

# Systematic Analysis of High-Throughput Transcriptomics to Identify Potential Carcinogens

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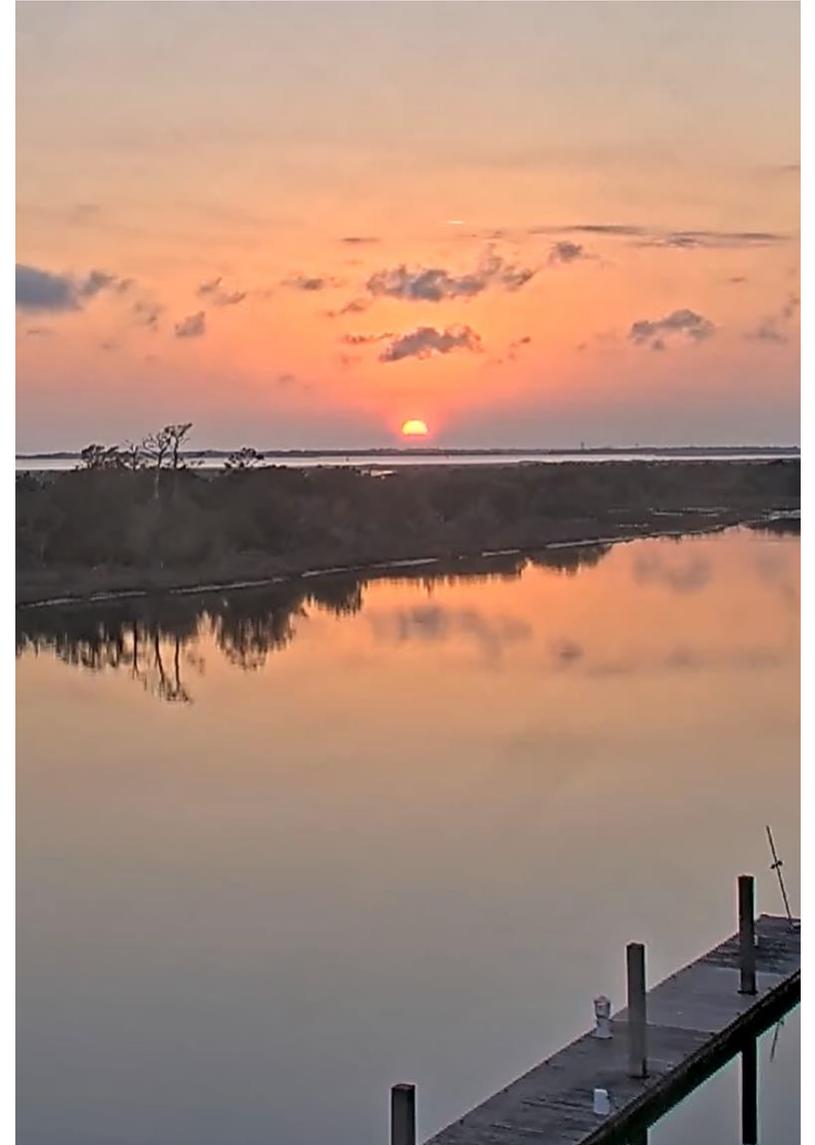


# Disclaimer

- The views expressed are those of Dr. Chris Corton and do not reflect US-EPA policy or product endorsement by the US-EPA.

# Sunsetting the 2-year Bioassay

- The 2-year bioassay – expensive, time-consuming, many animals used, questionable relevance to humans
- Many publications arguing that it is time to use modern approaches to replace the assay
- Complex problem – how to implement a testing strategy that is health protective and can be accepted by regulatory agencies?
- Will likely require both shorter-term exposures in vivo and assessment of effects in vitro



# Building an IATA to Identify Human Non-genotoxic Carcinogens (NGTxC)

A general integrated approach for the testing and assessment of non-genotoxic carcinogens

Archives of Toxicology (2020) 94:2899–2923  
https://doi.org/10.1007/s00204-020-02784-5

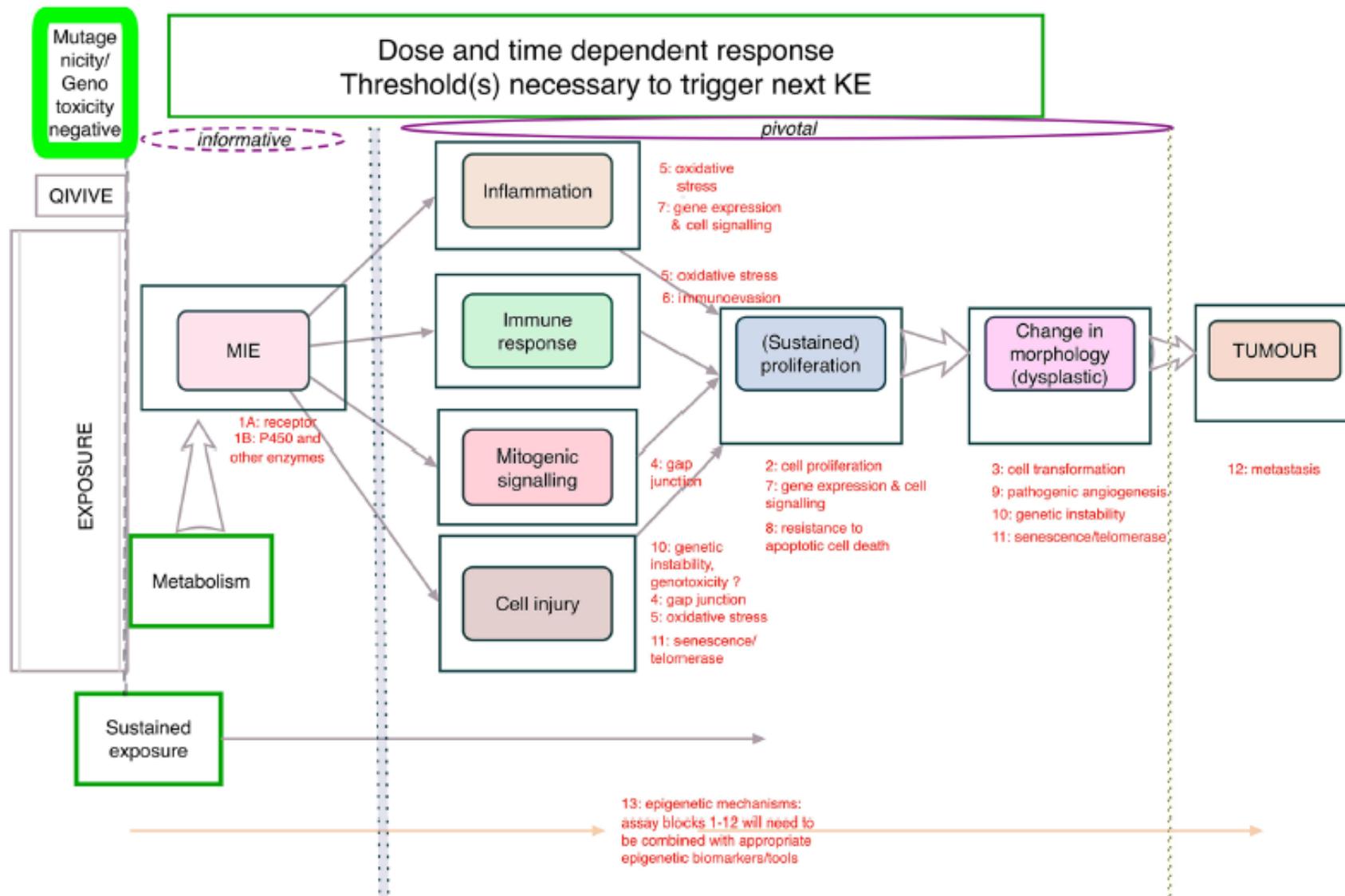
MEETING REPORTS



Chemical carcinogen safety testing: OECD expert group international consensus on the development of an integrated approach for the testing and assessment of chemical non-genotoxic carcinogens

Miriam N. Jacobs<sup>1</sup> · Annamaria Colacci<sup>2</sup> · Raffaella Corvi<sup>3</sup> · Monica Vaccari<sup>2</sup> · M. Cecilia Aguila<sup>4</sup> · Marco Corvaro<sup>5</sup> · Nathalie Delrue<sup>6</sup> · Daniel Desaulniers<sup>7</sup> · Norman Ertych<sup>8</sup> · Abigail Jacobs<sup>4</sup> · Mirjam Luijten<sup>9</sup> · Federica Madia<sup>2</sup> · Akiyoshi Nishikawa<sup>10</sup> · Kumiko Ogawa<sup>10</sup> · Kiyomi Ohmori<sup>11</sup> · Martin Paparella<sup>12</sup> · Anoop Kumar Sharma<sup>13</sup> · Paule Vasseur<sup>14</sup>

- OECD established an expert group to develop an IATA for identification of NGTxC
- Developed an overarching IATA framework based on key hallmarks of carcinogens –modules in boxes
- Identified in vitro and subchronic in vivo assays to measure the hallmarks in human cancer AOPs



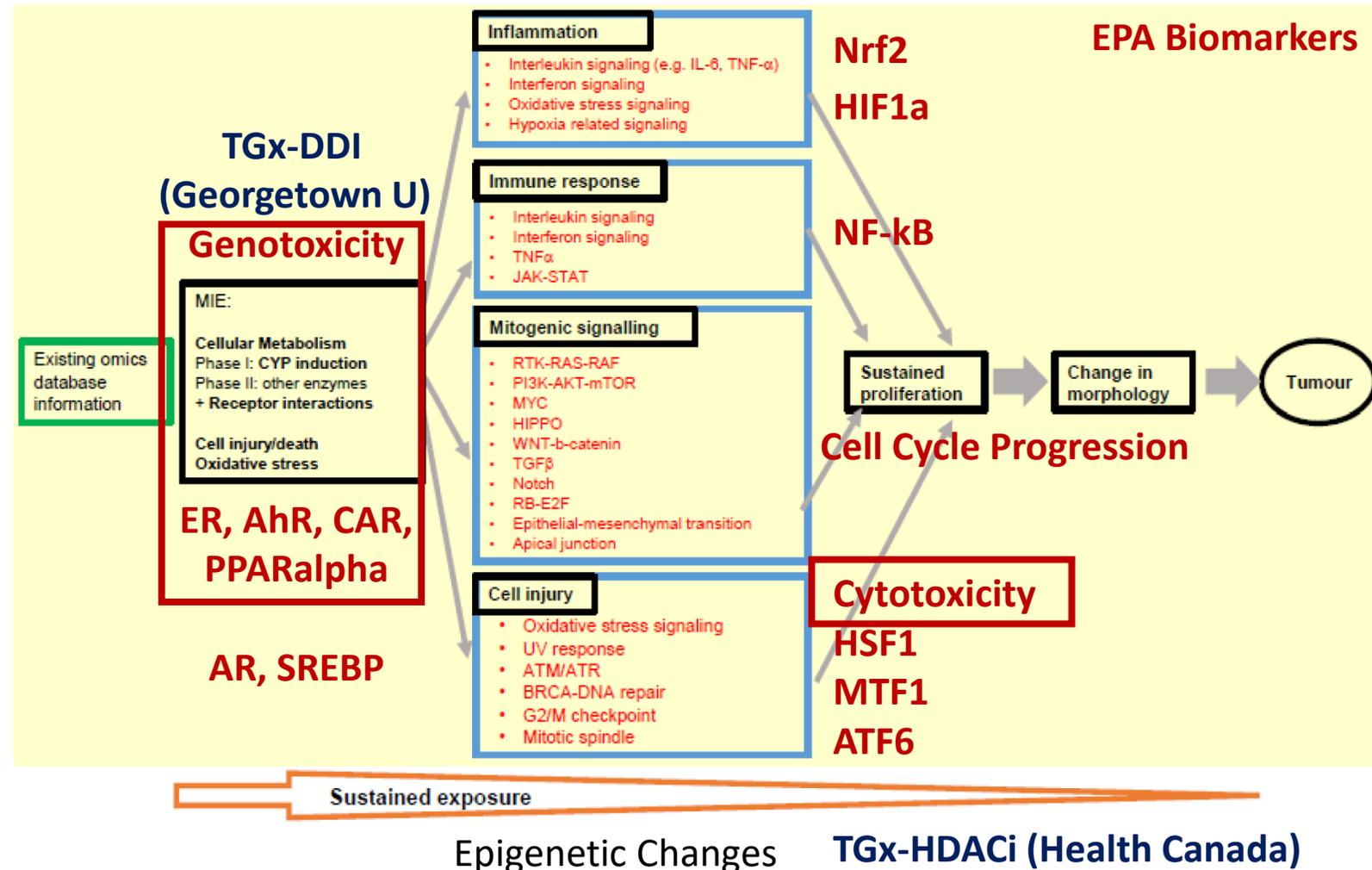
# Using Transcriptomics to Augment an IATA for Non-genotoxic Carcinogens

