform toxicity testing from a system based on whole-animal testing to one founded primarily on in vitro methods that evaluate changes in biologic processes using cells, cell lines, or cellular components, **preferably of human origin**”

“The vision emphasizes the development of suites of predictive, high-throughput assays” (p. 7)

“The mix of tests in the vision include tests that assess critical mechanistic endpoints involved in the induction of overt toxic effects rather than the effects themselves.” (p. 121)



Toxicity Testing in the 21st Century

How does the vision apply to Ecotoxicology?

- Ecotox is faced with the same competing objectives
- Need to generate useful hazard data cheaper, faster
- Need to prioritize what whole animal testing should be done

But....

- Human cell lines will not address species extrapolation
- Less willingness to apply precautionary approach – “show me the adversity”
- Units of concern are populations and ecosystem services/functions, not individuals.



New in this Issue:

ET&C FOCUS

In honor of ET&C's 30th anniversary, we are pleased to introduce the first in a regular series of succinct to timely articles to sharpen our understanding of current and emerging topics of interest to the scientific community at large.

Vision & Strategy:
Predictive Ecotoxicology in the 21st Century

Daniel L. Villeneuve* and Natalia Garcia-Reyero

Meeting the Scientific Needs of Ecological RISK ASSESSMENT in a Regulatory Context

STEVEN P. BRADBURY
U.S. EPA

TOM C. J. FEITEL
PROCTER & GAMBLE SERVICES COMPANY NV/SA (BELGIUM)

CORNELIS J. VAN LEEUWEN
EUROPEAN COMMISSION

Three strategies could move both science and regulation forward.

During the past decade, the field of ecological risk assessment has progressed considerably. Advances have come from such international bodies as

Increasing efficiency, cost-effectiveness, and focus. Risk assessment is a tiered process distinguished by levels of increasing complexity, beginning with the preliminary

Chemical Safety for Sustainability (CSS)

Task 2.1.1: AOP Discovery and Definition

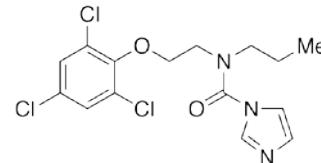
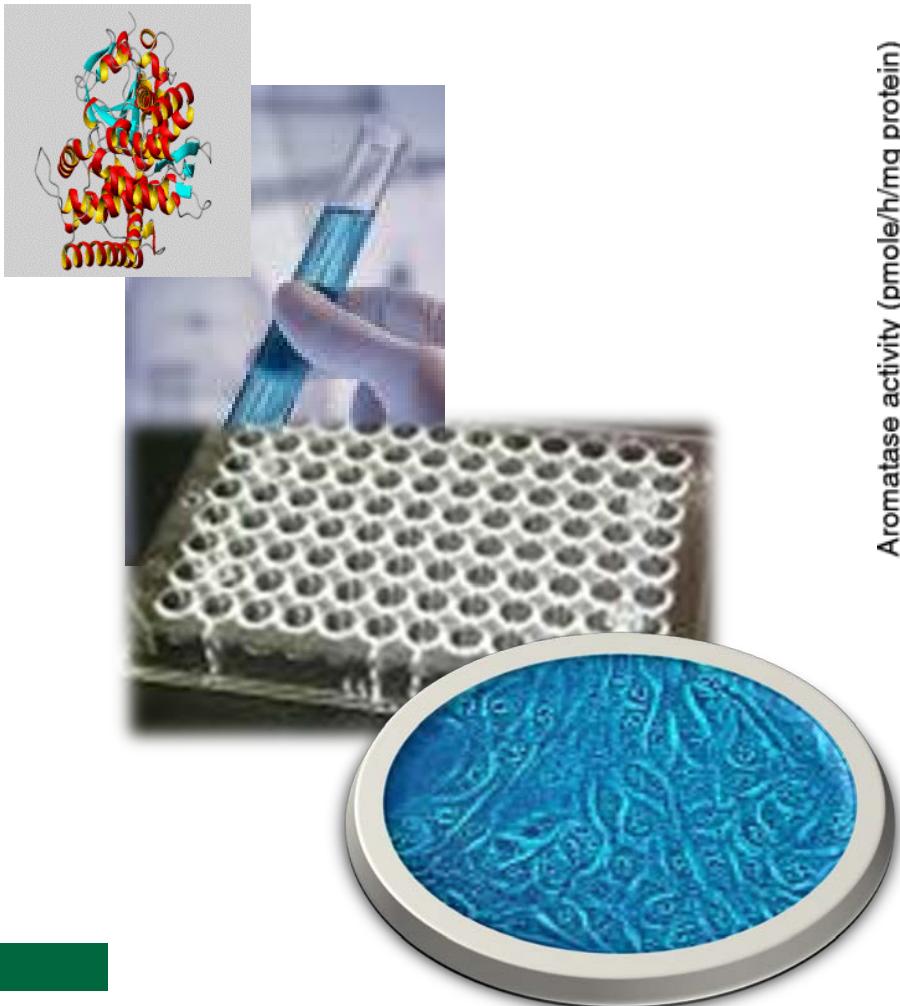
1. Developing AOP knowledge and populating an AOP knowledge-base.
2. Tools to evaluate conservation of molecular targets (i.e., molecular initiating events) as a basis for defining taxonomic susceptibility domains.
3. Supporting HTS assay development and application.
4. Supporting development of “virtual tissue” models.

Can we use in vitro data to predict eco hazards?

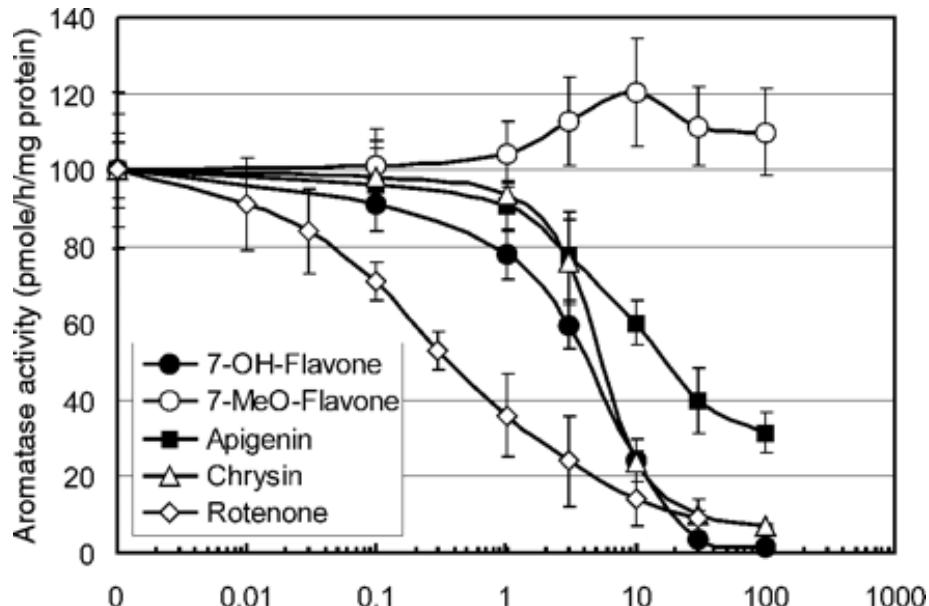


Example

Aromatase (cyp19) inhibition



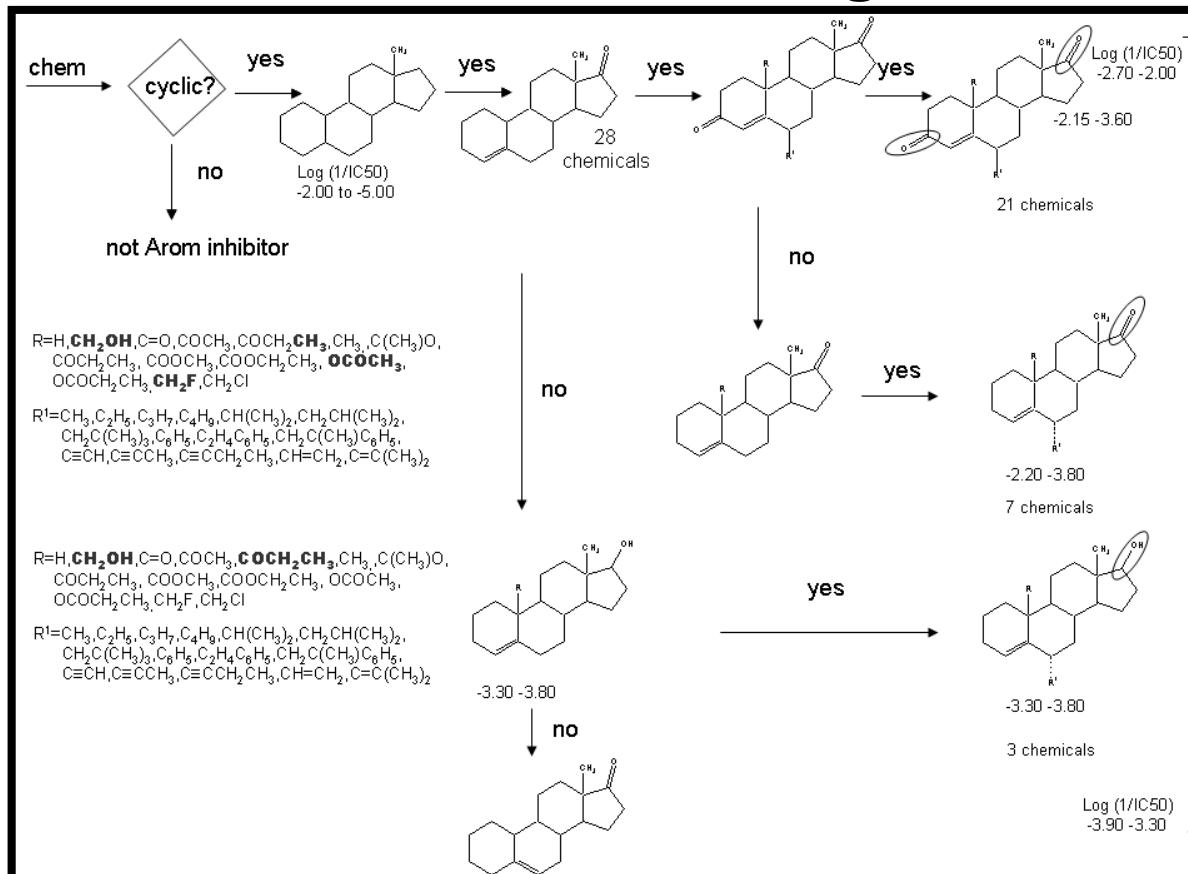
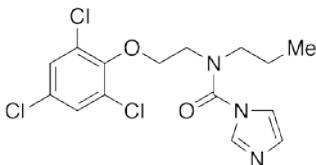
High throughput assay with human cell line



Sanderson et al. 2004, Toxicol. Sci. 82: 70-79

Aromatase Inhibition

QSAR, chemical categories

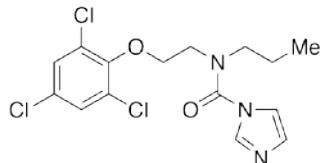


Mechanism-based categorization of aromatase inhibitors: a potential discovery and screening tool

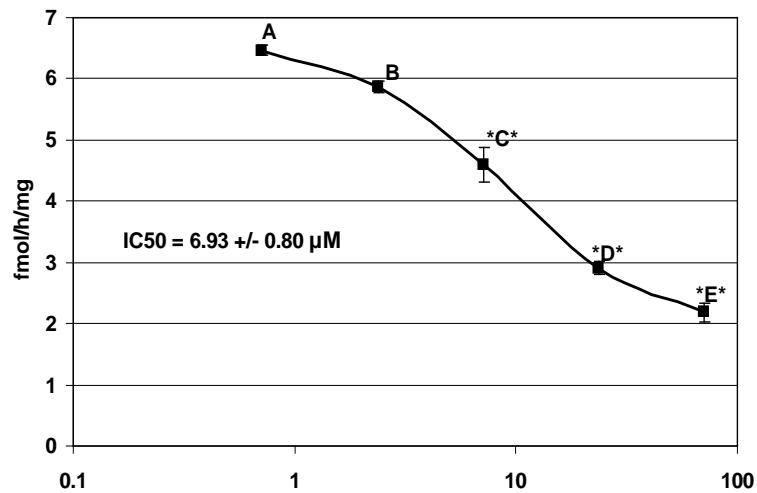
P. I. Petkov ^a; S. Temelkov ^a; D. L. Villeneuve ^b; G. T. Ankley ^b; O. G. Mekenyan ^a

^a Laboratory of Mathematical Chemistry, Bourgas As. Zlatarov University, Bourgas, Bulgaria ^b US Environmental Protection Agency, Mid-Continent Ecology Division, Minnesota, USA

SAR and QSAR in Environmental Research, 20: 657-678.



