



JOHNS HOPKINS

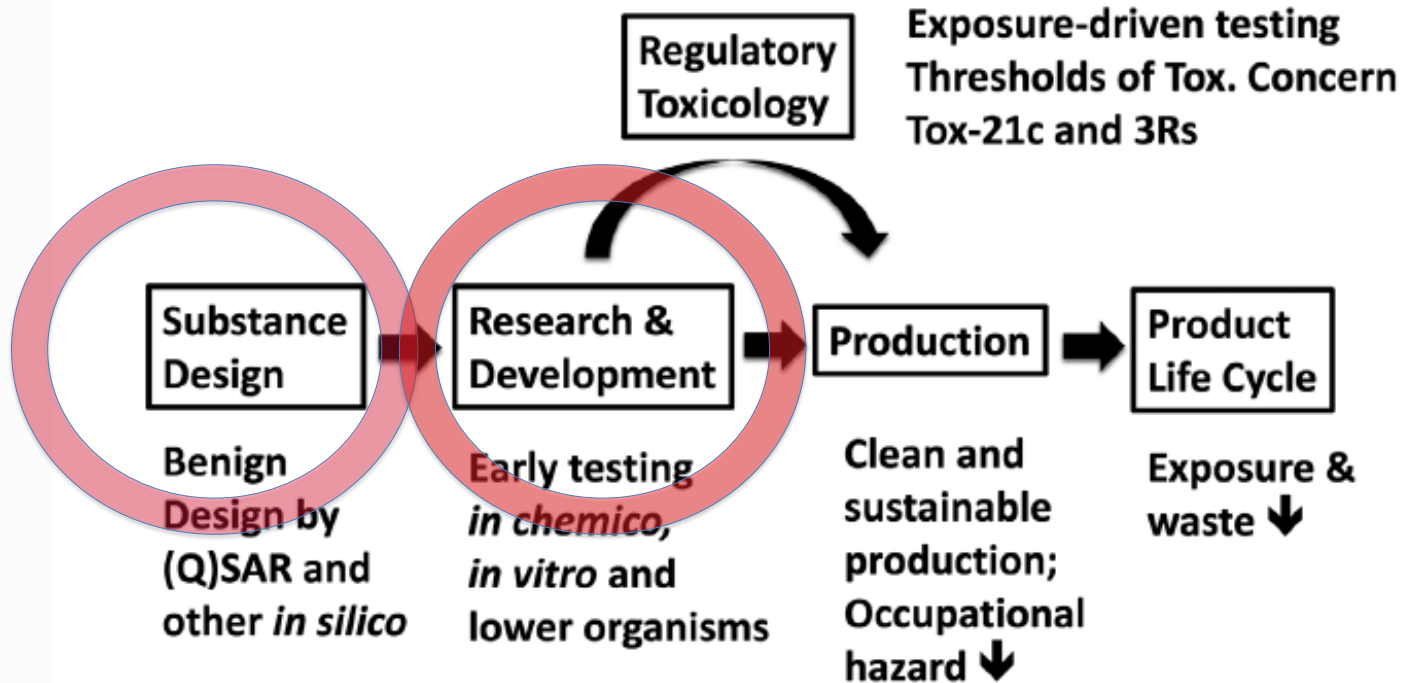
BLOOMBERG SCHOOL  
*of* PUBLIC HEALTH

# Green Toxicology: Using Big Data to Guide Design Safer Chemicals

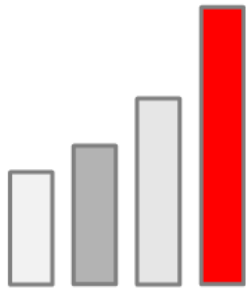
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Alexandra Maertens, Ph.D.

## Green Toxicology



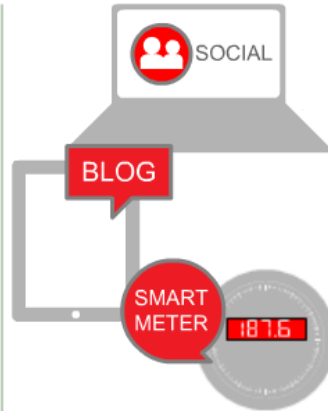
## What Makes it Big Data?