Review

## Analyses of Transcriptomics Cell Signalling for Pre-Screening Applications in the Integrated Approach for Testing and Assessment of Non-Genotoxic Carcinogens

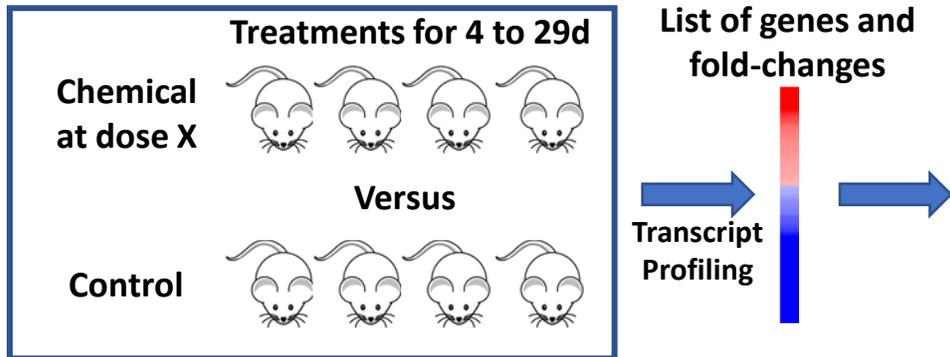
Yusuke Oku <sup>1,\*</sup>, Federica Madia <sup>2,†</sup>, Pierre Lau <sup>3</sup>, Martin Paparella <sup>4</sup>, Timothy McGovern <sup>5</sup>, Mirjam Luijten <sup>6</sup> and Miriam N. Jacobs <sup>7,\*</sup>

- Proposed to use available omics database information to monitor the key events of inflammation, immune response, mitogenic signaling and cell injury, in the NGTxC IATA
- Signaling pathways contributing to carcinogenesis** linked to the key hallmarks in the IATA
- Transcriptomics would be used in conjunction with cell-based assays
- Their proposal utilizes lists of genes that are linked to key hallmarks from MSigDB
- Weaknesses of the gene lists
  - Likely cell- or tissue-specific
  - Lists of genes have not been examined for ability to predict an effect
- Hypothesis: Biomarkers with known context of use and accuracy would complement the MSigDB gene lists to predict effects**
- Two examples of the use of biomarkers
  - Identify rat liver tumorigens
  - Identify chemicals that perturb pathways relevant to human chemical carcinogenesis
- Many predictions from one gene list



# NAM: Prediction of rat liver tumor induction using toxicogenomics analysis of short-term exposures

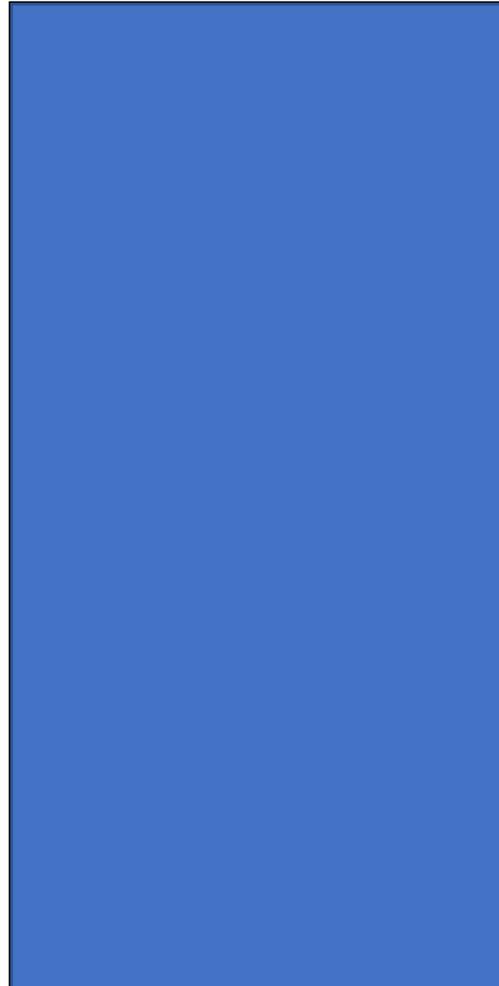
Would a chemical candidate at dose X cause increases in liver tumors in chronic studies?



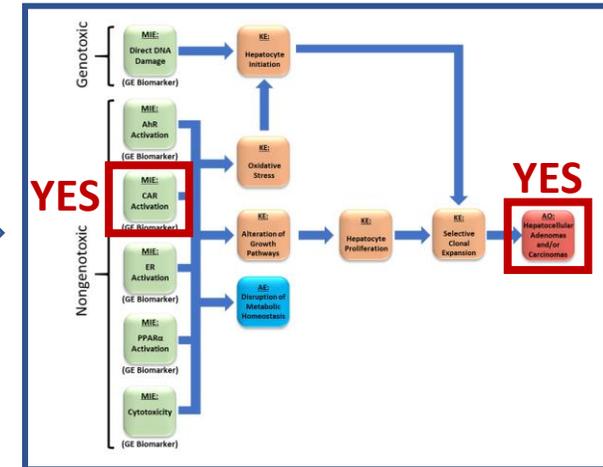
## Data Used to Construct the Model

- Microarray data
  - TG-GATES
  - DrugMatrix
- 2-year cancer data
  - Lhasa carcinogenicity database

## NAM Computational Model



## Network of Liver Cancer AOPs

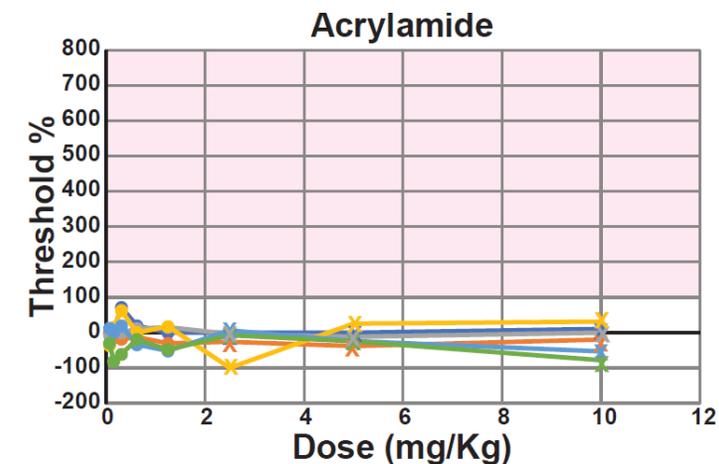
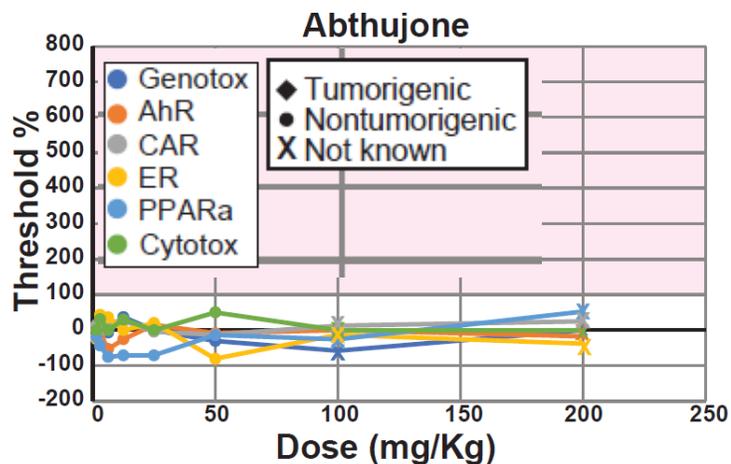
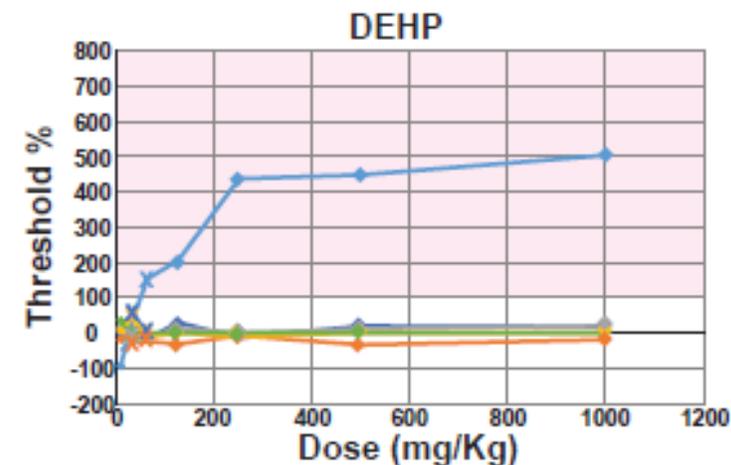
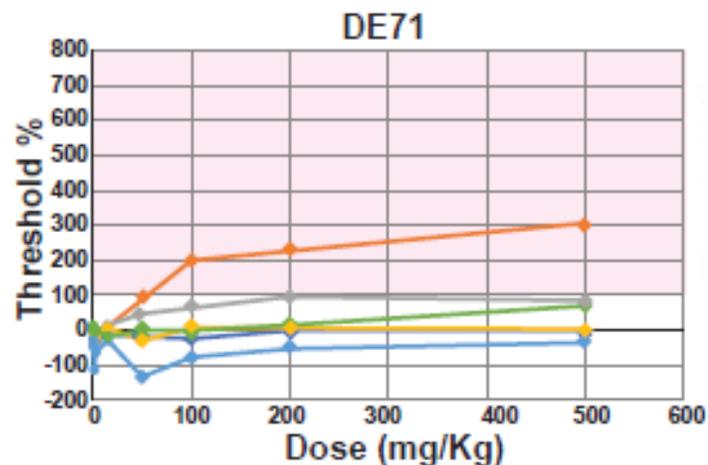


- Is the dose tumorigenic?
- Which mode(s) of action is activated?
- Is the mode(s) of action human irrelevant?
- Is a waiver for testing appropriate?