IC₅₀ = 7.0 ± 0.8 μM



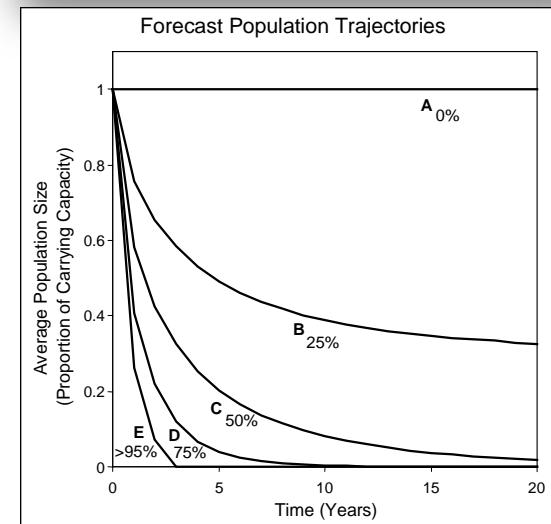
Villeneuve et al. 2006. Aquat. Toxicol. 76:353-368





OPP-EFED
OPPT

Population Impacts



- QSARs
- Read across
- High throughput screening
 - In vitro
- Biomarkers
- Genomics

Use in hazard/risk assessment has been limited by lack of well defined (predictive) linkages between these alternative types of data and adverse outcomes traditionally considered relevant to risk assessment.

Molecular

Cellular

Tissue

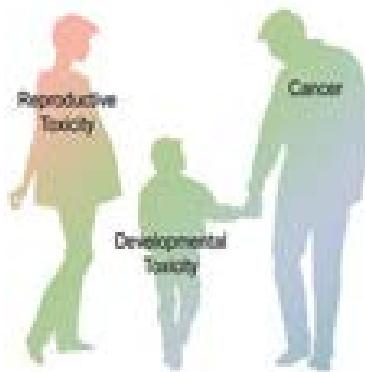
Organ

Organ System

Individual

Population

Ecosystem



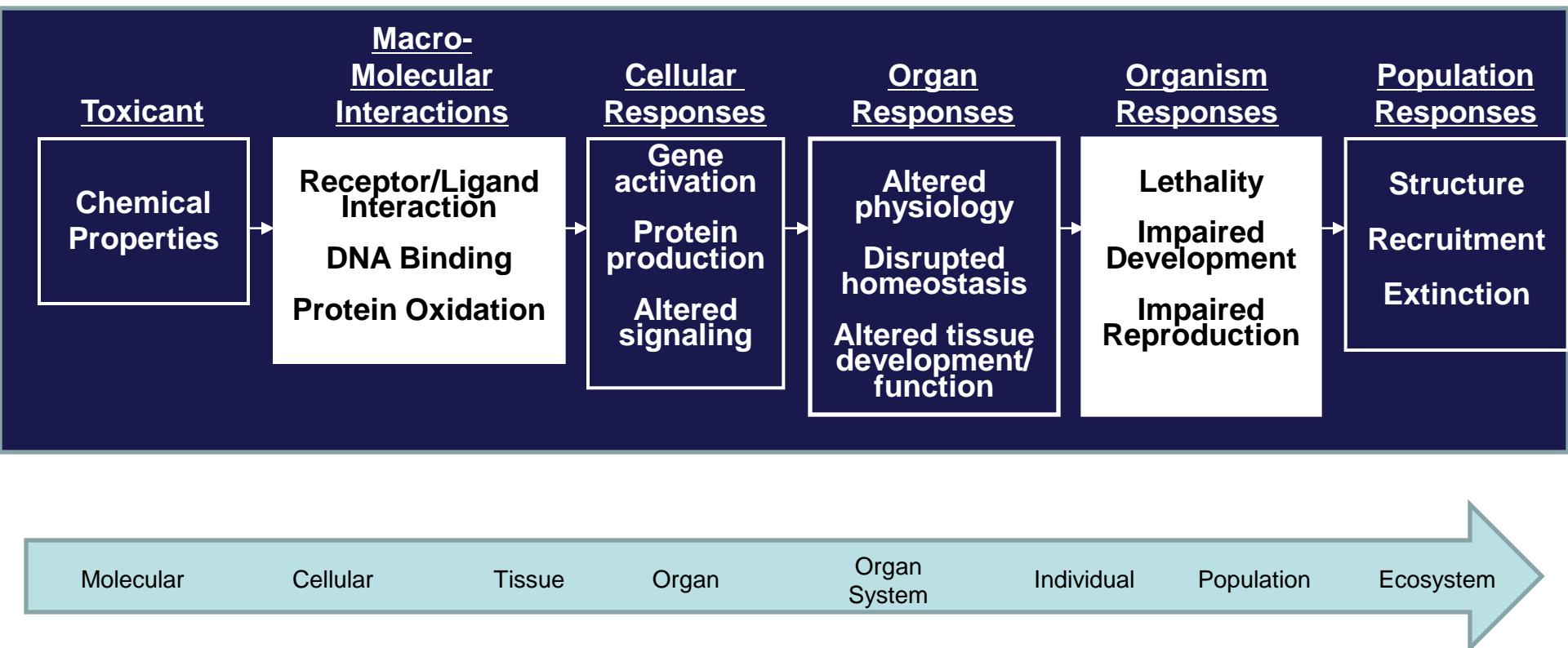
Human Health



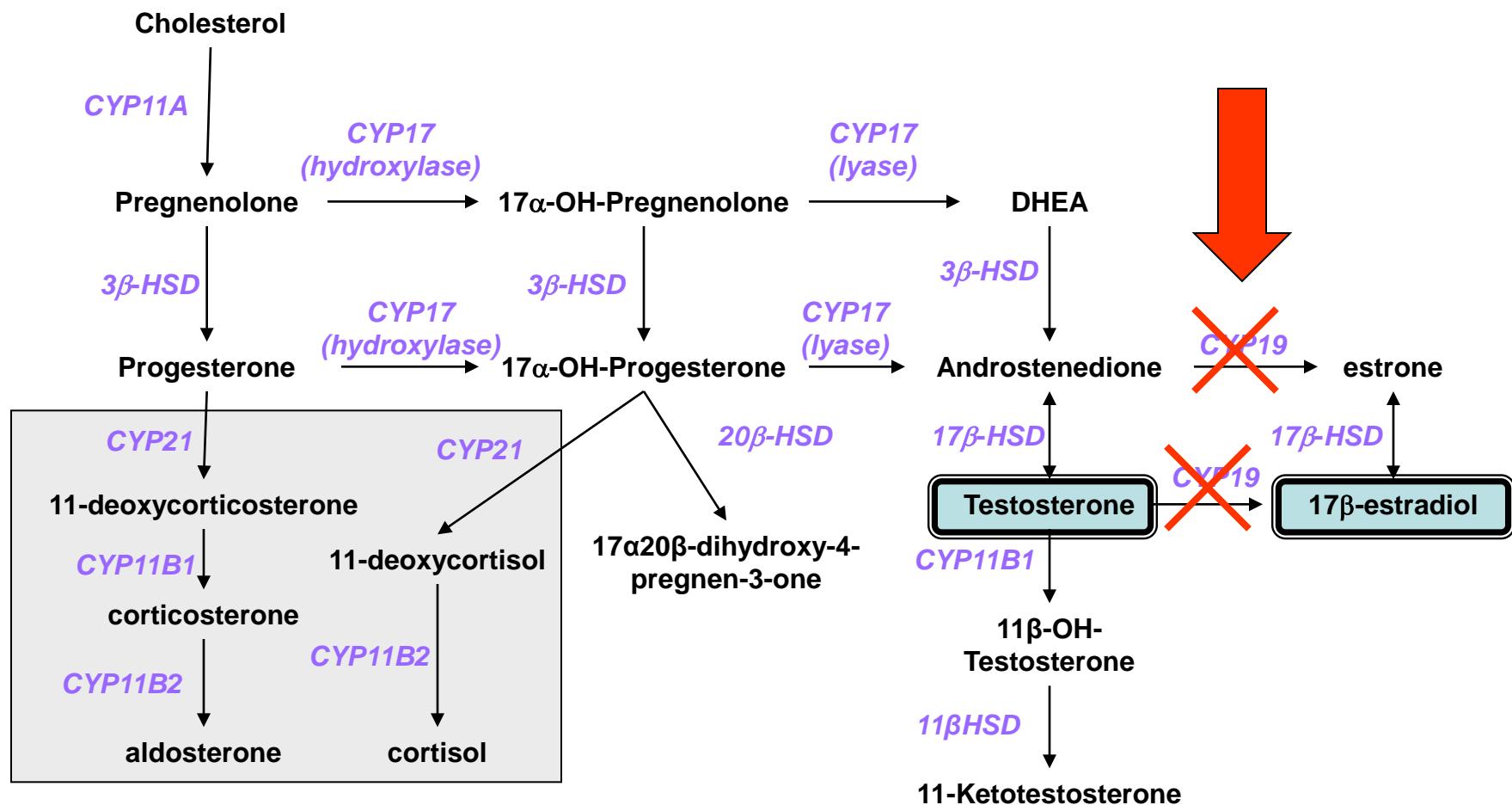
Ecosystem
Health

An Adverse Outcome Pathway (AOP) is a conceptual framework that portrays existing knowledge concerning the linkage between a direct molecular initiating event and an adverse outcome, at a level of biological organization relevant to risk assessment.

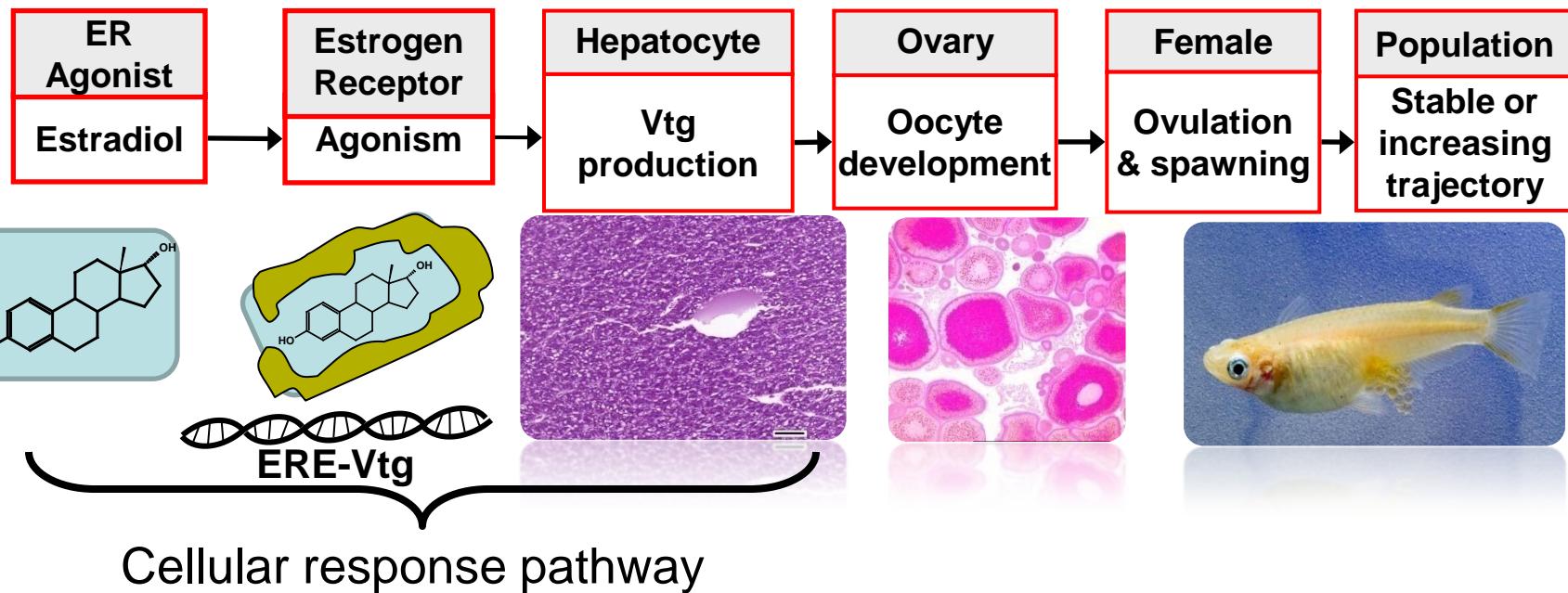
(Ankley et al. 2010, Environ. Toxicol. Chem., 29(3): 730-741.)