VOLUME



VELOCITY



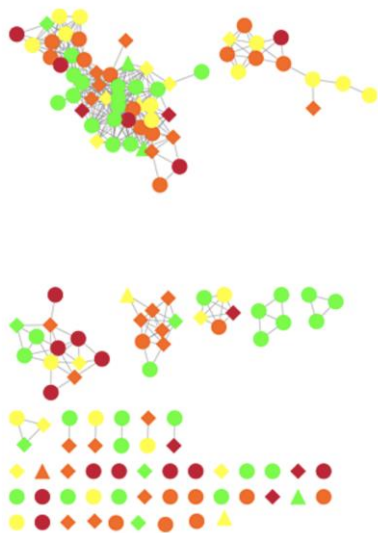
VARIETY



VALUE

# Big Data. . . for a Toxicologist

2015



- More information is not always better; model improved when we used feature elimination
- Model improved when dose-response information included
- Variable ranking indicated in vitro assays provided information

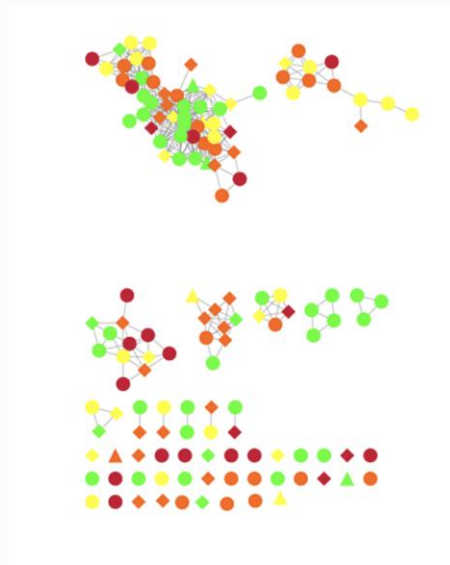
**Probabilistic hazard assessment for skin sensitization potency by dose-response modeling using feature elimination instead of quantitative structure-activity relationships**

Thomas Luechtefeld<sup>a†</sup>, Alexandra Maertens<sup>a†</sup>, James M. McKim<sup>b</sup>,  
Thomas Hartung<sup>b,c\*</sup>, Andre Kleensang<sup>a</sup> and Vanessa Sá-Rocha<sup>a,d</sup>