- Examined ~250 chemicals (~50 caused liver tumors)
- Accuracy was ~75-95% depending on the dataset used
- Accuracy is independent of platform used to assess gene expression

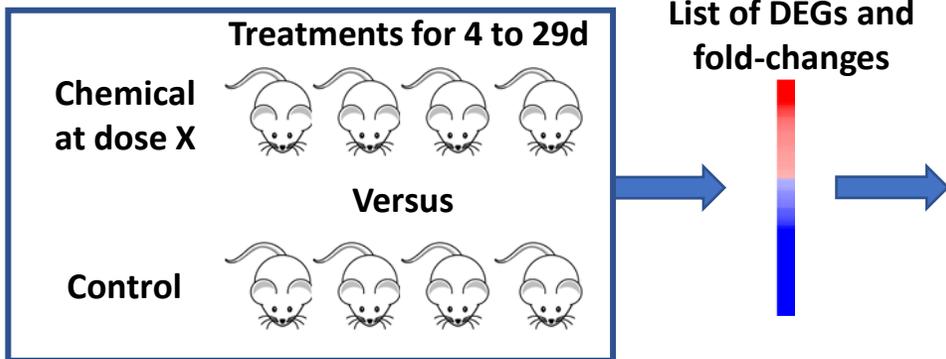
# NAM identifies chemical-dose pairs that are tumorigenic in the liver using TempO-Seq

- Examined 16 chemicals at up to 10 doses; 5d exposures (Gwinn et al., 2021 ToxSci)
- Liver gene expression analyzed using full genome TempO-Seq
- Model correctly identified all tumorigenic chemicals
- Balanced accuracies = 74-91% depending on the tumorigenic activation level used and whether individual chem-doses were considered or all doses for a chemical



# NAM: Prediction of rat liver tumor induction using toxicogenomics analysis of short-term exposures

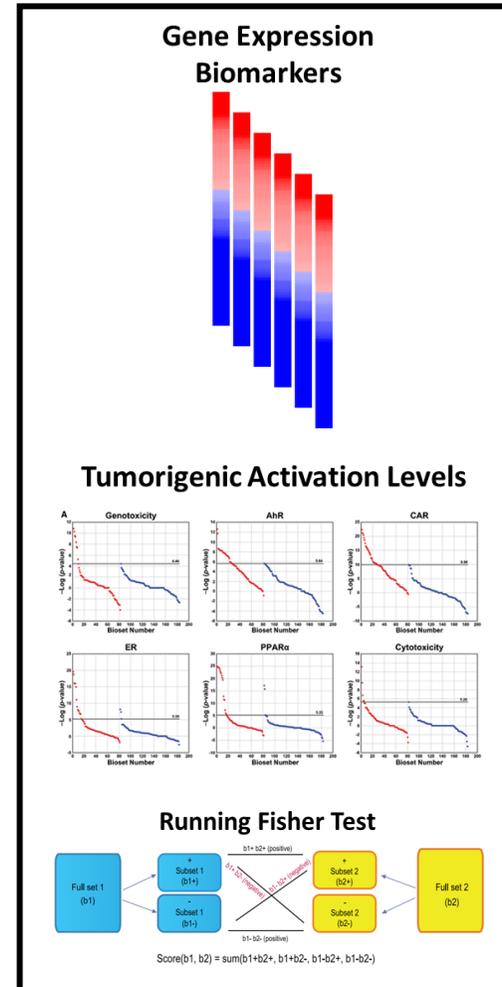
Will a chemical candidate at dose X cause increases in liver tumors in chronic studies?



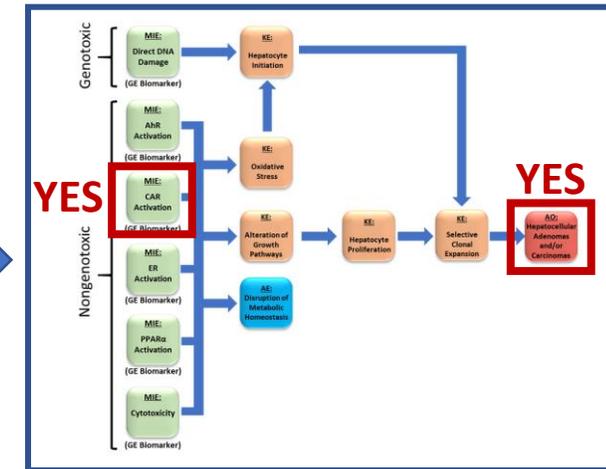
## Questions still to be addressed:

- Can we improve accuracy by incorporating
  - More data?
  - A greater diversity of chemicals?
  - Wild-type and null rat comparisons?

## NAM Computational Model



## Network of Liver Cancer AOPs



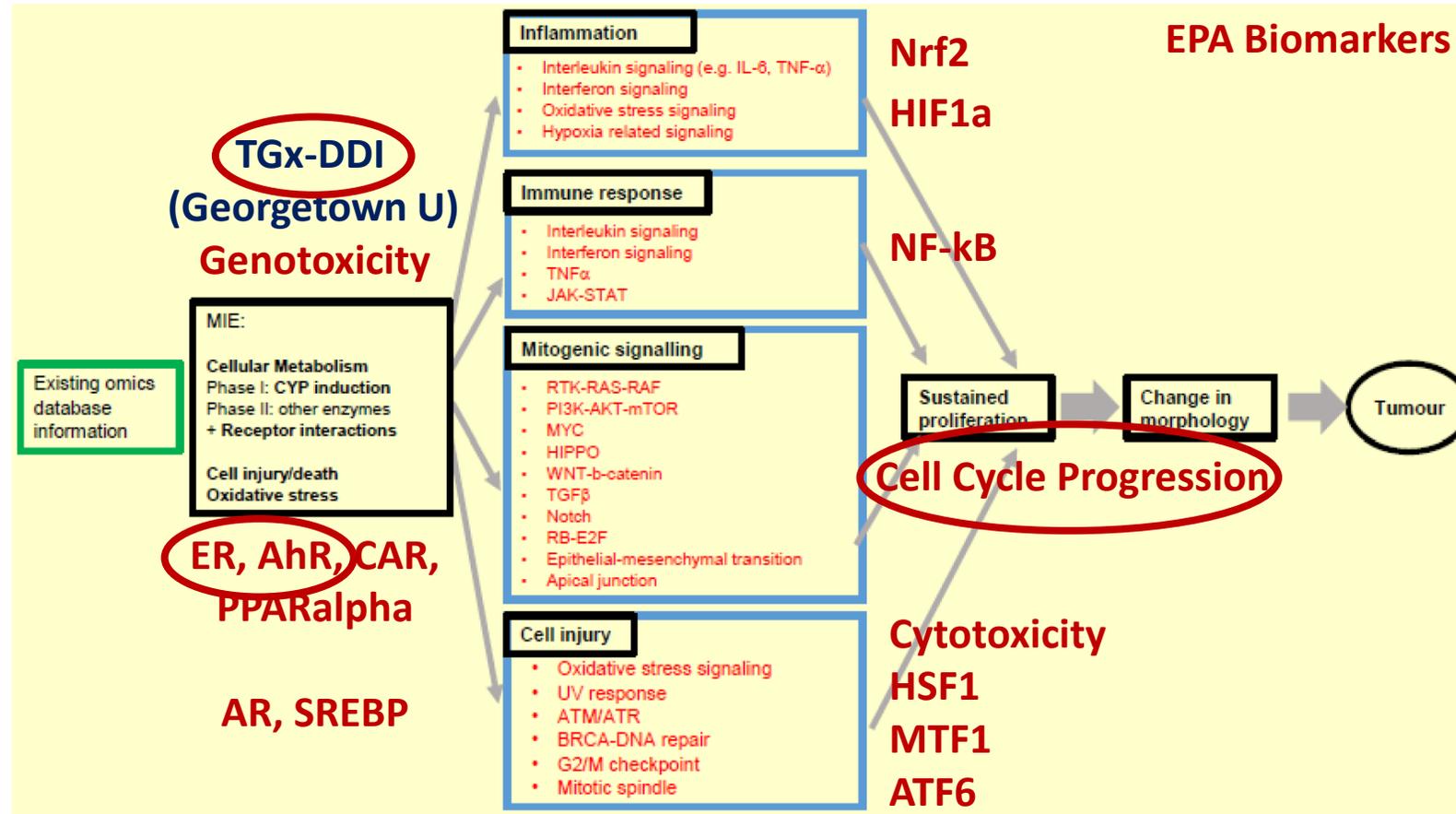
- Is the dose tumorigenic?
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Emerging Systems Toxicology for  
the Assessment of Risk (eSTAR)  
Committee

## Future Studies:

- Studies conducted through the HESI eSTAR Carcinogenomics Workgroup

# Application of biomarkers to identify effects of chemicals in human cells



**Sustained exposure** (orange arrow)

Epigenetic Changes

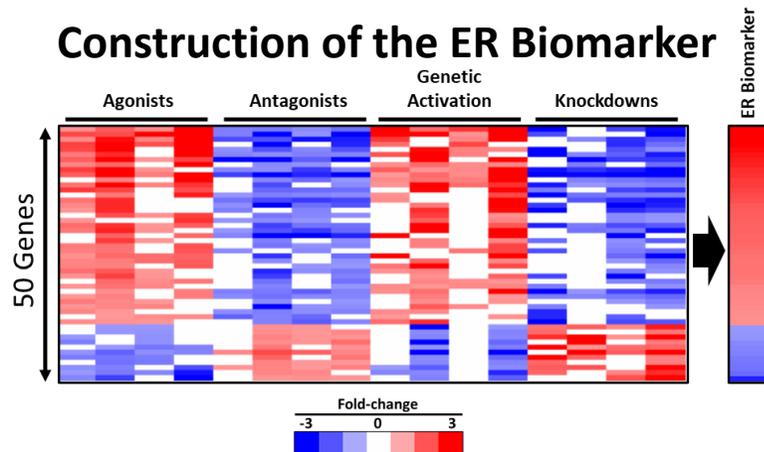
TGx-HDACi (Health Canada)



# Creating Predictive Biomarkers

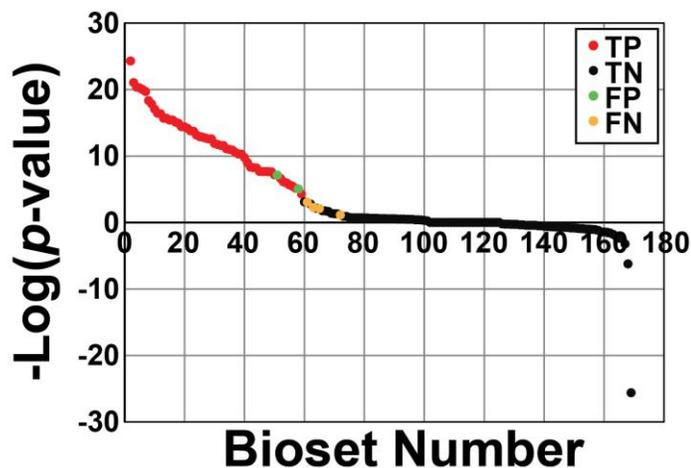
- Assemble your tools
  - Reference compounds – what is their predicted behavior?
  - Are there any examples of the gene knocked out/down or overexpressed/activated?
- Generate the profiles in which the factor is activated or suppressed in the system of interest
- Use computational approaches for identification of predictive gene sets
  - Machine learning
  - Weight of evidence

## Construction of the ER Biomarker

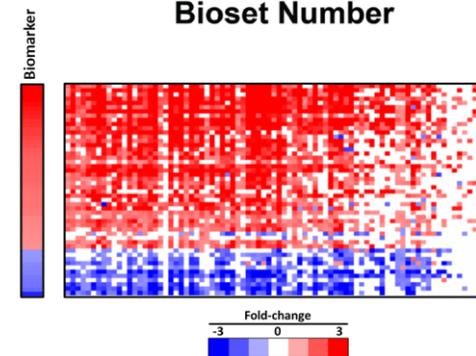
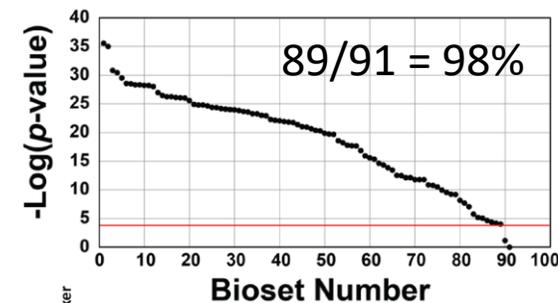


50-gene biomarker built from profiles of

- 4 ER agonists
- 4 ER antagonists
- 4 constitutively active ER mutants
- 4 knockdowns of *ESR1* expression