Aromatase Inhibition



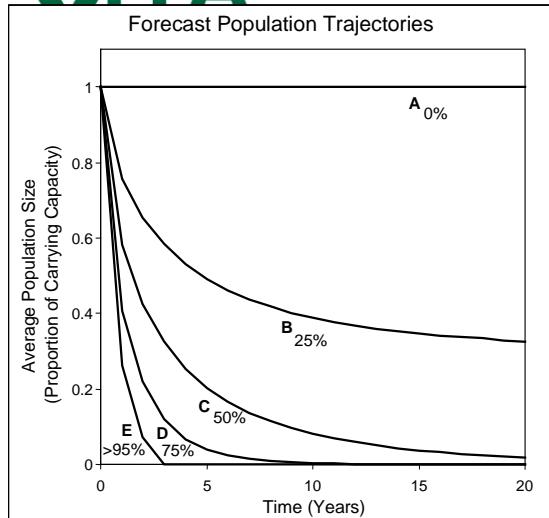
A bit of fish reproductive biology



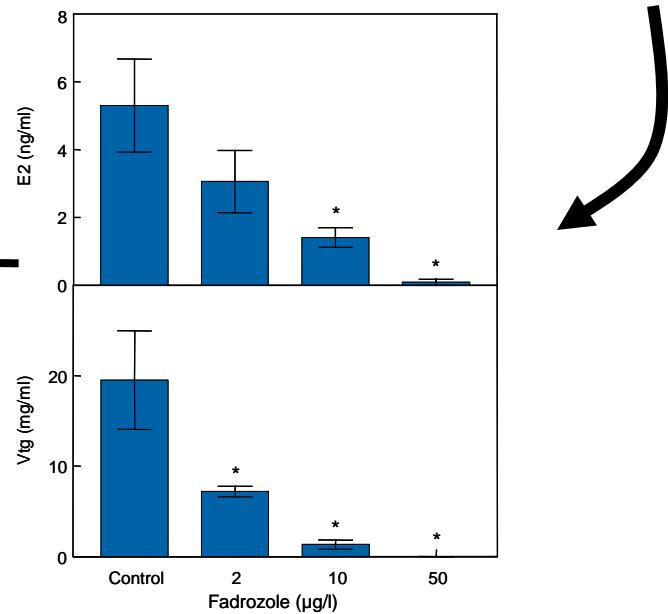
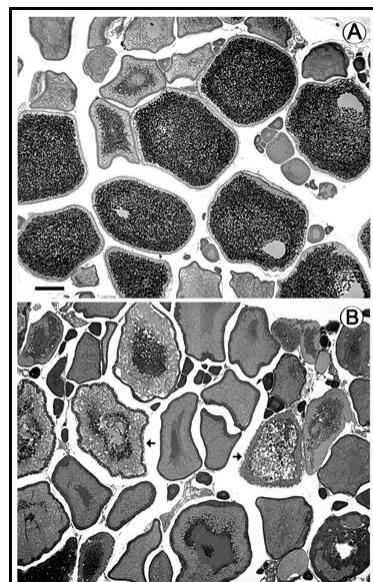
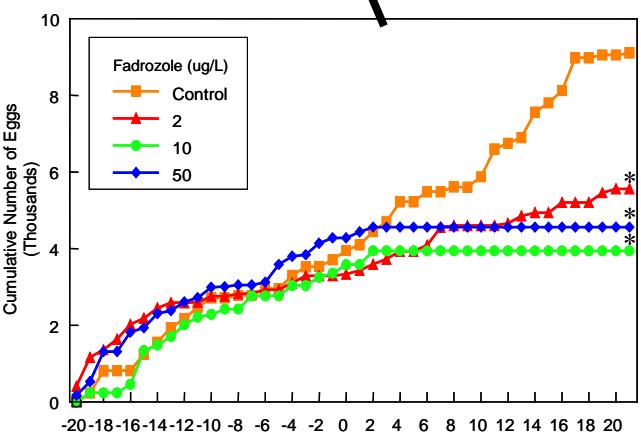
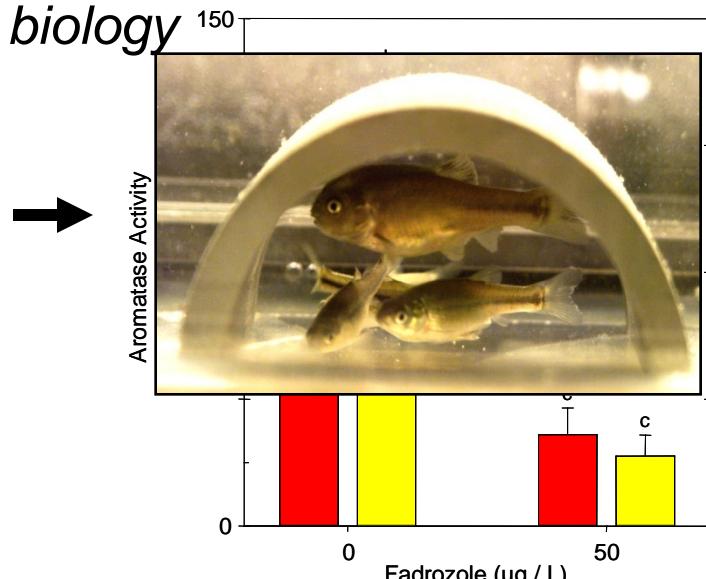
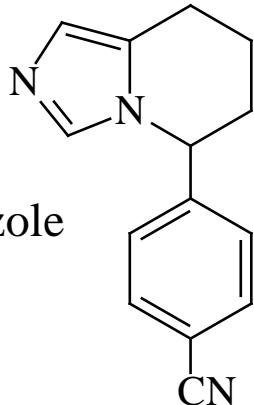
Biologic inputs

"Normal"
Biological Function

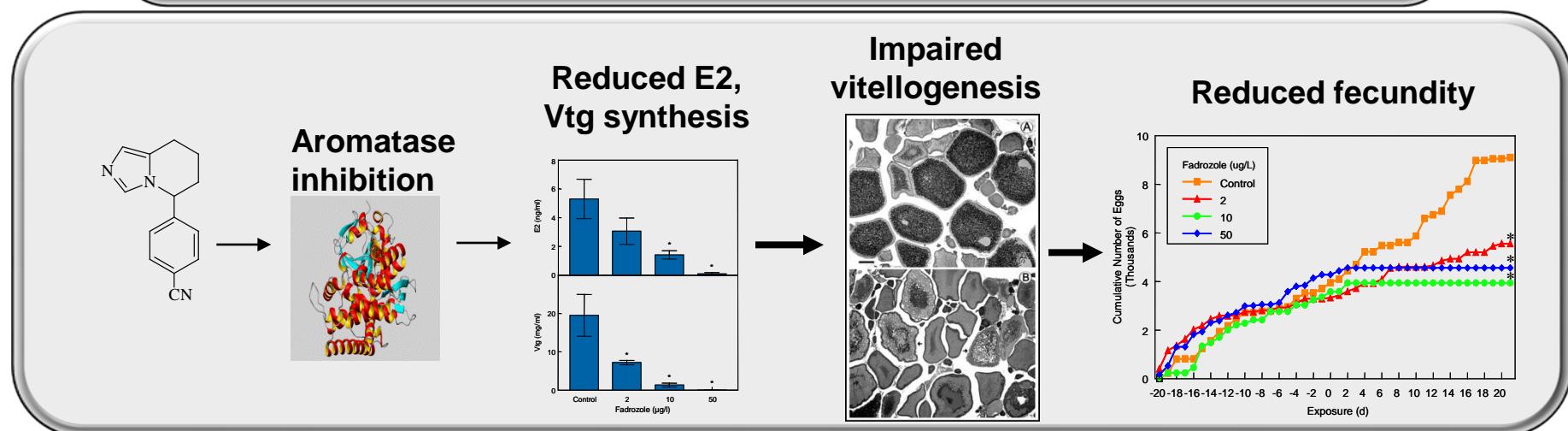
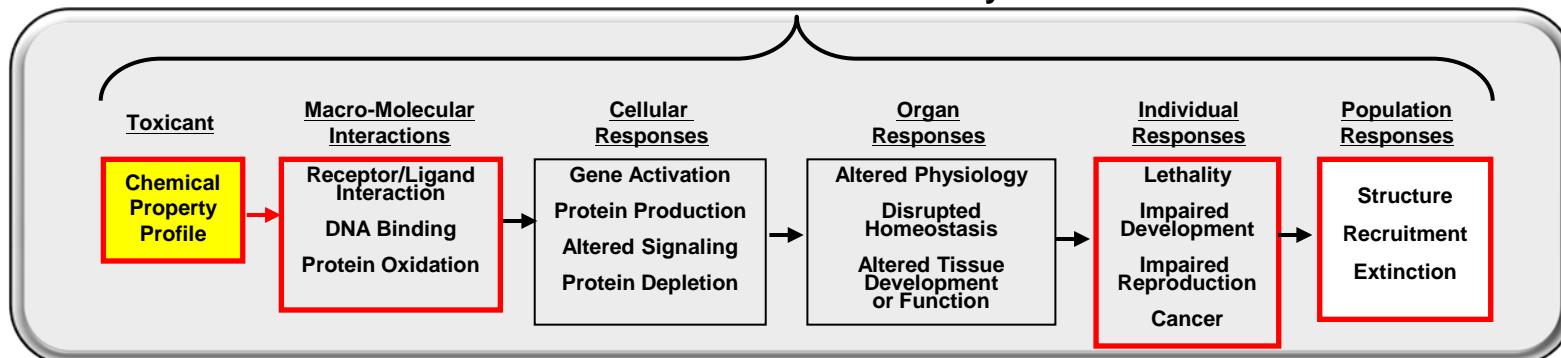
Adverse Outcome Pathways – definition and example



Disturbed fish reproductive biology



Adverse Outcome Pathway

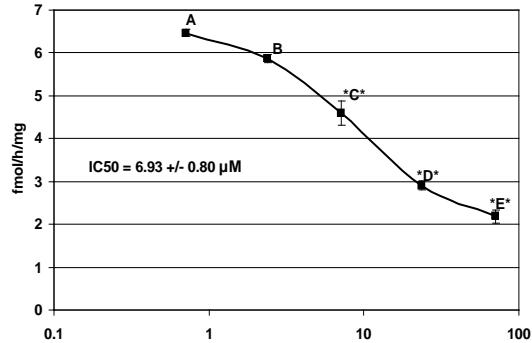


Molecular initiating event

Key events or predictive relationships spanning levels of biological organization

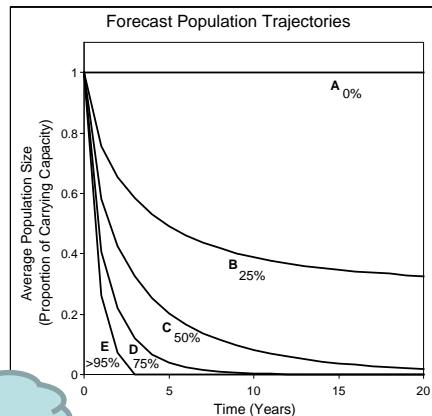
Adverse outcome relevant to risk assessment

$IC_{50} = 7.0 \pm 0.8 \mu M$



AOP

Population Impacts



Aha!



Hazard identification

Prioritization of testing

Risk Assessment

- AOPs assume sufficient perturbation to cause the adverse outcome.



- Risk assessment – probability of an adverse outcome under defined circumstances.
- Requires an understanding of adaptive/homeostatic capacity of biological systems and their limits, relative to concentration and duration of exposure.