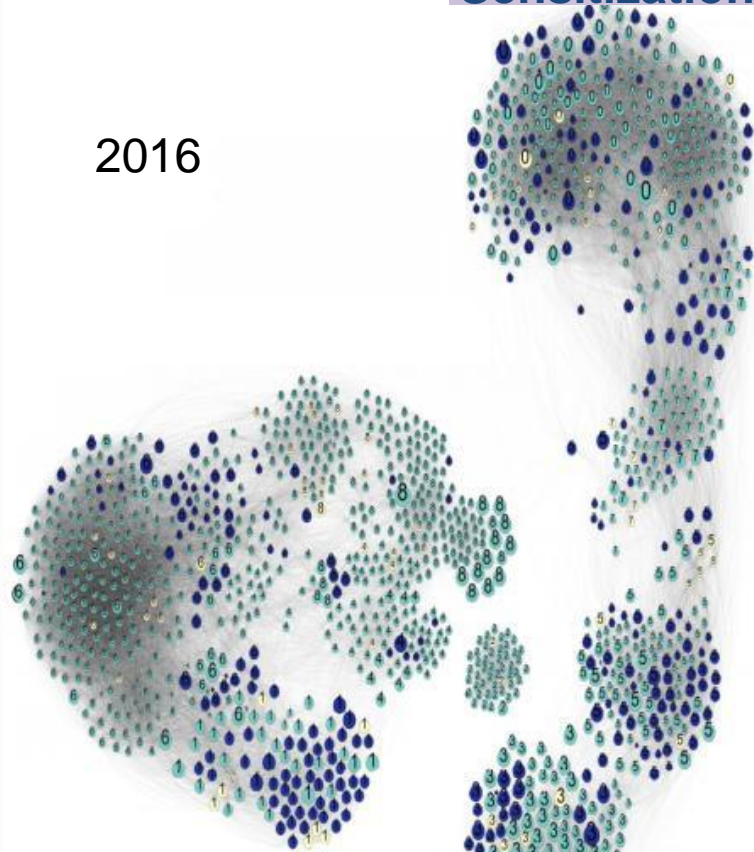
# Big Data. . . For a Toxicologist

**REACH** Skin  
Sensitization

2015



2016



## *Skin sensitization: Simple classification by nearest neighbor*

Min. Similarity	Chemicals	Accuracy
0.95	525	0.92
0.9	1189	0.85
0.85	1738	0.84
0.75	2288	0.80

**Accuracies better than different animal TG against each other**

**From:** Luechtefeld et al. Predicting Skin Sensitization with REACH Dataset

# Big Data: Better Data

**Tab. 1: Classification agreement on chemicals with at least two sensitization studies in REACH dossiers from 2008-2014**  
Studies found by searching for all studies with *studytype* = Buehler, GPMT, Patch-Test or LLNA.

	Buehler	GPMT	Patch-test	LLNA
Buehler	95.1% (344 chem.)	91.8% (364 chem.)	87.8% (58 chem.)	76.8% (212 chem.)
GPMT		93% (624 chem.)	90.5% (107 chem.)	77.4% (403 chem.)
Patch-test			92.1% (24 chem.)	78.3% (40 chem.)
LLNA				88.5% (296 chem.)