## The ER Biomarker identifies E2 treatments in MCF-7 cells

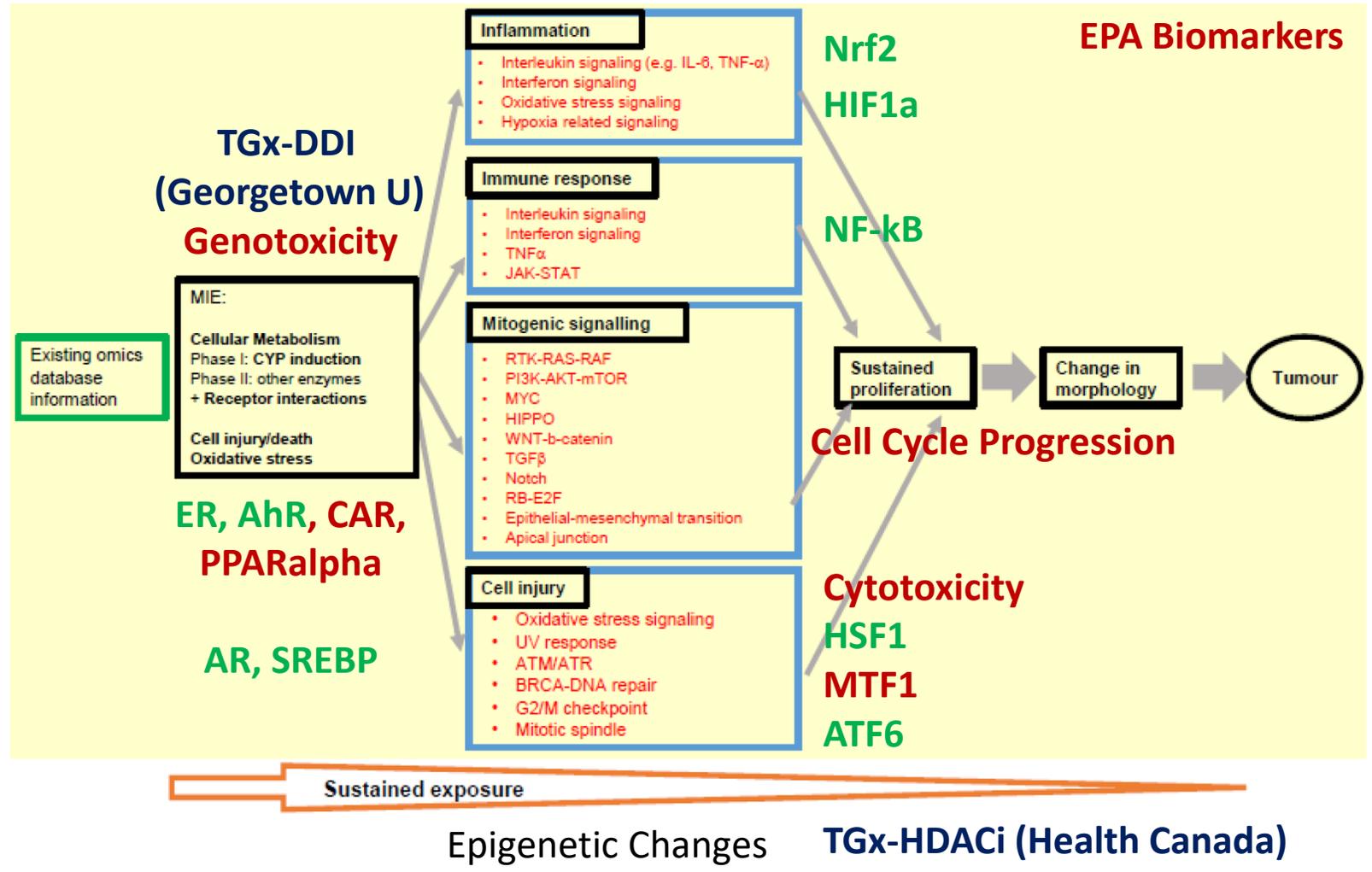


Using the NCATS Tox21 ER trans-activation assays as the reference data set:

- Balanced accuracy = 96%
- Context of use: ER positive human breast cancer cell lines

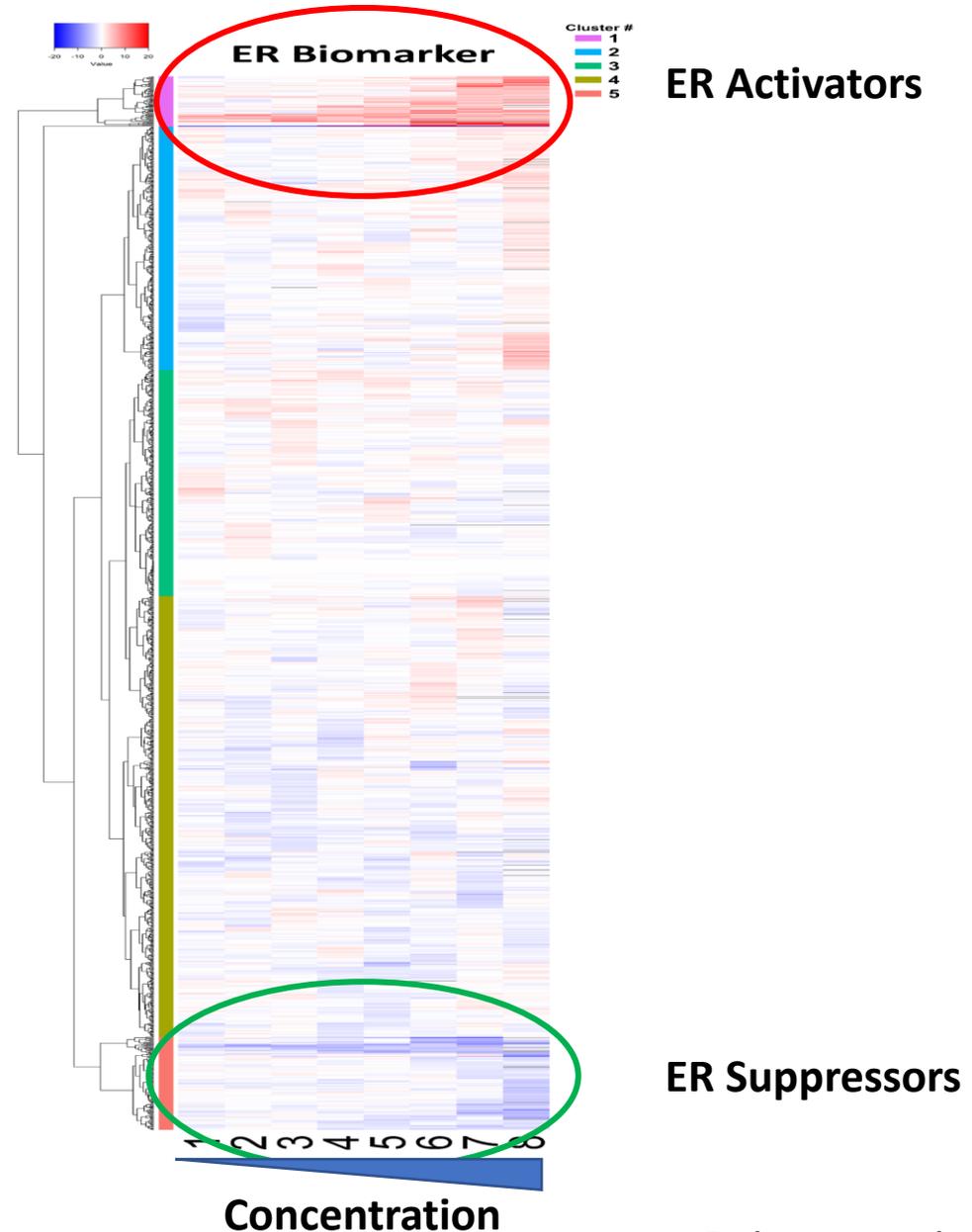
# Application of biomarkers to identify effects of chemicals in human cells

Biomarkers built using profiles from cells in which the factor was genetically modified



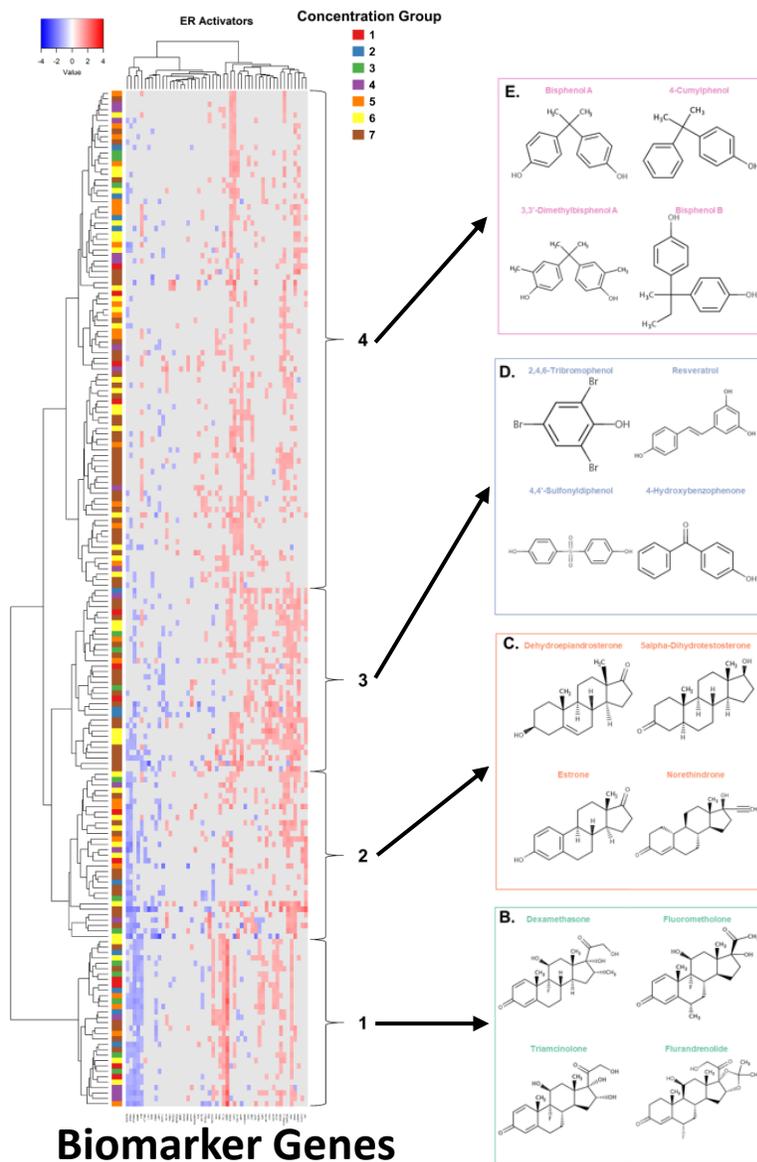
# Identification of ER modulators using an estrogen receptor biomarker in MCF-7 cells

- Estrogen receptor activation is associated with increases in cancers of breast and ovaries
- Examined transcript changes in MCF-7 cells treated with ~1600 chemicals at 8 concentrations (~12,800 comparisons)
- Compared the profiles to the 50-gene estrogen receptor (ER) biomarker
- Values expressed as  $-\text{Log}(\text{p-value})$ s of the correlation between the profile and the ER biomarker
- 1D hierarchical clustering of chemicals across 8 concentrations



# ER activators regulate ER biomarker genes in a structure-dependent manner

- 2D hierarchical clustering of ~120 chem-concentration pairs that activated ER vs. ER biomarker genes
- 4 major clusters of chemical-concentration pairs