Systems Biology Approach

Integrated Genomic and Proteomic Analyses of a Systematically Perturbed Metabolic Network

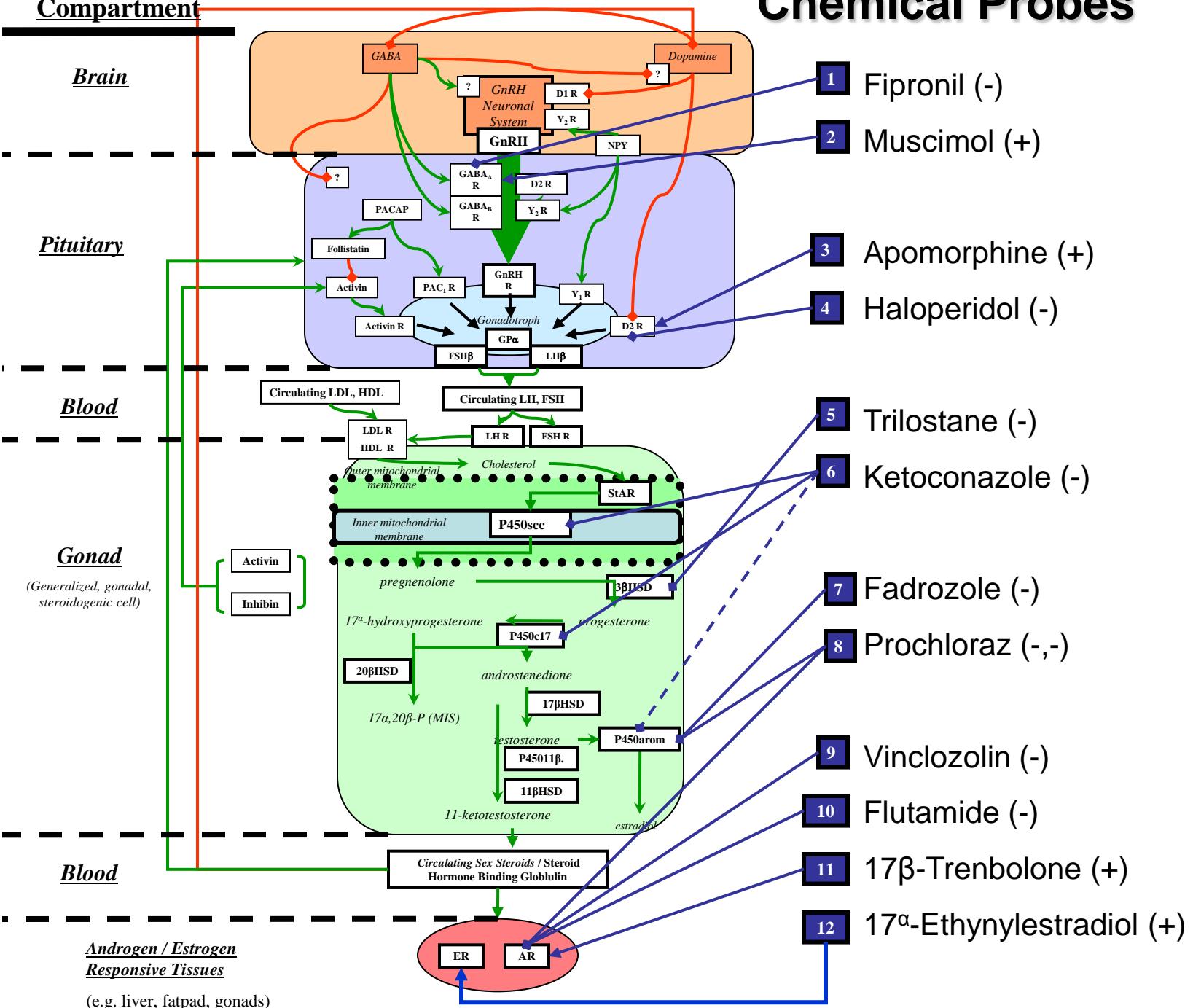
Trey Ideker,^{1,2*} Vesteinn Thorsson,^{1,2} Jeffrey A. Ranish,^{1,2}
Rowan Christmas,¹ Jeremy Buhler,³ Jimmy K. Eng,¹
Roger Bumgarner,⁴ David R. Goodlett,¹ Ruedi Aebersold,^{1,2}
Leroy Hood^{1,2}

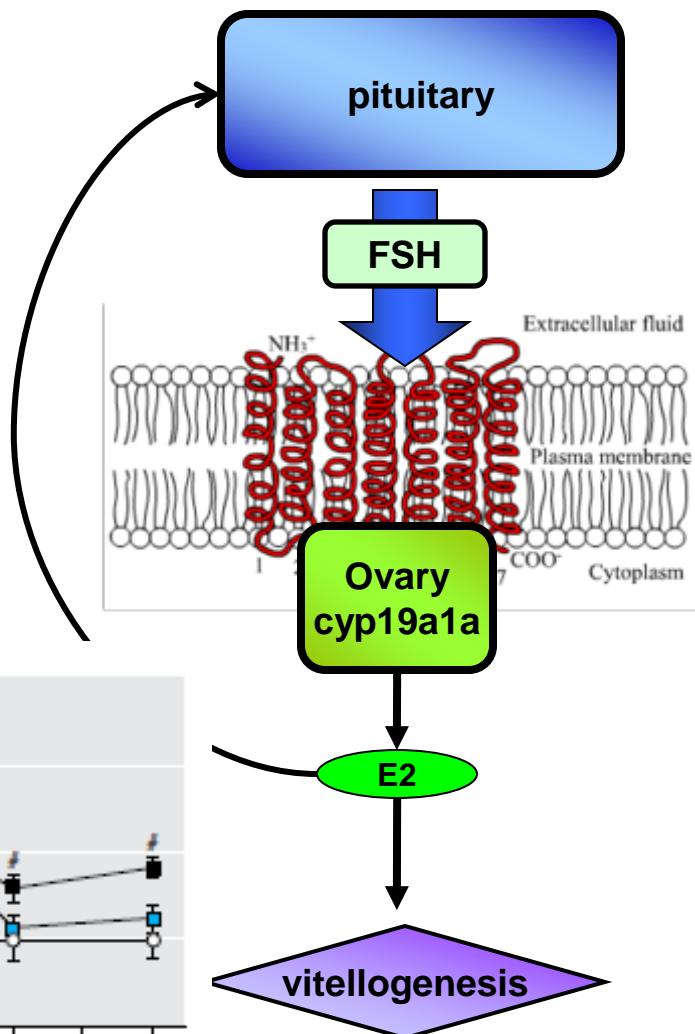
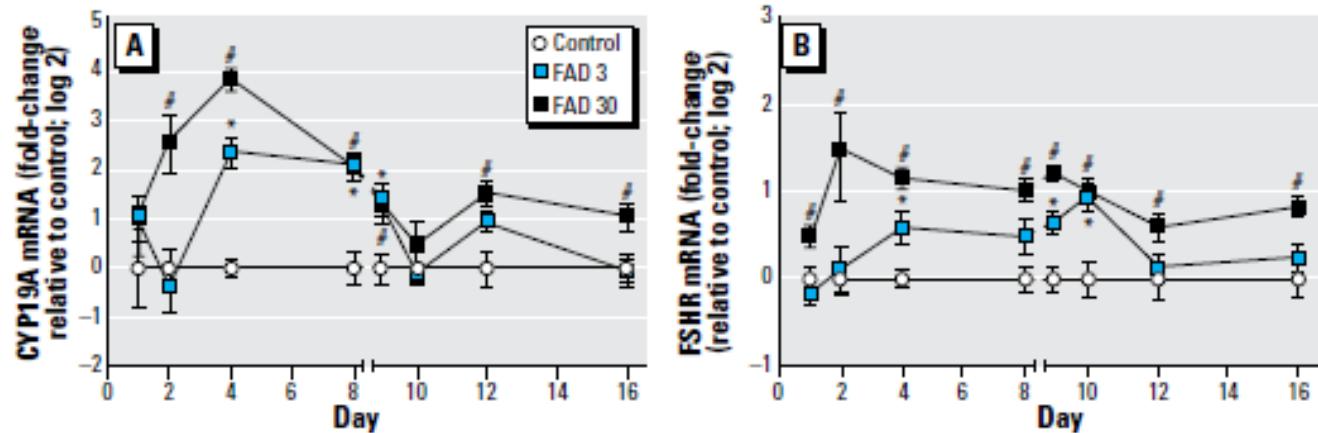
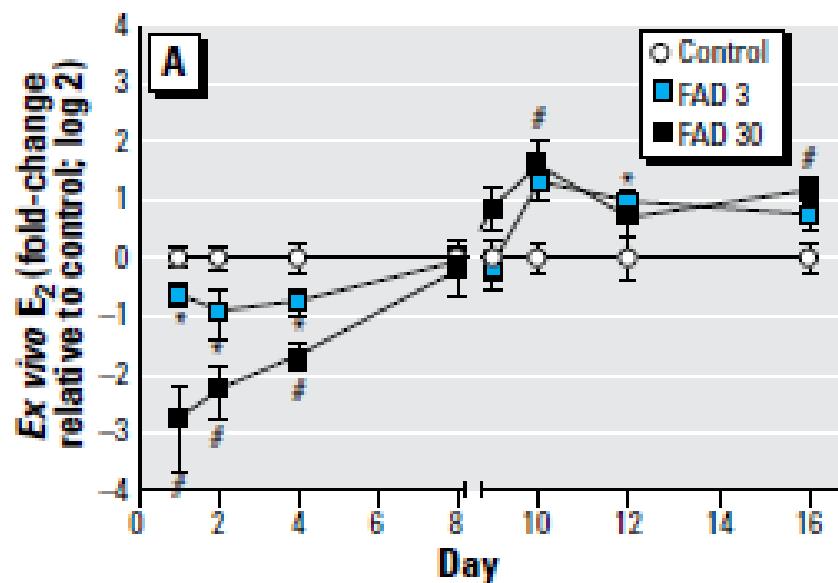
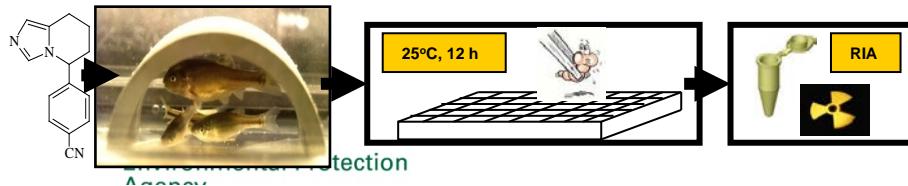
www.sciencemag.org SCIENCE VOL 292 4 MAY 2001

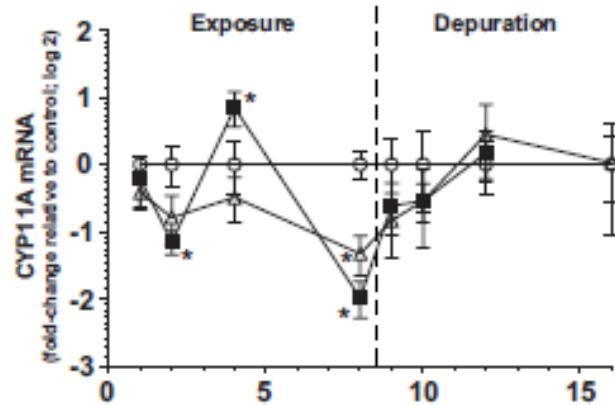
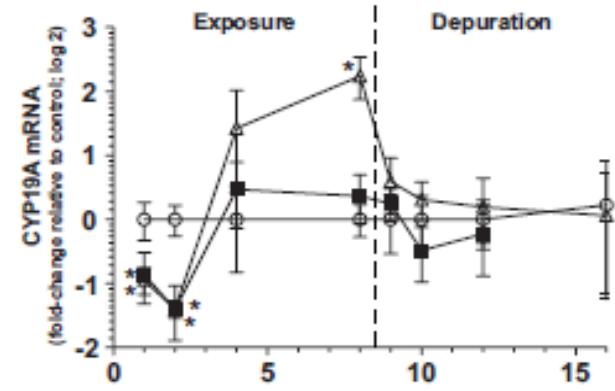
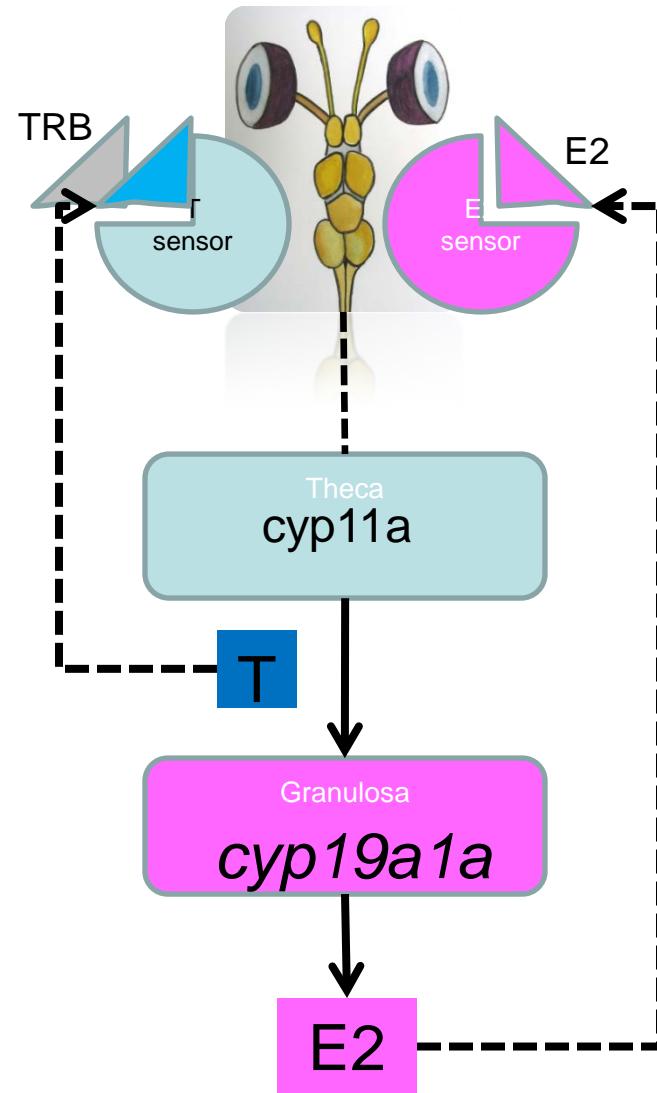
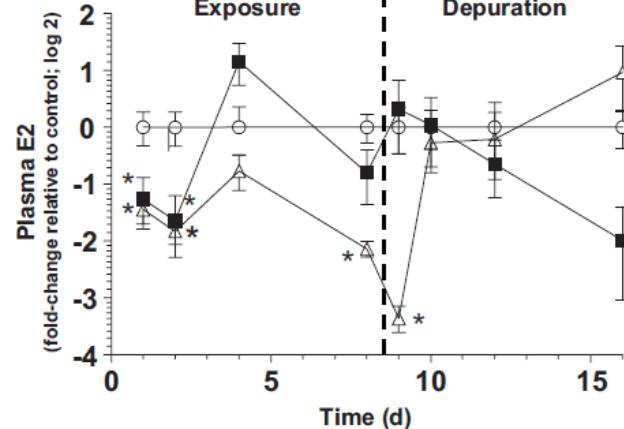
- (i) Define the subset of genes, proteins, and other small molecules constituting the pathway of interest.
- (ii) Perturb each pathway component
Detect and quantify the corresponding global cellular response to each perturbation
- (iii) Integrate the observed responses with the current, pathway specific model
- (iv) Formulate new hypotheses to explain observations not predicted by the model.