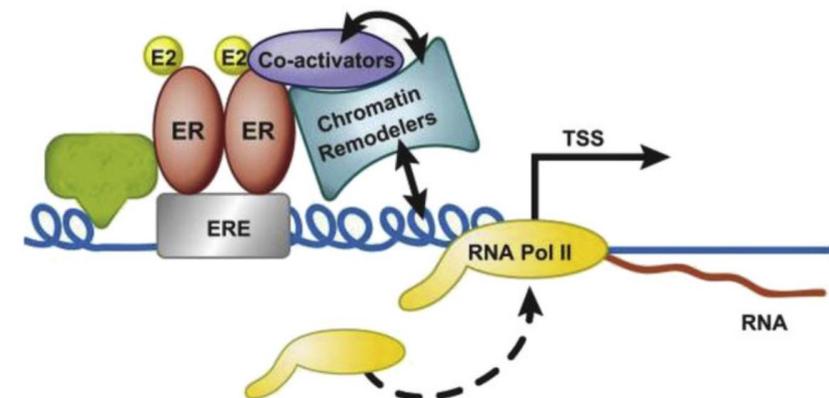


**Bisphenols**

**Misc  
activators**

**Classical  
estrogens**

**GR and PR  
agonists**



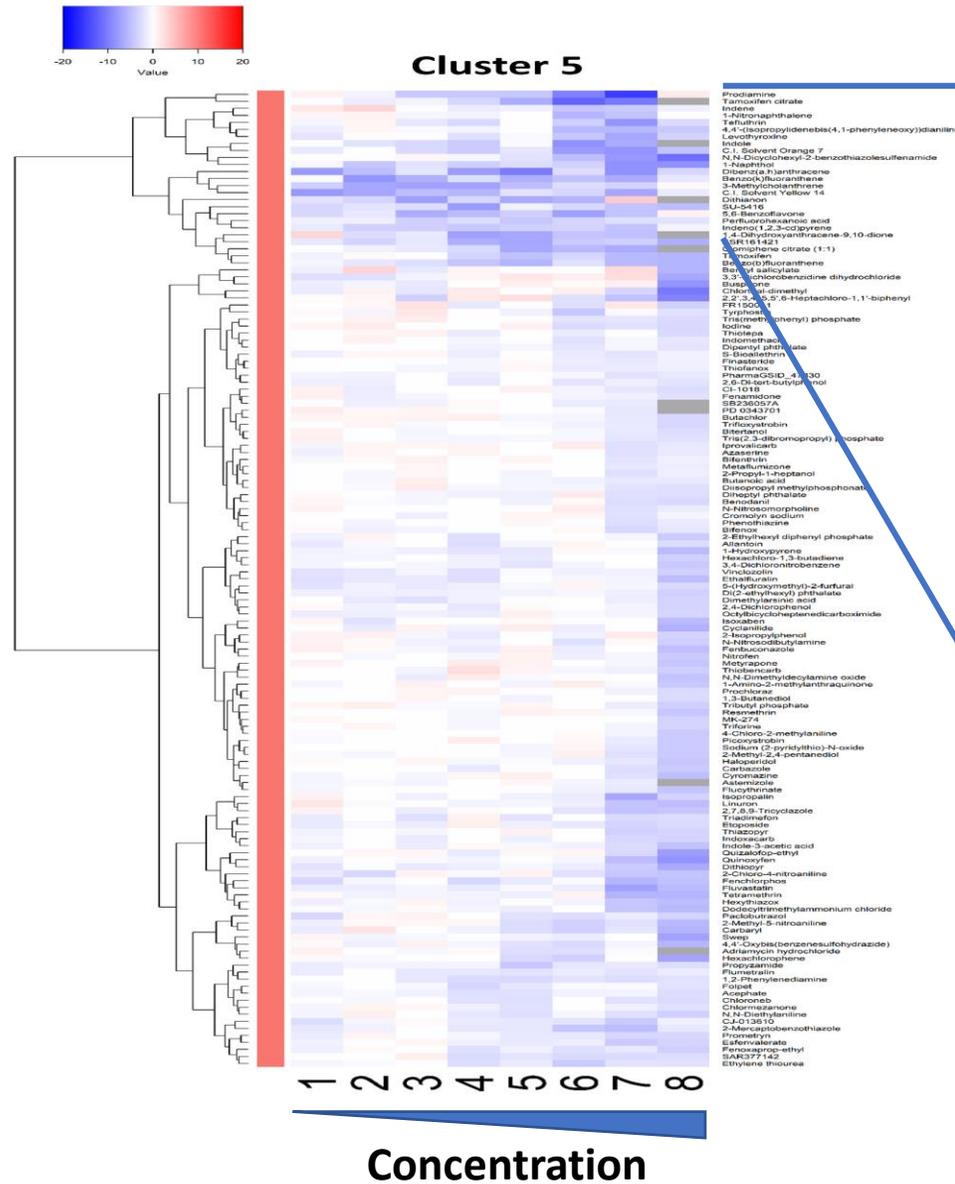
**Results consistent with**

- Agonists induce different conformations of the receptor
- ER conformation determines which co-activators interact
- ER-co-activator complexes determine which genes are activated

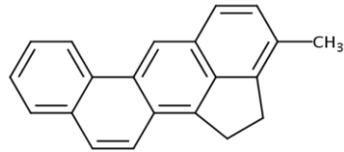
Robarts et al., in preparation

# Many ER suppressors appear to be AhR activators

- Examined transcript changes in MCF-7 cells treated with ~1600 chemicals at 8 concentrations (~12,800 comparisons)
- Compared the profiles to the 50-gene estrogen receptor (ER) biomarker
- 1D hierarchical clustering of chemicals across 8 concentrations

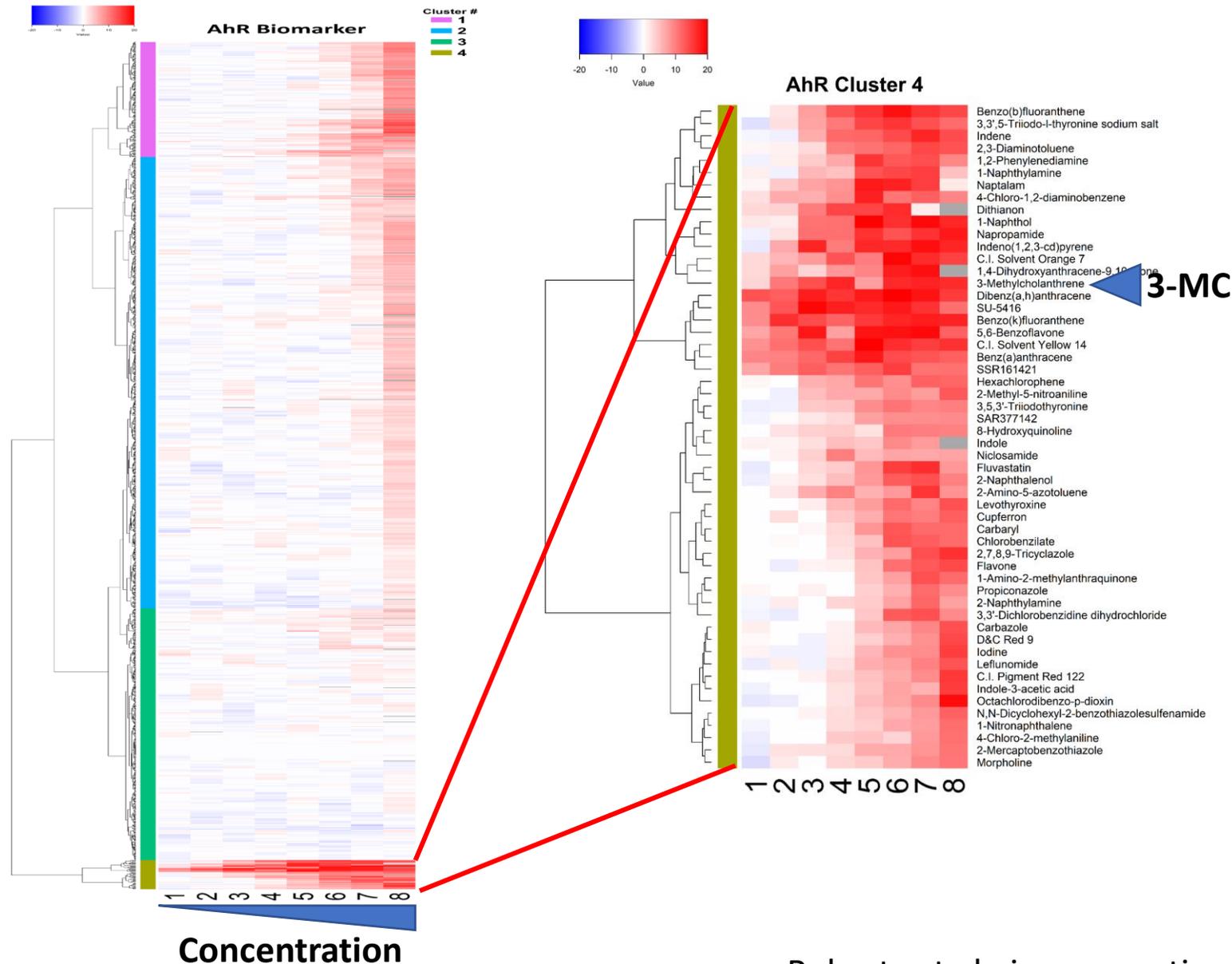


- Prodiamine
- Tamoxifen citrate
- Indene
- 1-Nitronaphthalene
- Tefluthrin
- 4,4'-(Isopropylidenebis(4,1-phenyleneoxy))dianiline
- Levothyroxine
- Indole
- C.I. Solvent Orange 7
- N,N-Dicyclohexyl-2-benzothiazolesulfenamide
- 1-Naphthol
- Dibenz(a,h)anthracene
- Benzo(k)fluoranthene
- 3-Methylcholanthrene
- C.I. Solvent Yellow 14
- Dithianon
- SU-5416
- 5,6-Benzoflavone
- Perfluorohexanoic acid
- Indeno(1,2,3-cd)pyrene
- 1,4-Dihydroxyanthracene-9,10-dione



# Identification of AhR activators in an HTTr screen in MCF-7 cells

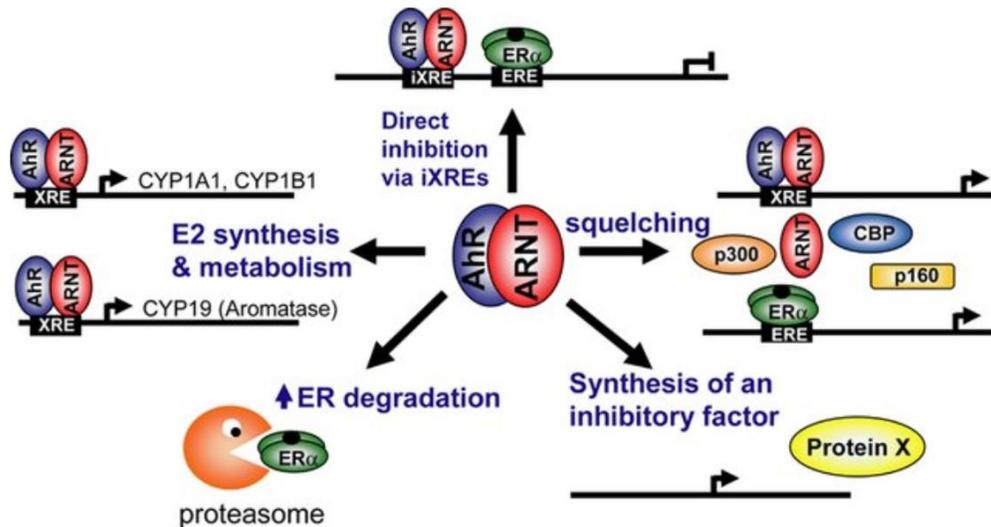
- Activation of AhR by TCDD is associated with a number of human tumors (e.g., breast, endometrium, testicular, liver, lung)
- Built and characterized a gene expression biomarker to identify AhR activators in MCF-7 cells
- 16 genes consistently regulated by 12 AhR activators and in the opposite direction by knockdown of *AHR* using gene-specific siRNA
- Compared the ~12,800 profiles to the AhR biomarker