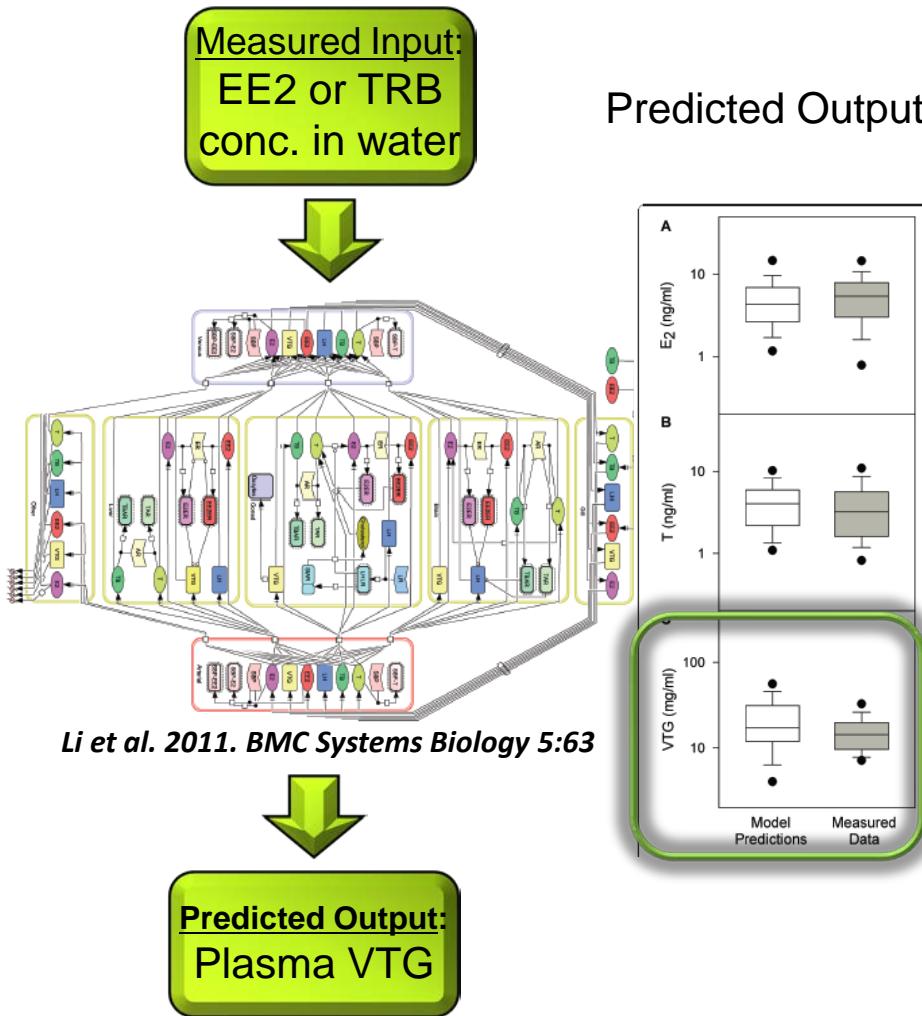
Design additional perturbation experiments to test these, and iteratively repeat steps (ii), (iii), and (iv).

Chemical Probes



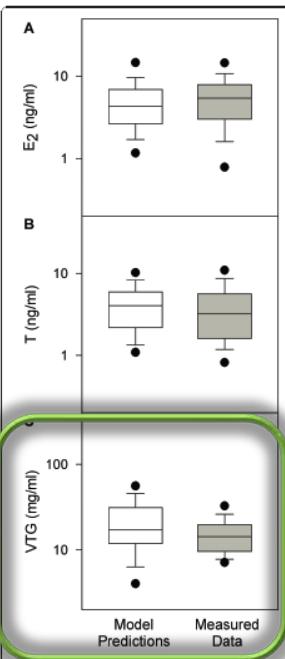


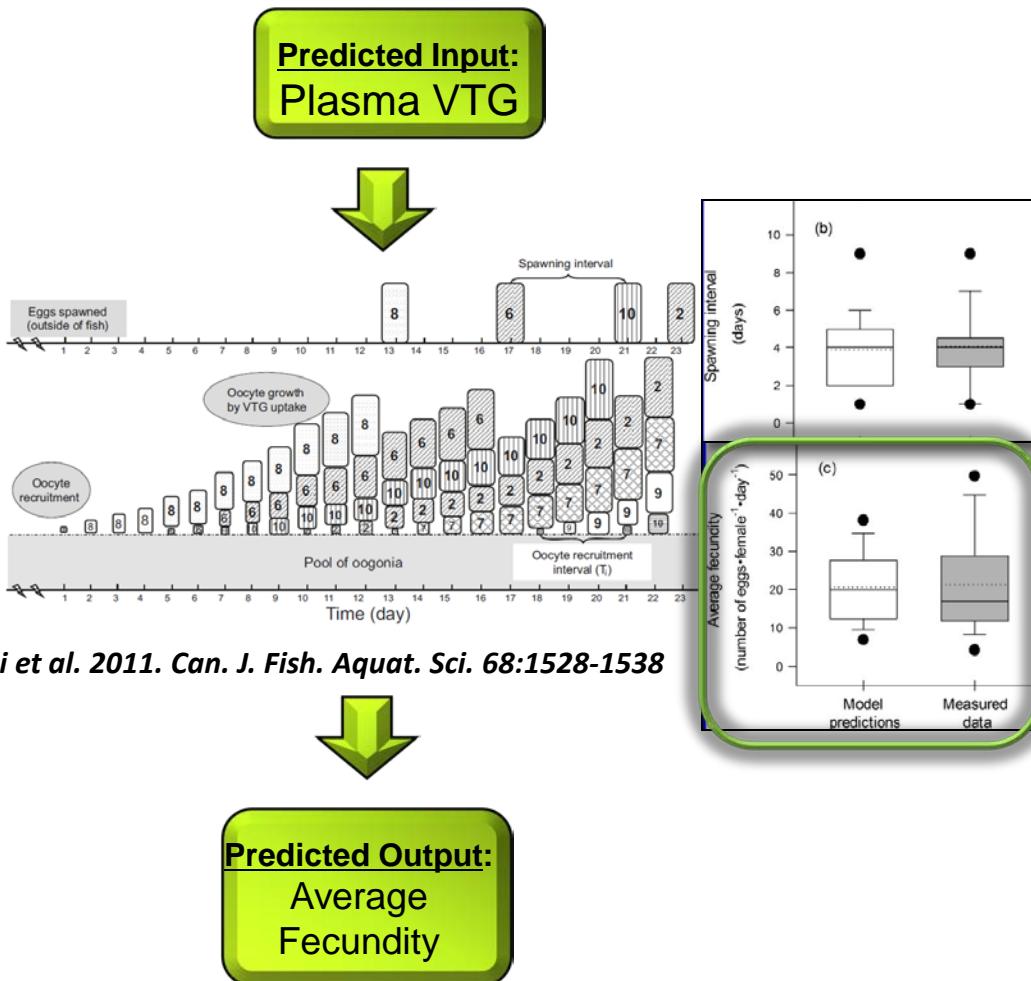
A**B****A**



Mechanistic Modeling

- To aid extrapolation from molecular perturbations to outcomes.
- How much perturbation is too much?





Mechanistic Modeling

- To aid extrapolation from molecular perturbations to outcomes.
- How much perturbation is too much?
- Predictive toxicology

Predicted Input:
 Average
 Fecundity



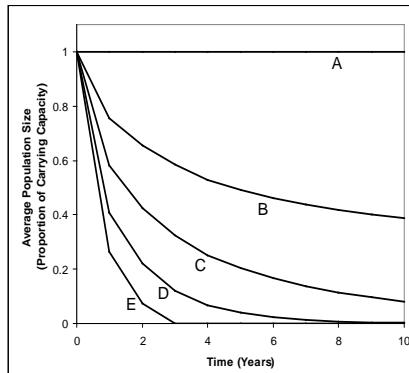
Life table for the fathead minnow survival rates derived from populations (Westman, 1938; Zischke et al., 1983; Duda, 1989; Gleason and Nacci, 2001)

Age (years)	Survival (per year)	Fecundity (eggs/year)
0	1	0
1	0.001	750
2	0.00039	1500
3	0.0001521	3000
4	0	N/A

$$M_{\text{fathead minnow}} = \begin{bmatrix} 0.75 & 1.5 & 3 \\ 0.39 & 0 & 0 \\ 0 & 0.39 & 0 \end{bmatrix}$$

Fig. 1. Leslie matrix for the fathead minnow derived from field studies (Westman, 1938; Zischke et al., 1983; Duda, 1989; Gleason and Nacci, 2001) and developed using birth pulse survival and fertility rates and a prebreeding census (Gotelli, 1998; Caswell, 2001).

Miller and Ankley, 2004. Ecotoxicol. Environ. Saf. 59:1-9



Mechanistic Modeling

- To aid extrapolation from molecular perturbations to outcomes.
- How much perturbation is too much?
- Predictive toxicology