# AhR activators suppress ER responses

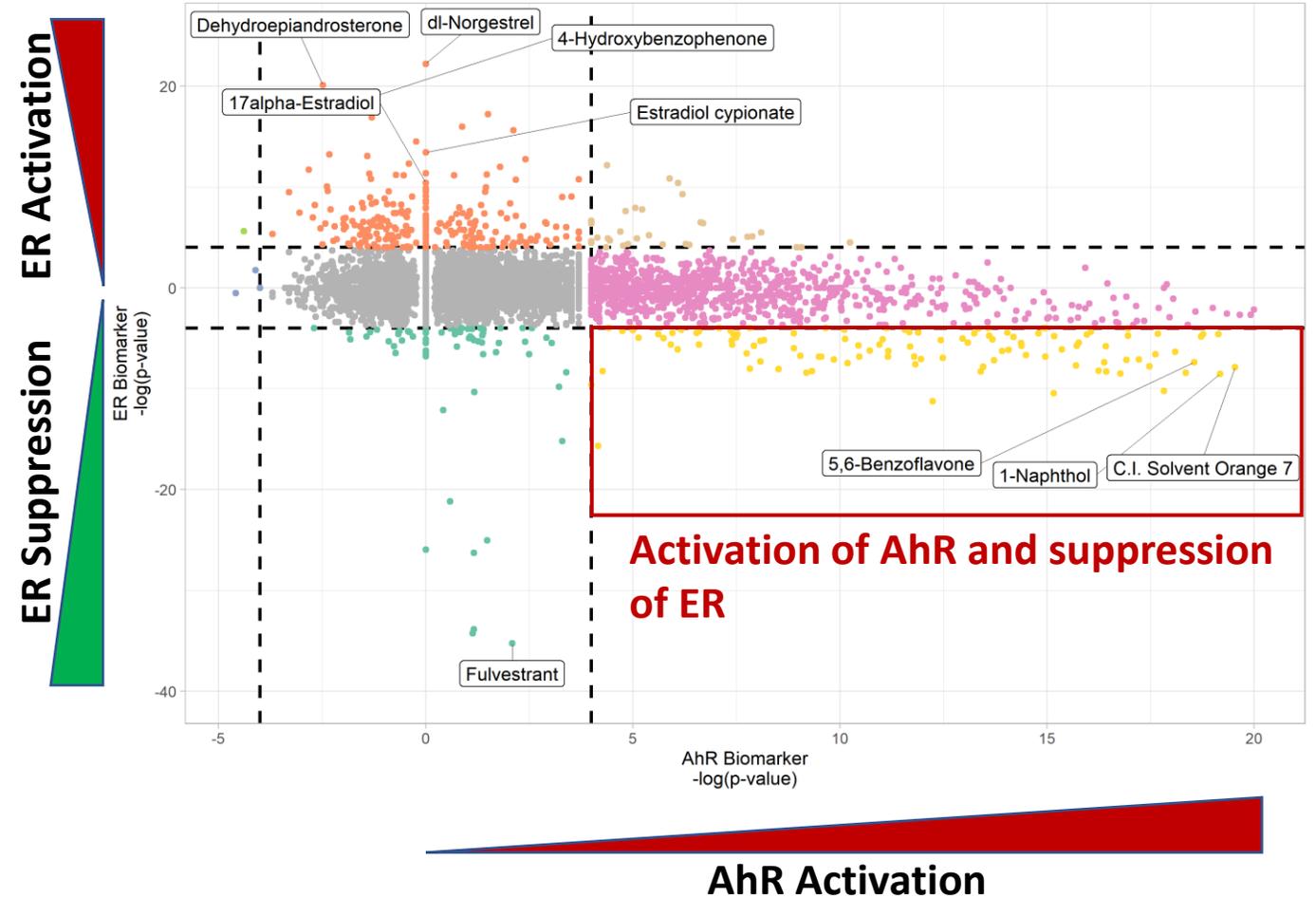
- Compared each of the ~12,800 profiles to the estrogen receptor (ER) and aryl hydrocarbon receptor (AhR) biomarkers

## Molecular basis for suppression of ER by AhR



From Nuclear Receptor Signaling 4(1):e016

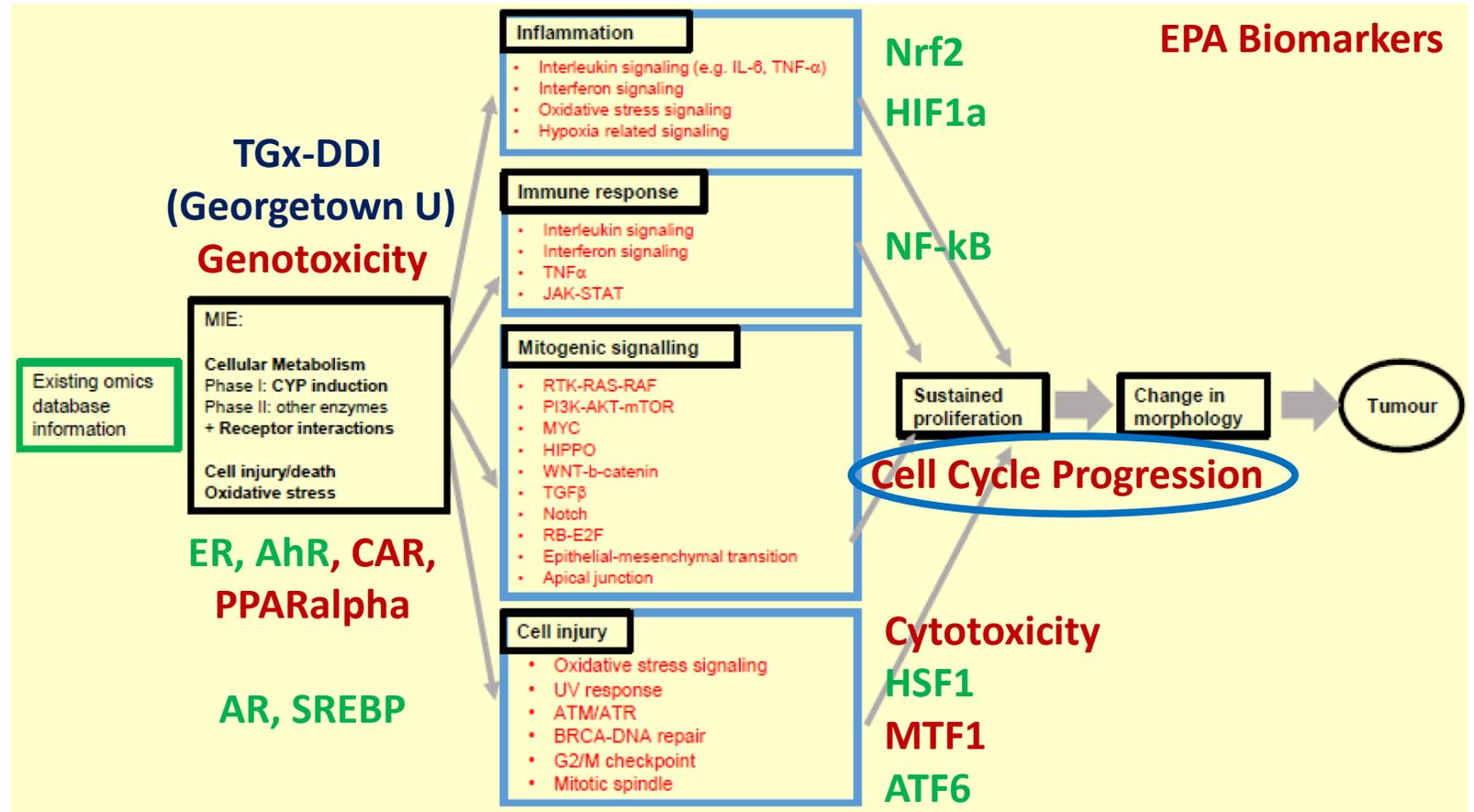
## AhR Biomarker vs ER Biomarker Activity



Robarts et al., in preparation

# Application of biomarkers to identify effects of chemicals in human cells

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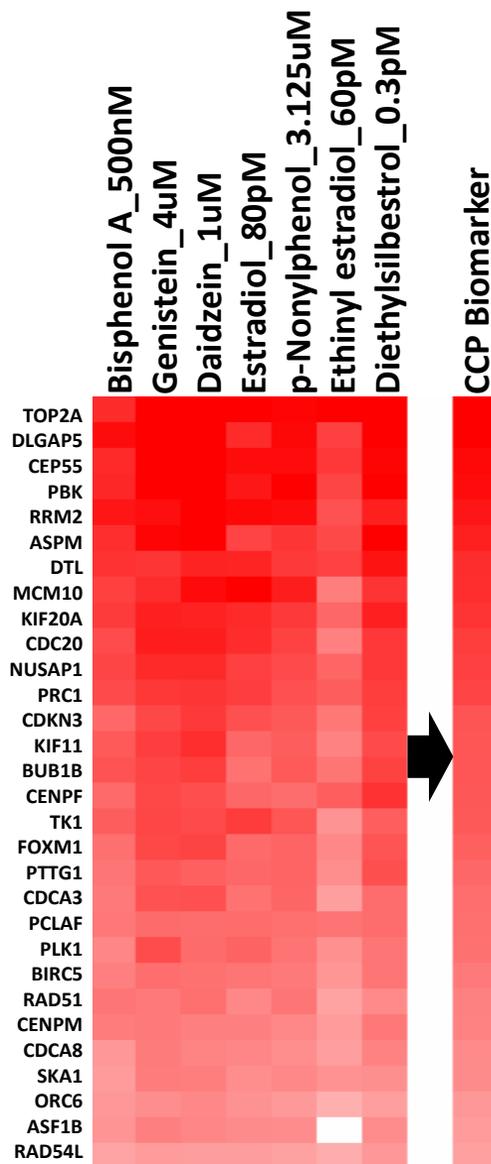
**Sustained exposure** (orange arrow)

Epigenetic Changes      **TGx-HDACi (Health Canada)**

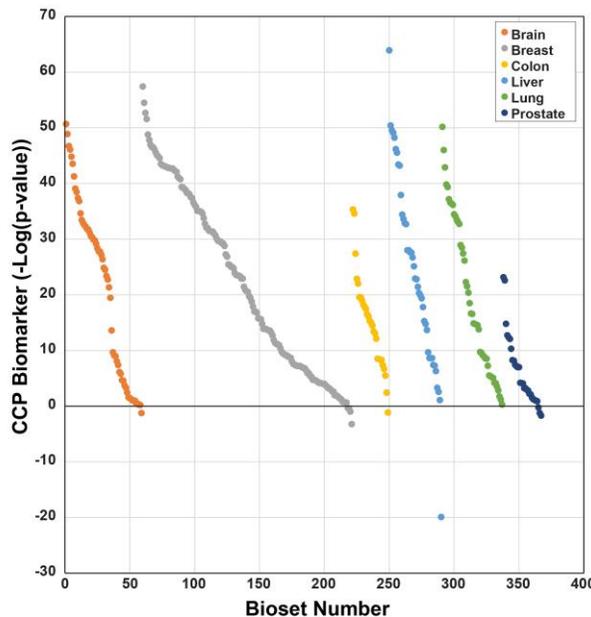


# The Cell Cycle Progression Biomarker

- 30 genes identified as being involved in cell cycle progression in human prostate tumors (Cuzick et al. (2011). Lancet Oncol. 12:245) – expression of genes associated with death from prostate cancer
- Examined expression of the genes after 48 hrs of treatment with 7 estrogen receptor activators in MCF-7 cells at concentrations known to induce cell proliferation
- Examined responses to chemicals and stressors in humans, rats and mice (~120K comparisons)

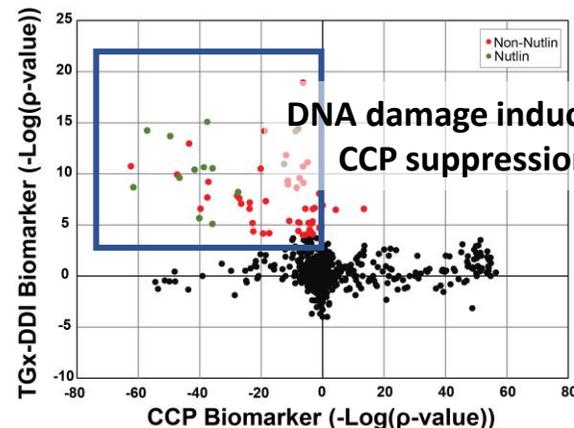


Can the biomarker identify proliferation in human tumor samples?