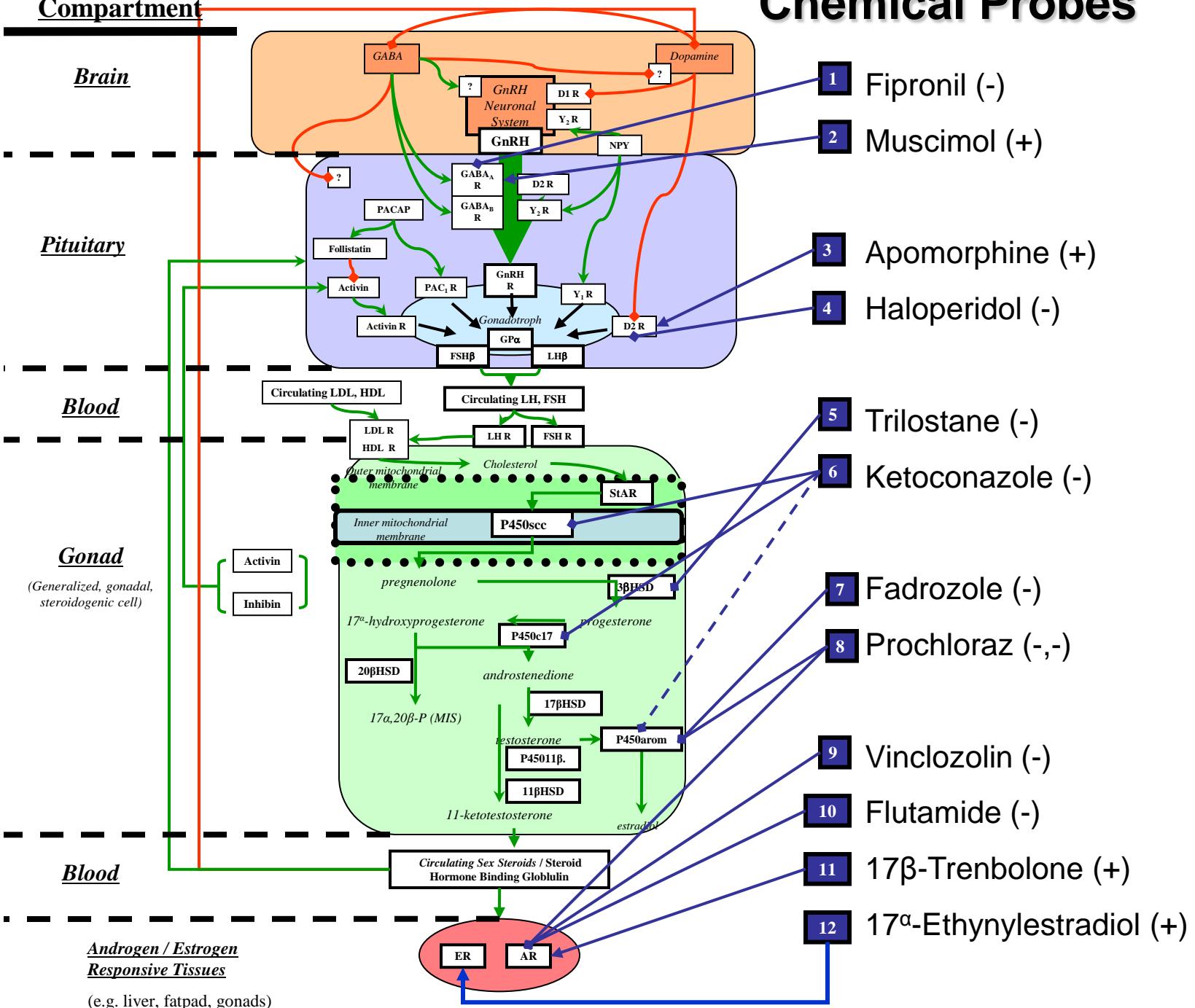


Predicted Output:
 Population
 forecast

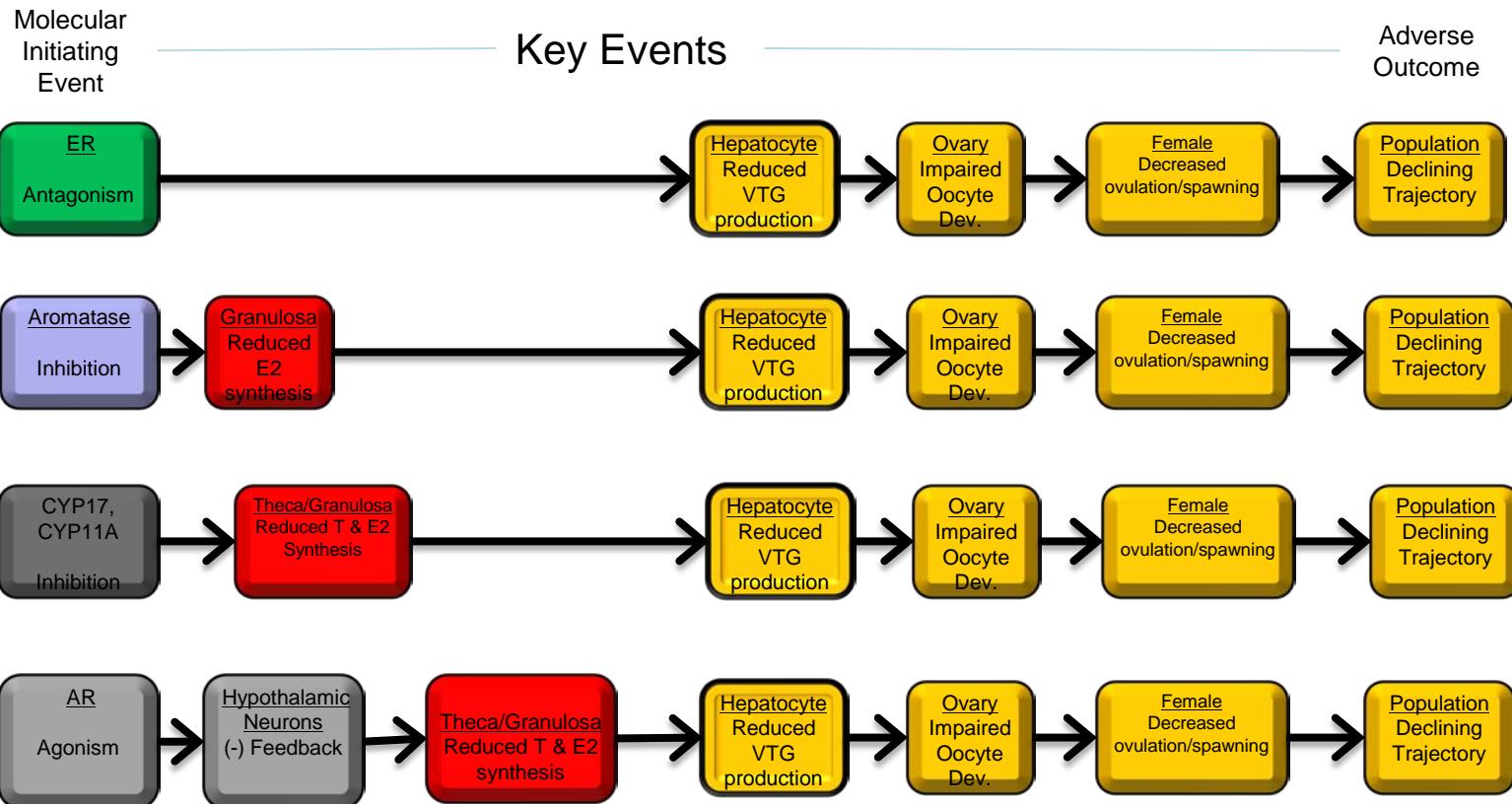
Scaling it up: Can we use human-oriented HTS data to predict eco hazards?



Chemical Probes



Assemble AOPs into AOP Knowledge-base





Biochemical Assays **ToxCast™ Assays** Cellular Assays

(Primarily Human / Rat)

- Protein families

- GPCR
- NR
- Kinase
- Phosphatase
- Protease
- Other enzyme
- Ion channel
- Transporter

- Assay formats

- Receptor binding
- Enzyme activity
- Co-activator recruitment



**~600 Total
Endpoints**

Model Organisms

- Zebrafish embryo development
- *C. elegans* growth

- Cell lines

- HepG2 human hepatoblastoma
- A549 human lung carcinoma
- HEK 293 human embryonic kidney
- T47D human breast carcinoma
- PC12 rat neuronal

- Primary cells

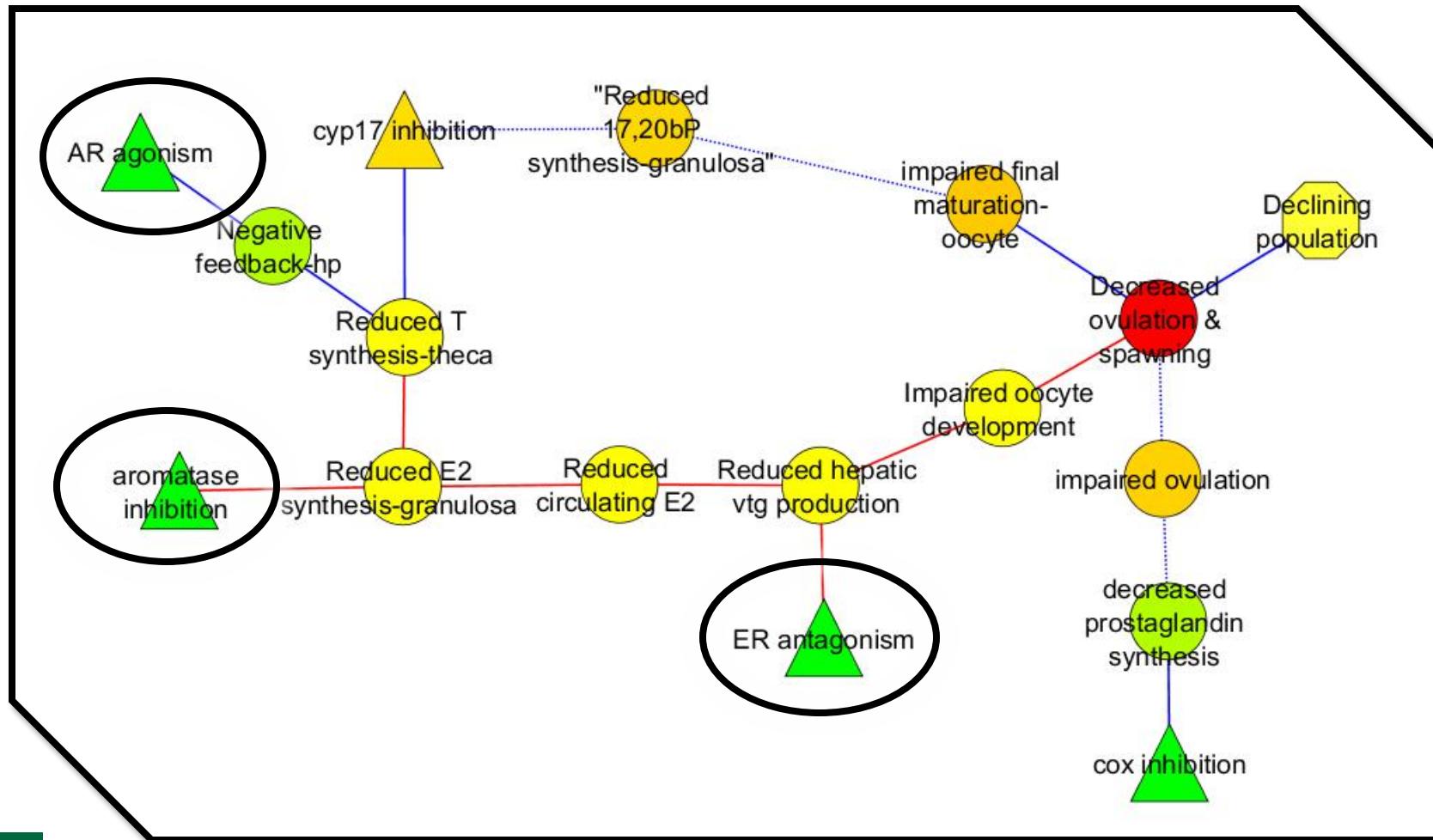
- Human vascular endothelial cells
- Human monocytes
- Human keratinocytes
- Human fibroblasts
- Human proximal tubule kidney cells
- Human small airway epithelial cells
- Rat hepatocytes
- Mouse embryonic stem cells

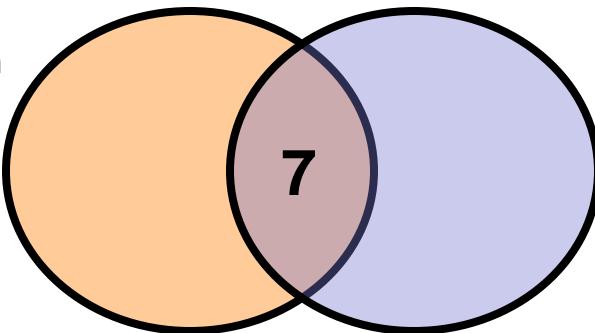
- Assay formats

- Cytotoxicity
- Reporter gene
- Gene expression
- Protein expression
- High-content imaging

AOP Networks

- MIE (\blacktriangle) and Key Events (\bullet) aligned with molecular screening assays





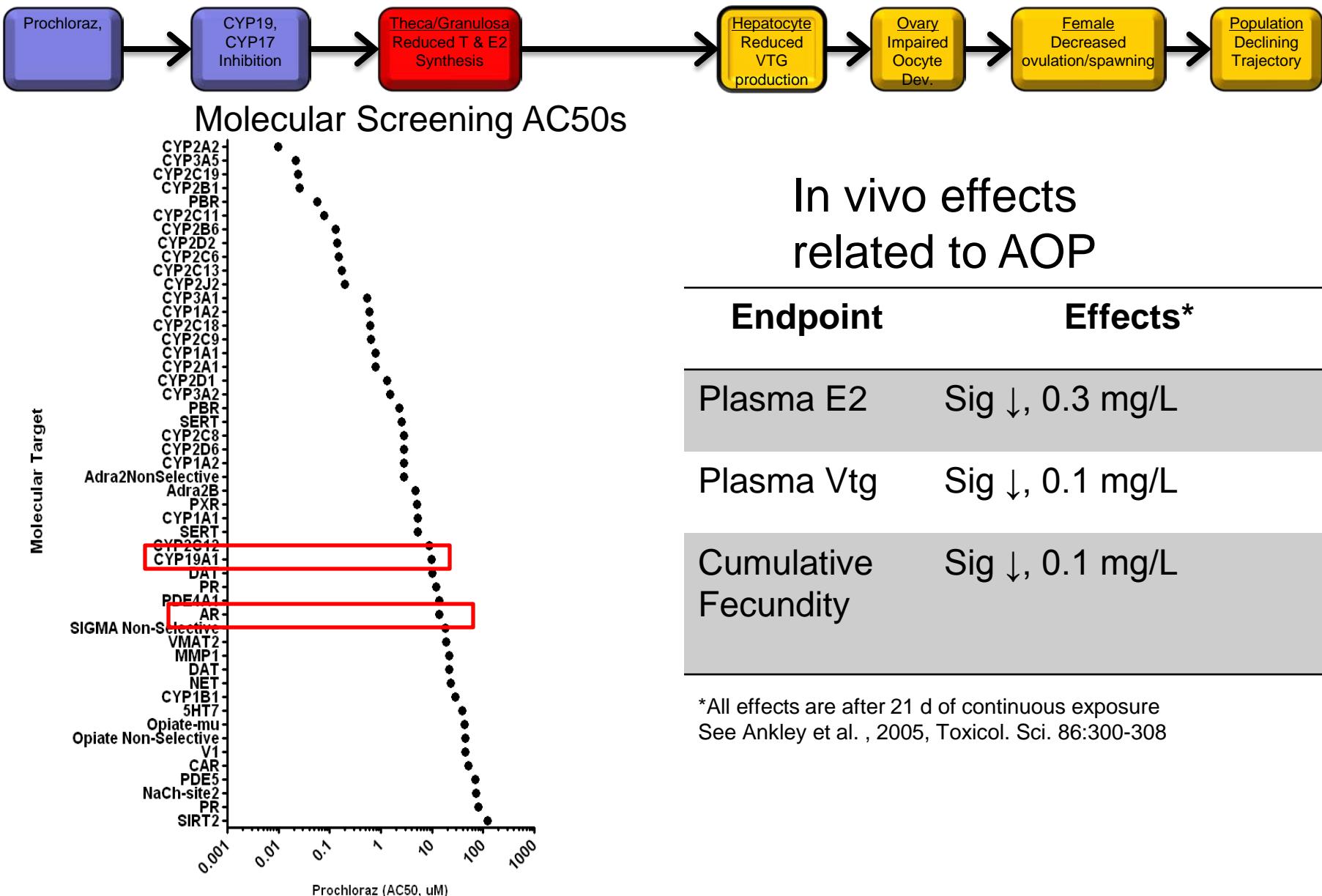
Seven Test chemicals	Putative M.I.E.
Prochloraz	Cyp19, cyp17 inhib.
Propiconazole	Cyp inhib.
Fenarimol	ER antagonist
Bisphenol A	ER agonist, AR ant.
Vinclozolin	AR antagonist
Prometon	Photosystem II inhib.
Fipronil	GABA-A receptor, chloride channel blocker

Post-hoc analysis

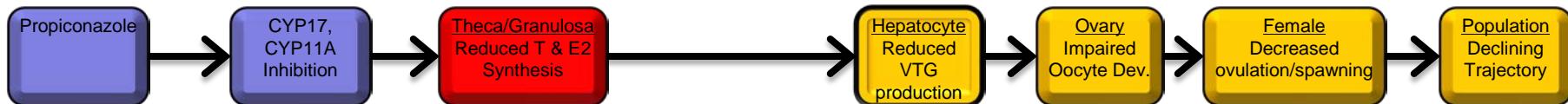
Did MS flag a MIE known to be relevant to fish reproduction?

Were the *in vivo* effects consistent with our AOP(s)?

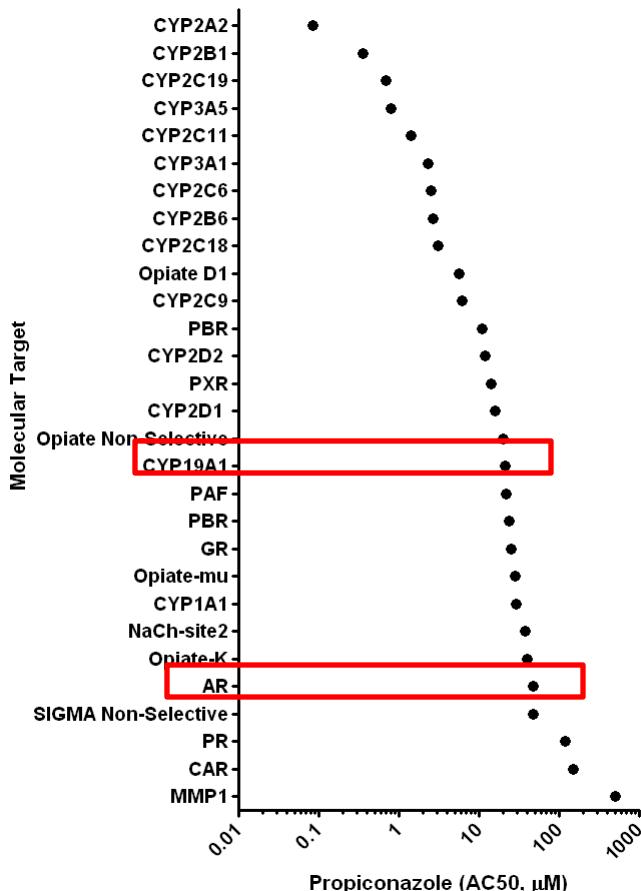
Prochloraz



Propiconazole



Molecular Screening AC50s



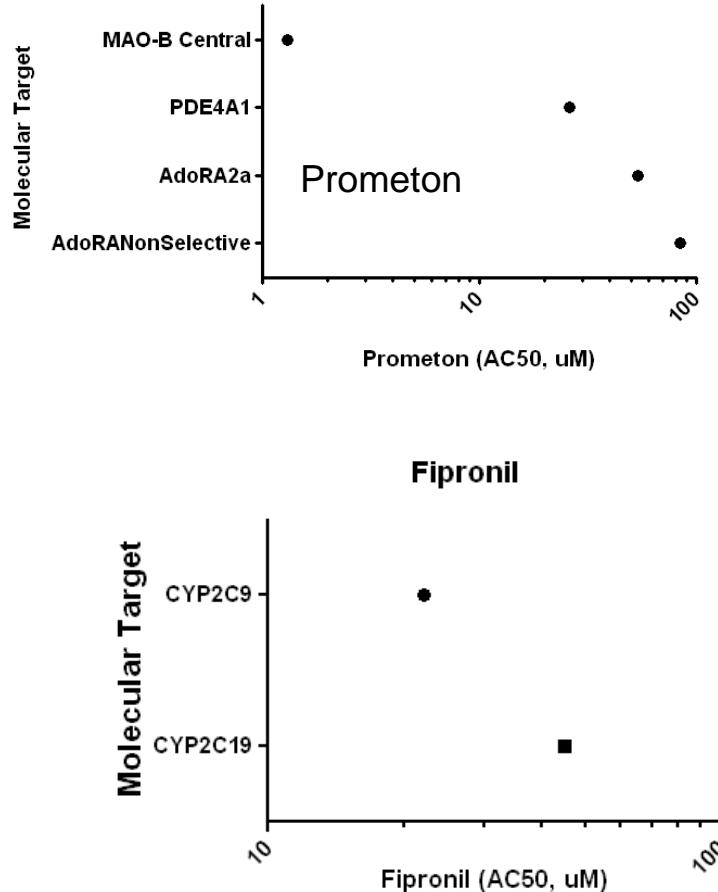
In vivo effects related to AOP

Endpoint	Effect
Plasma E2	Sig ↓, 0.5 mg/L
Plasma Vtg	Sig ↓, 0.5 mg/L*
Cumulative Fecundity	Sig ↓, 0.5 mg/L

* From 96 h range finding study, all other data are from a 21 d exposure. Vtg data for 21 d exposure not yet available.
From Skolness et al. (in press, Toxicol. Sci. 2012)

Prometon, Fipronil

Molecular Screening AC50s



In vivo

Endpoint effects Effects*

Plasma E2	No effect (up to 1 mg/L)
Plasma Vtg	No effect (up to 1 mg/L)
Cumulative fecundity	No effect (up to 1 mg/L)

*Villeneuve et al. 2006, Environ. Toxicol. Chem. 25: 2143-2153

Endpoint Effects*

Plasma E2	No effect (up to 5 $\mu\text{g}/\text{L}$)
Plasma Vtg	No effect (up to 5 $\mu\text{g}/\text{L}$)
Cumulative fecundity	No effect (up to 5 $\mu\text{g}/\text{L}$)

*Bencic et al., in preparation

Results

- Overall, when MS indicated perturbation of molecular target associated with AOP, adverse effects on reproduction and related key events were observed.
- Predicted pattern of effects observed, despite other targets being affected.
- Two chemicals with least impact on fish repro, also quite inactive in MS assays.
- Testing of *a priori* predictions based on MS and AOPs needed

Prospective Assessments



Data poor
Large Inventory

e.g., TSCA

Screening

- Efficient/cost effective methods to predict hazard.
- Get more information from “alternative” data that are available.

Prioritization

- Conduct only the tests most likely to drive assessment
- More effective use of testing resources increase efficiency.



Data-rich
Programs

e.g., FIFRA

Diagnostic Assessments

Regulatory context



Environmental Monitoring

e.g., Regions,
OW, GLNPO

- **Complex and undefined exposures**
- **If apical responses are observed, it's too late**

Need

- Ability to cast a broad net
- Early warning signs
- Indication/elimination of cause(s) - diagnostic



Early environmental risk assessors.

Diagnostic Assessments

- Chemical monitoring strategies are effective for chemicals whose hazards are well understood and for which sensitive analytical methods are available.
- Biological effects monitoring can be a powerful complement to chemical monitoring
 - Can detect exposure to chemicals for which analytical methods are unavailable or impractical.
 - Can provide insight into the potential biological consequences of those exposures.

Chemical Safety for Sustainability (CSS)

Task 2.1.2: AOP-Based Effects Monitoring and Exposure Reconstruction

1. Apply AOP knowledge to development of effects-based monitoring approaches.
2. Methods for collecting and preparing environmental samples for HTS.
3. AOPs to support identification/elimination of causes of biological responses [exposure reconstruction]



One Common Approach to Effects-Based Monitoring

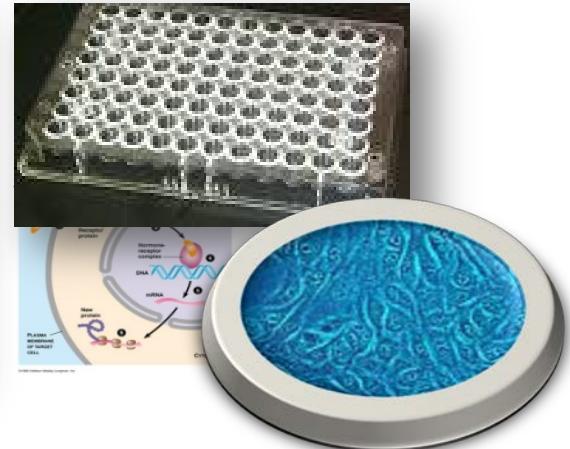
Environmental Sampling



Sample Extraction/Prep



In vitro bioassay



Supervised/Targeted Effects-based Monitoring

Looking under the biological lamp-post

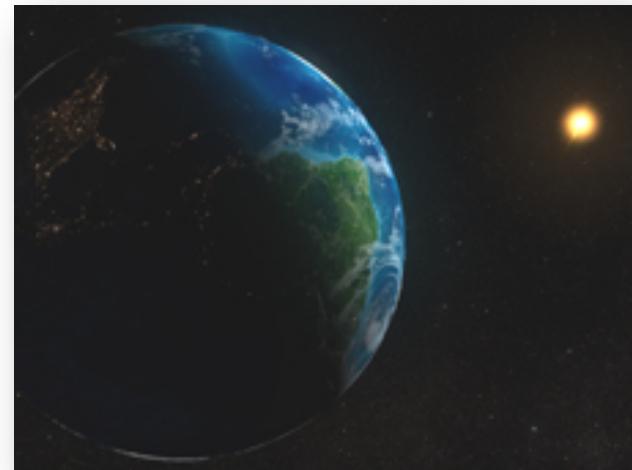
- Will only detect the biological activities we look for
- Effective once a hazard of concern has been identified
- Not ideal for surveillance
- May miss activities that influence in vivo biological/ecological outcomes



Unsupervised Effects-based Monitoring

Ability to detect what we might not expect

- Take advantage of HTP to rapidly/cost effectively screen wide range of biological activities
- Ideal for surveillance
- Identification of targeted assays/endpoints for subsequent monitoring of status and trends
- More complete picture of how mixed biological activities influence *in vivo* biological/ecological outcomes