- Examined ~360 tumor vs surrounding tissue comparisons in 6 tissues
- Almost all give a positive response
- 84% were identified as positive for cell proliferation

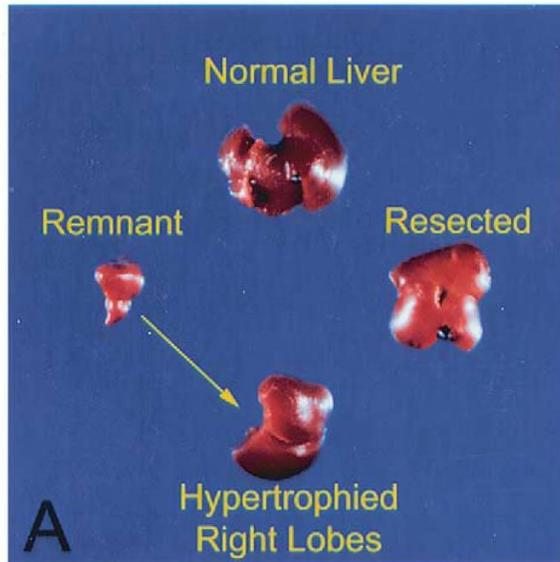
Can the biomarker identify conditions in which cell cycle is arrested in human cells?



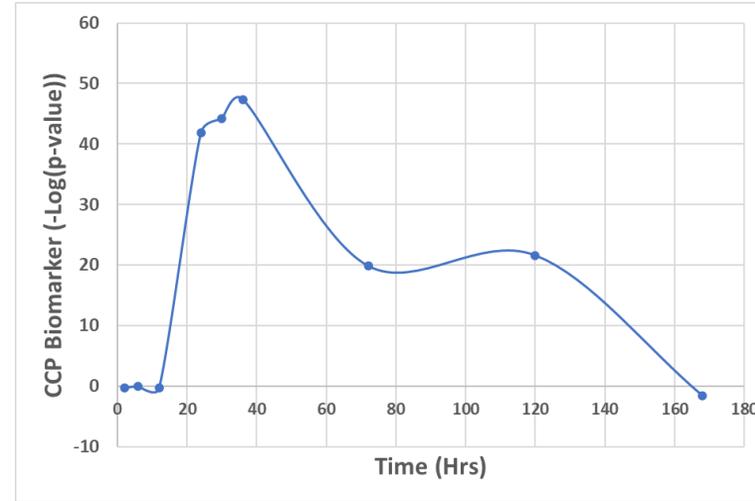
- Compared responses of the CCP biomarker to TGx-DDI biomarker
- P53 activators including **Nutlin** (stabilizes p53) and **genotoxic chemicals** suppress cell proliferation

# The CCP Biomarker Identifies Proliferation in Rats

- The 2/3rds partial hepatectomy is a classic model for studying liver regeneration

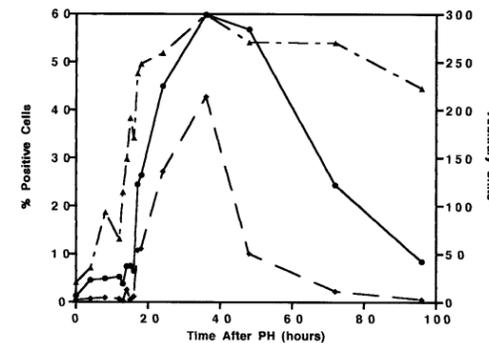


- Examined activation of CCP biomarker from 2 – 168 hrs after partial hepatectomy in male Sprague-Dawley rats
- Affymetrix data from GSE63742



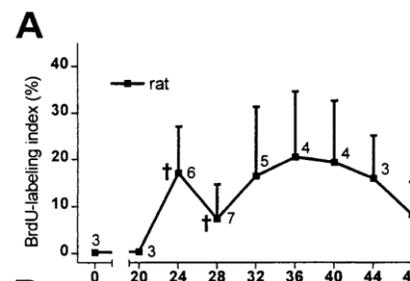
CCP Biomarker:  
Peak at 36 hrs

Gerlach et al. Ki-67 Expression During Rat Liver Regeneration After Partial Hepatectomy. *Hepatology* 1997;26:573-578.



Male Wistar rats  
BrdU: Peak at 36 hrs  
Ki-67: Peak at 36 hrs

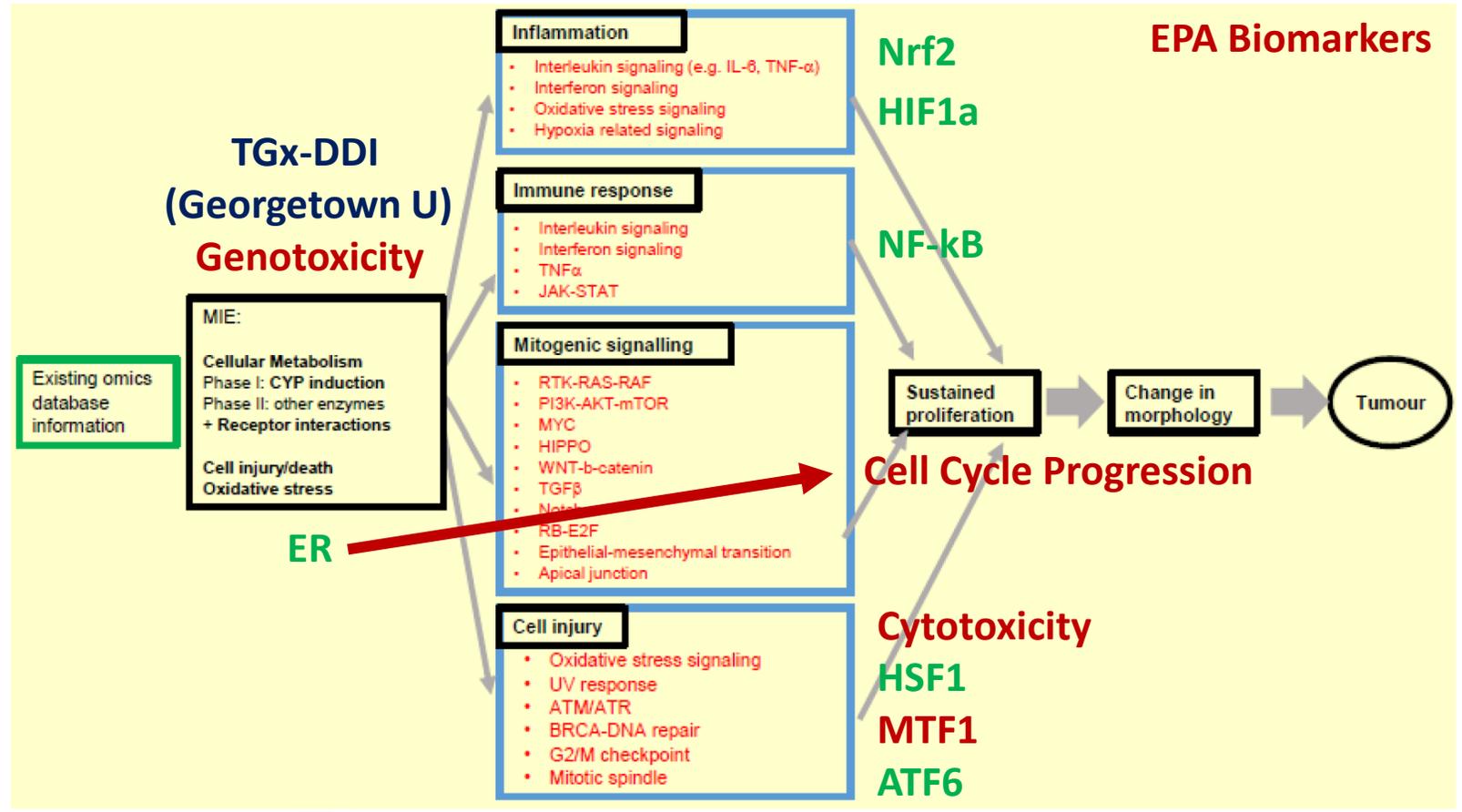
Weglaz and Sandgren Timing of hepatocyte entry into DNA synthesis after partial hepatectomy is cell autonomous. *PNAS* 2000 97: 12595



Male Fisher 344 rats  
BrdU: Peak at 36 hrs

# Application of biomarkers to identify effects of chemicals in human cells

Biomarkers built using profiles from cells in which the factor was genetically modified



ER

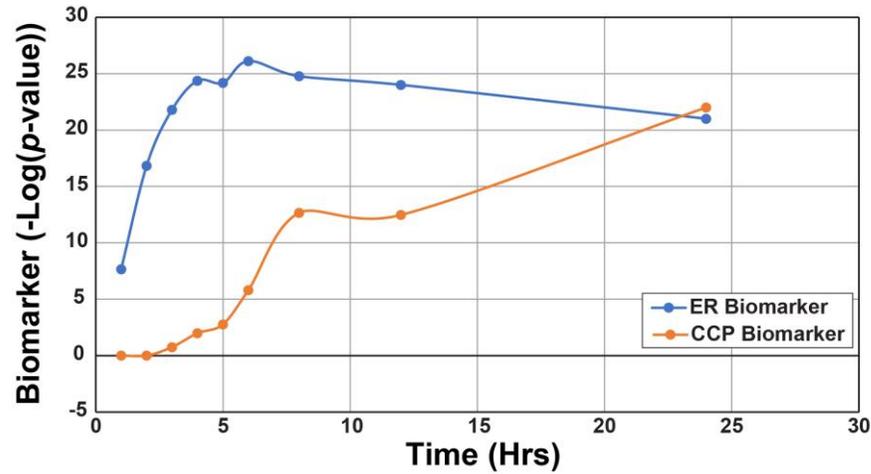
Sustained exposure

Epigenetic Changes TGx-HDACi (Health Canada)



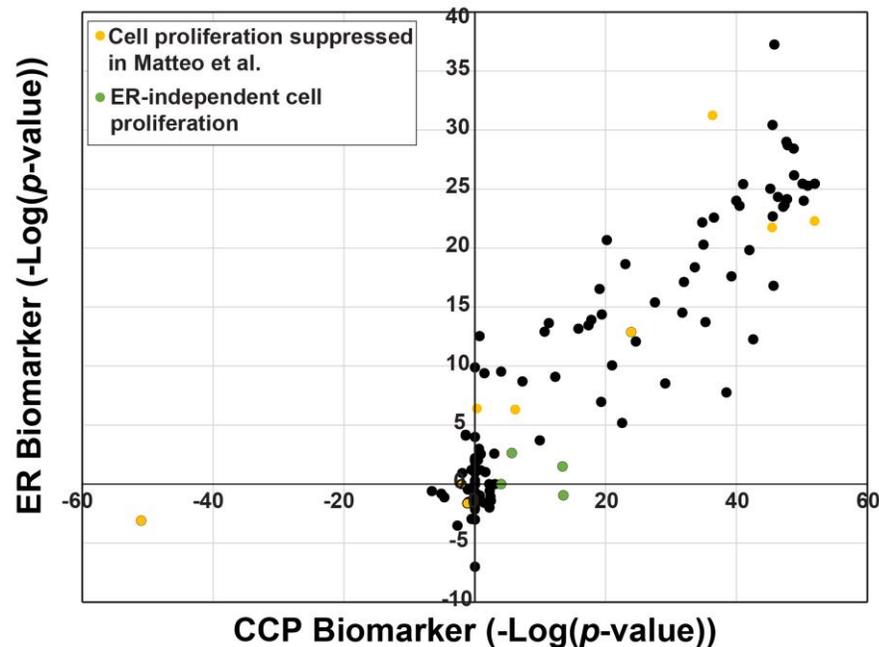
# Linking Estrogen Receptor Activation with Cell Proliferation

## ER activation precedes cell proliferation



- Treated MCF-7 cells with 10nM estradiol and examined gene expression out to 24 hrs
- Data from GSE78167 (Baran-Gale et al., 2016; RNA 22:1592)

## Relationship between ER activation and cell proliferation across 15 BPA alternatives

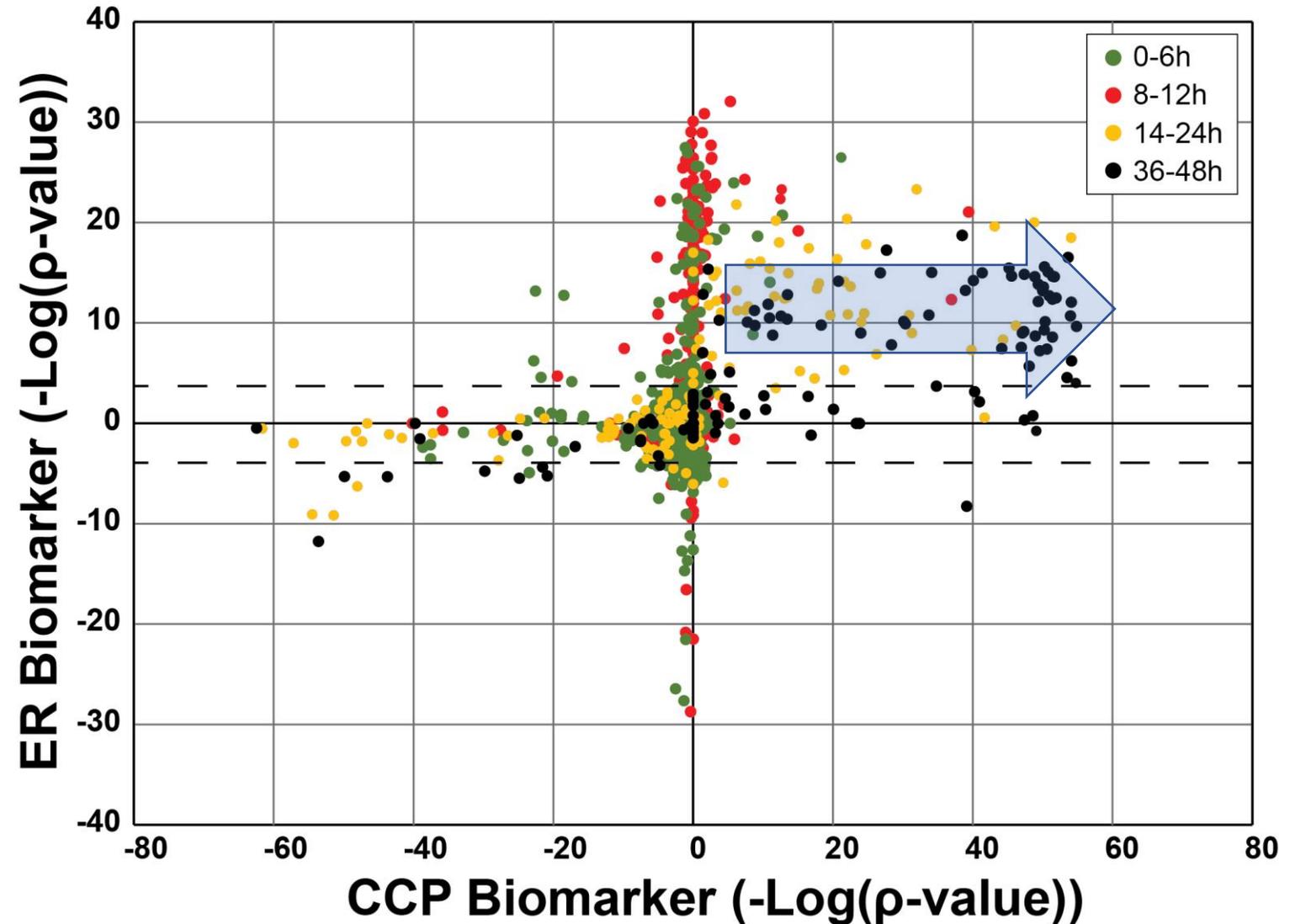


- Dataset from Matteo et al. 2023 ToxSci 191(2):266-275
- MCF-7 cells treated with BPA and 15 alternatives
- 0.0005-100  $\mu$ M for each chemical and 48 hr treatment times – 143 comparisons
- The level of ER activation determines the level of cell proliferation response – is there a threshold?

# Linking Estrogen Receptor Activation with Cell Proliferation

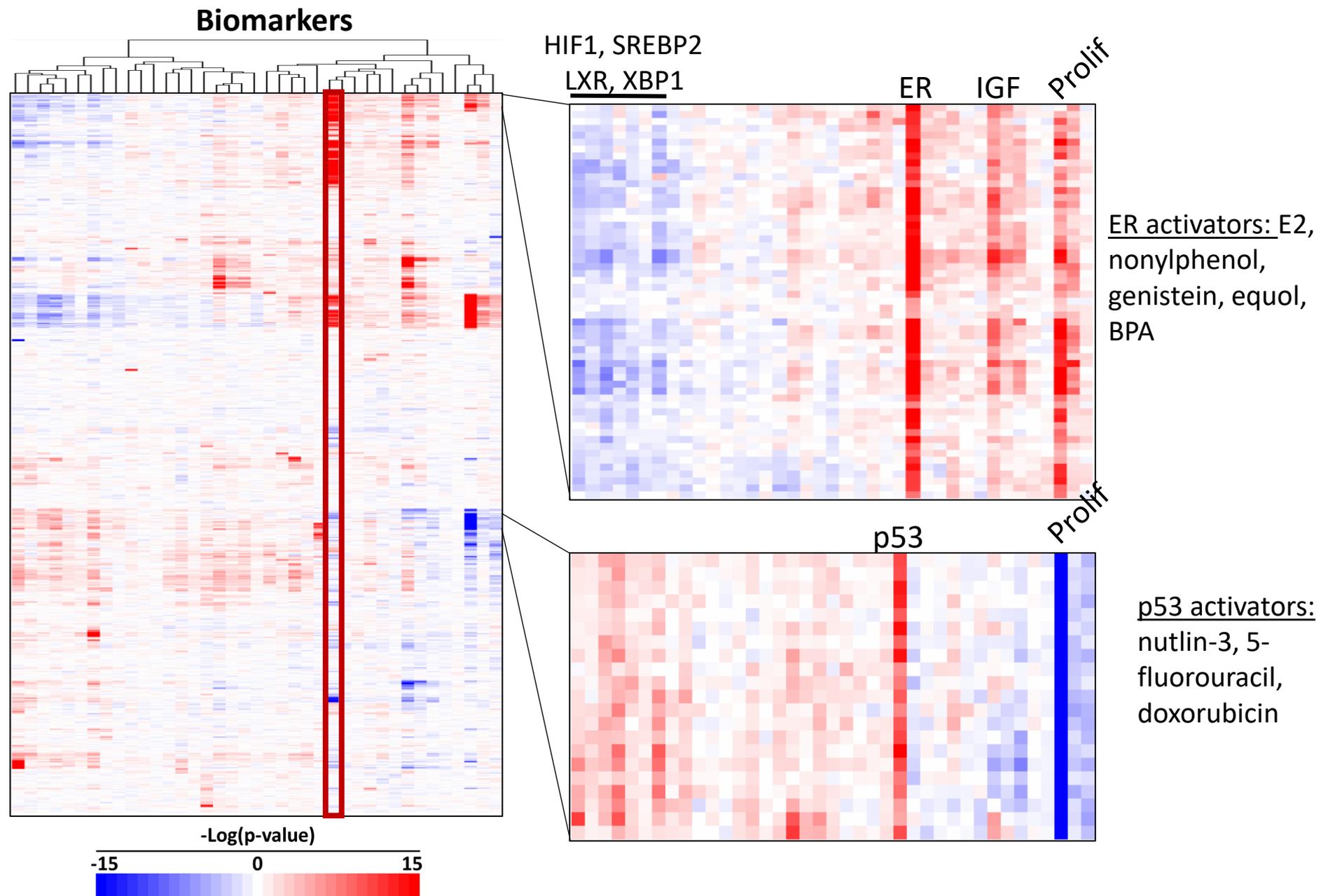
ER activation precedes cell proliferation for a large number of ER activators

- Compared 2006 chemical treatments in MCF-7 cells (1431 chemicals) to the ER and CCP biomarkers
- Grouped by time of treatment
- In general, the longer the exposure the greater the activation of the CCP biomarker



# Behavior of Biomarkers in MCF-7 cells

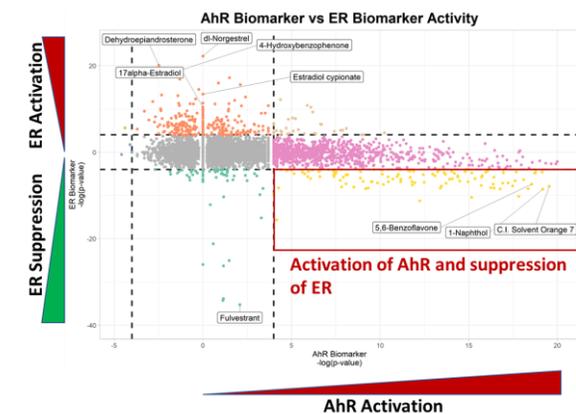
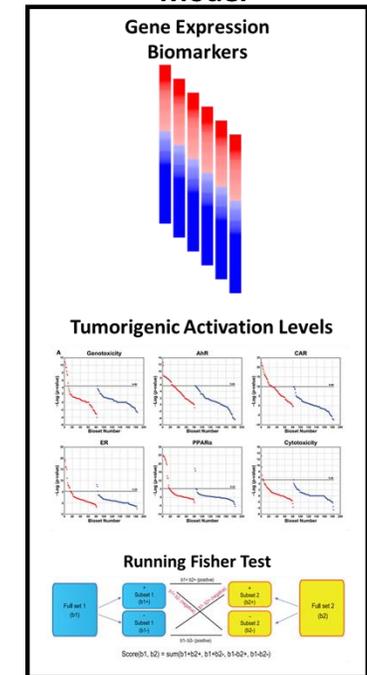
- Examined relationships between 2165 microarray comparisons in MCF-7 cells across 39 biomarkers
- Includes chemicals, various stressors, cytokines
- Two-dimensional hierarchical complete linkage clustering
- Efforts are ongoing to integrate predictions into prioritization schemes and into the AOP network



# Summary

- Moving away from the 2-year bioassay will likely require both short-term exposures in vivo and assessment of effects in vitro
- Integration of gene expression into carcinogenicity testing will be facilitated using biomarkers with known predictive accuracies and context of use
- Biomarkers for screening in rats to reduce unnecessary testing
  - Identification of mode of action
  - Identification of chemical doses that would cause cancer
- A growing number of human biomarkers are characterized for Tier 1 screening in high throughput transcript profiling
  - ER, AhR, TGx-DDI, CCP biomarkers
- Screening strategies should consider
  - Multiple cell lines (organotypic models)
  - Range of concentrations to allow response modeling
  - Range of times of exposure to capture molecular and cellular events (cell fate)

## NAM Computational Model





# ACKNOWLEDGEMENTS

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**Thanks for listening!**

**Questions?**