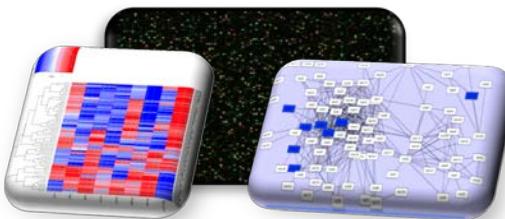


Fish exposed *in situ*

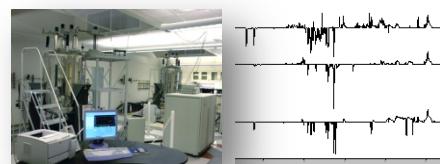


Unsupervised

DNA-microarray Transcriptomics



Metabolomics

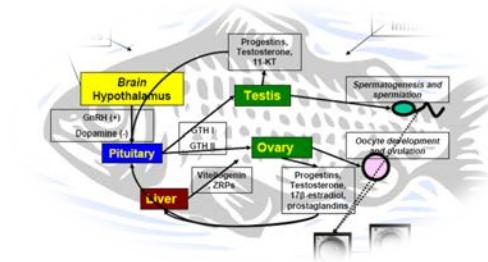


Surface water samples/extracts



Supervised

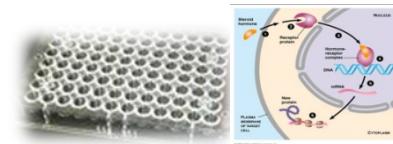
- Endpoints associated with established adverse outcome pathways:



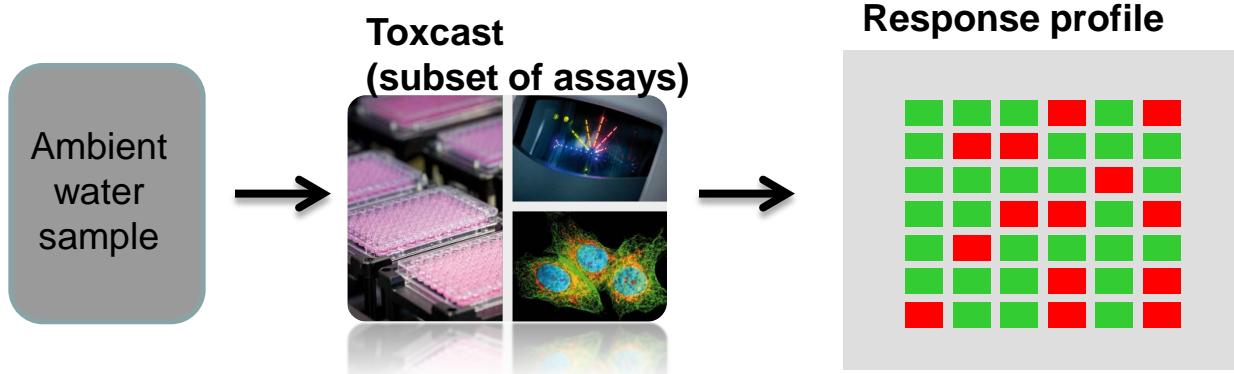
- E.g., Biochemical and molecular markers of endocrine disruption and adverse reproductive outcomes.

In vitro bioassays

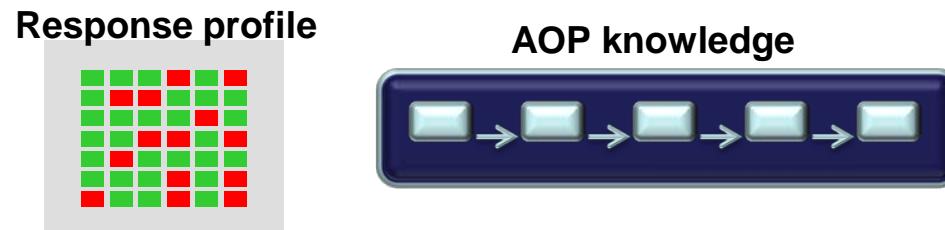
- MDA-kb2: (anti)androgenic activity
- T47D: (anti)estrogenic activity
- H4IIE: dioxin-like contaminants
- H295R: steroidogenesis inhibitors
- Others



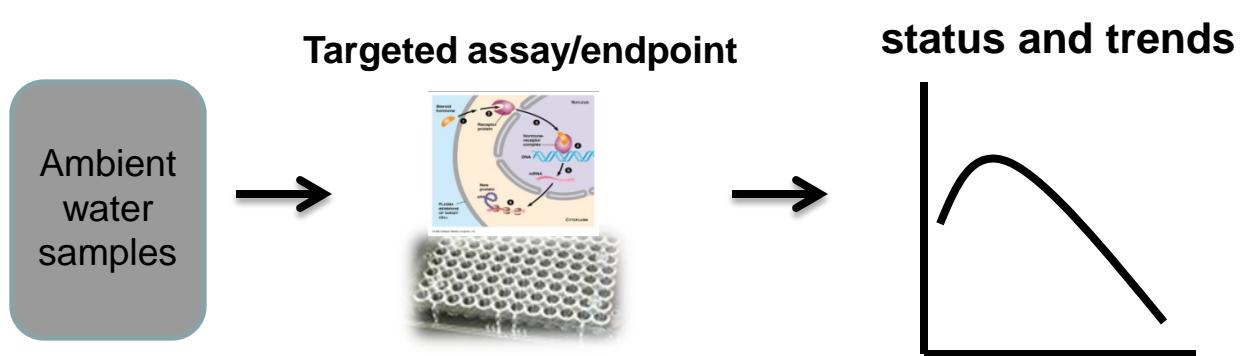
Surveillance



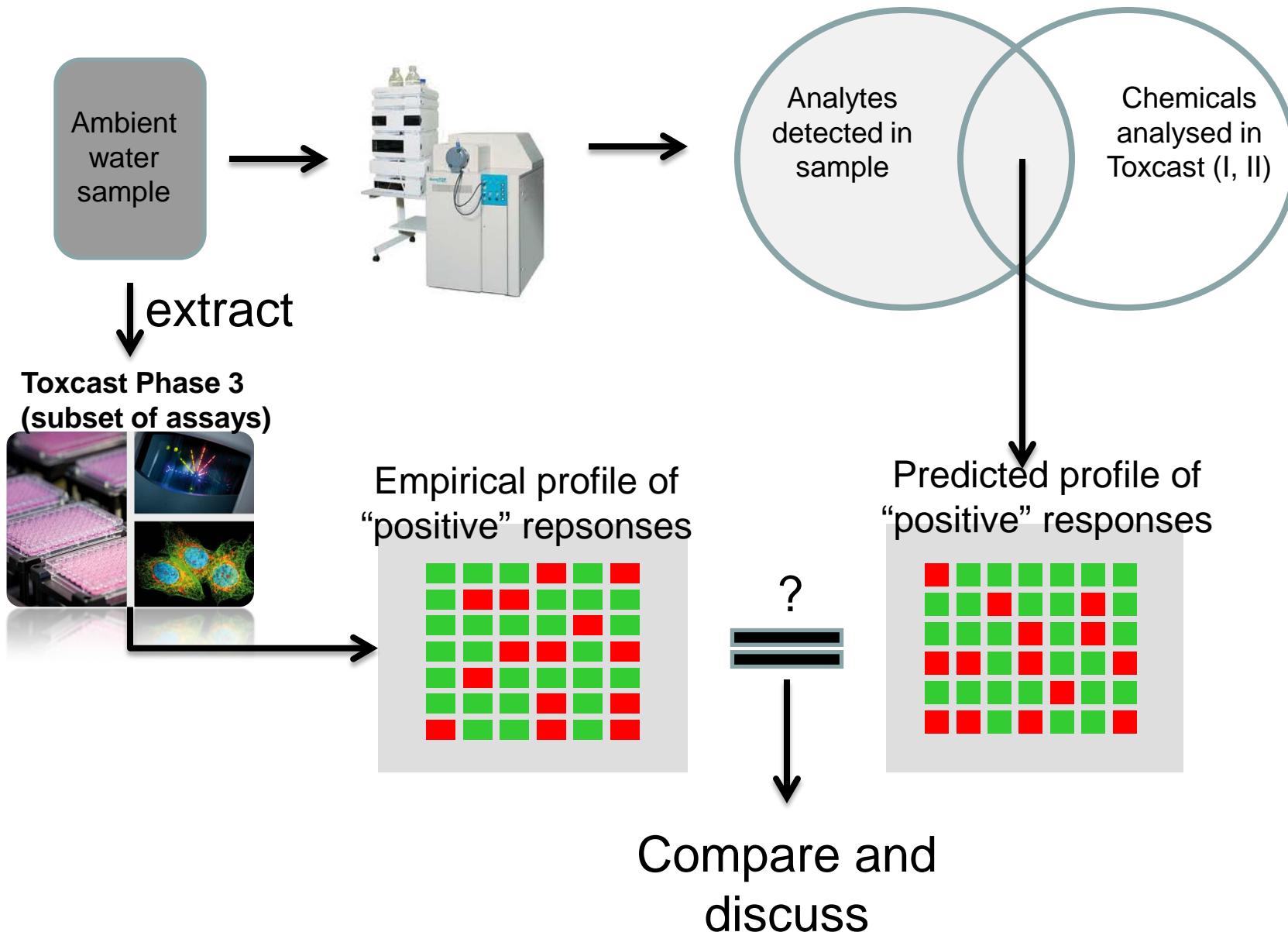
Assay/Endpoint Selection



Monitoring



Predicting effects of mixtures



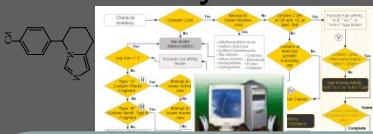
Conclusions

- Computational toxicology has important role to play in 21st C ecotoxicology
- AOPs are a critical foundation

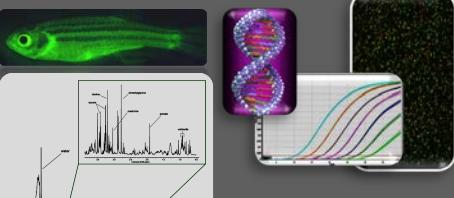
Alternative Data for Prospective Assessments

Molecular Cellular Tissue Organ Organ System Individual Population Ecosystem

QSAR, expert systems



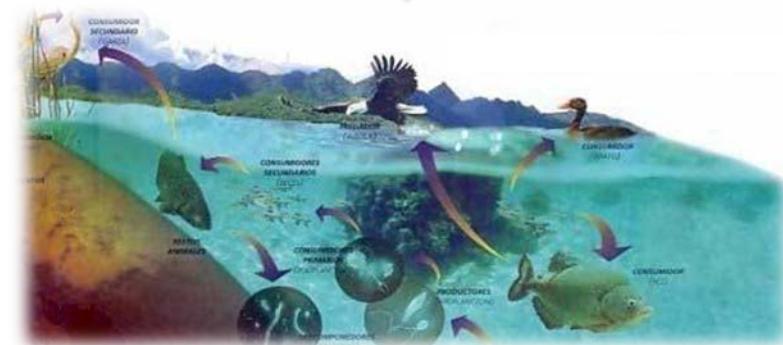
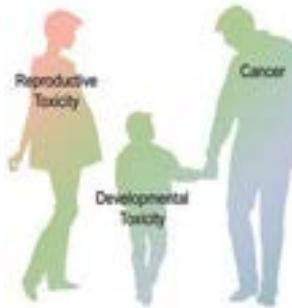
Biomarkers



Screening & Prioritization Toxcast, Tox21, in vitro



AOP



Demographic significance

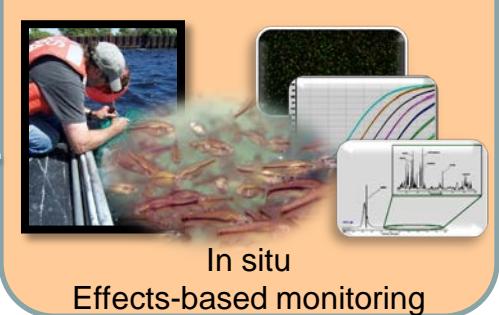
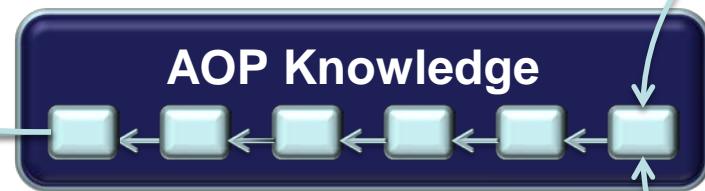
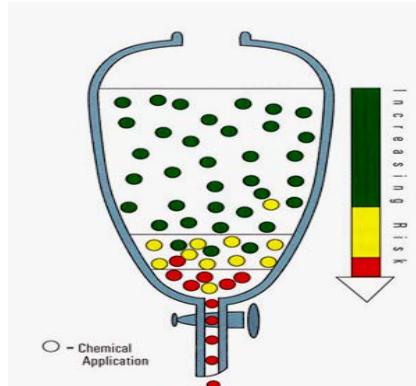
- Survival
- growth-development
- reproduction

Effects Data for Diagnostic Assessments

Molecular Cellular Tissue Organ Organ System Individual Population Ecosystem

Endpoint selection for monitoring

- Remediation/restoration
- Effectiveness of regulation



Diagnostic assessments
High priority – chemical(s),
class(es)

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