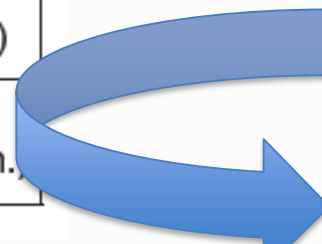
Guinea  
pig

77%



Mouse

89%



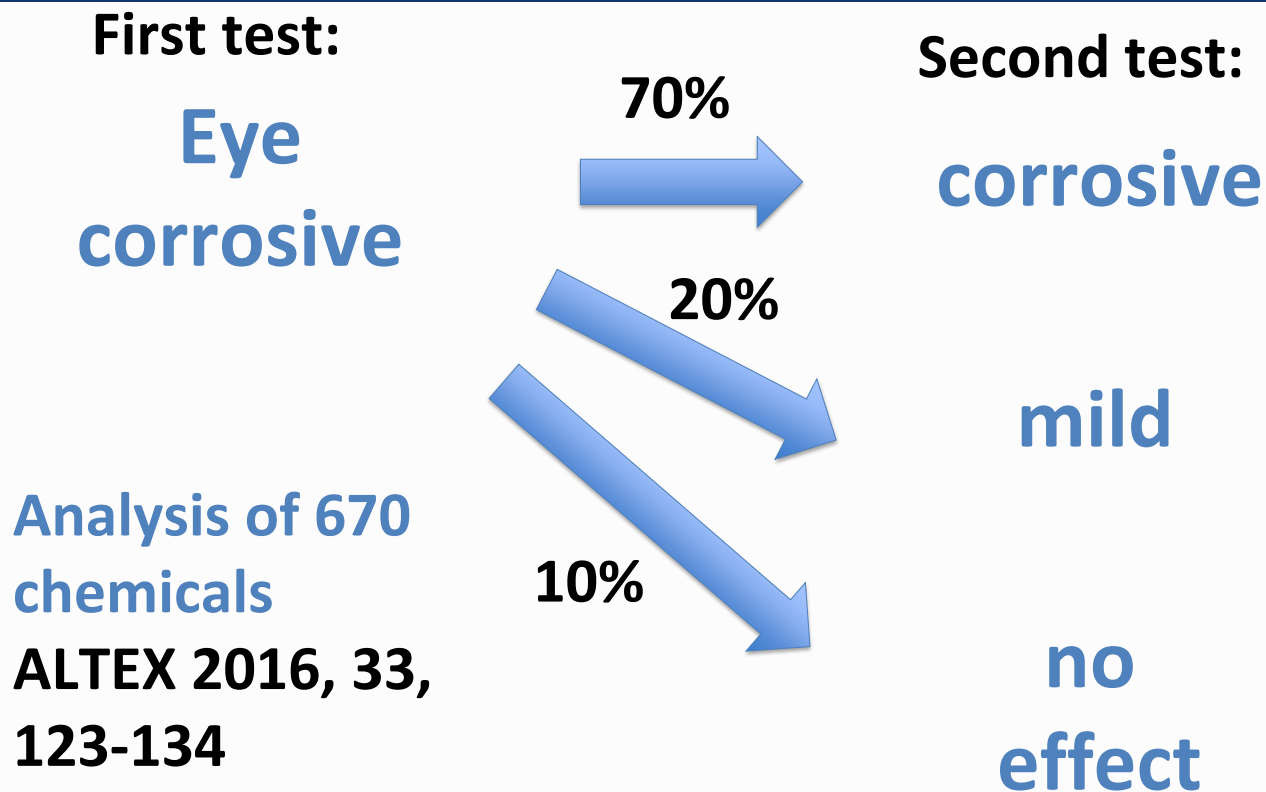
# Big Data: Public Data is Useful Data

## *Repetitions of Draize rabbit eye test (TG 405)*

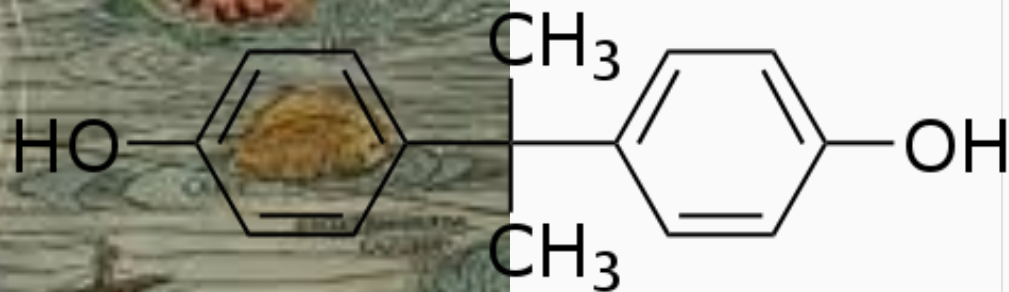
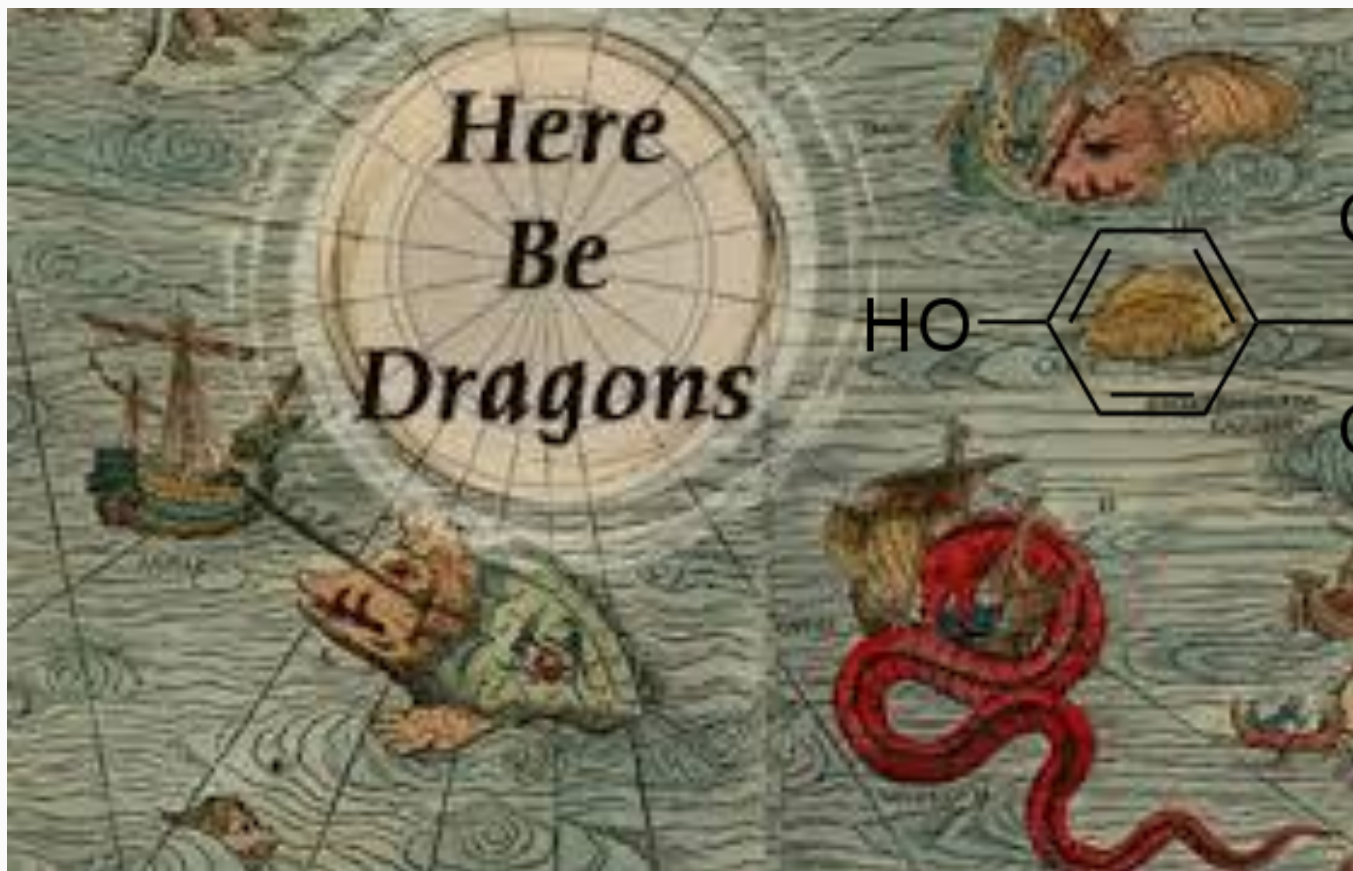
<b>Repeats</b>	<b>Number substances</b>	<b>Example ECNumber</b>
90	2	613-683-0,295-445-2
45	69	940-595-2,295-431-6
18	1	931-203-0
15	2	700-762-0,692-840-5
13	38	934-268-3,931-515-7
12	2	918-317-6,500-513-4
11	2	931-700-2,226-109-5
10	2	232-395-2,939-581-9
9	1	267-291-6
8	27	940-730-5,940-728-4
7	32	940-727-9,940-726-3
6	75	931-745-8,300-226-2
5	56	939-578-2,939-575-6
4	135	939-693-8,939-621-5
3	254	939-715-6,939-688-0
2	593	208-778-5,941-224-7
1	2388	293-029-5,273-224-1



# Big Data: Public Data is Useful Data



# Mapping the Human Toxome



## Big Data. . . Of the Wrong Kind

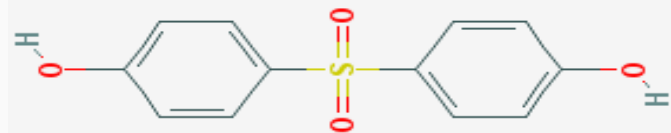
- BPA has been subjected to multiple guideline studies in different species/strains
- Pubmed returned over 10,800 abstracts for BPA
- Comparative Toxigenomics Database shows BPA effects almost 2,000 genes when restricted to humans
- HSDB has over 79 studies laboratory animal studies



THE LD<sub>50</sub> OF TOXICITY DATA IS  
2 KILOGRAMS PER KILOGRAM.

Source: XKCD

Aha! BPA binds to the Estrogen Receptor!  
If I add a sulphonyl group, it will no longer bind!



Bisphenol Sulfate

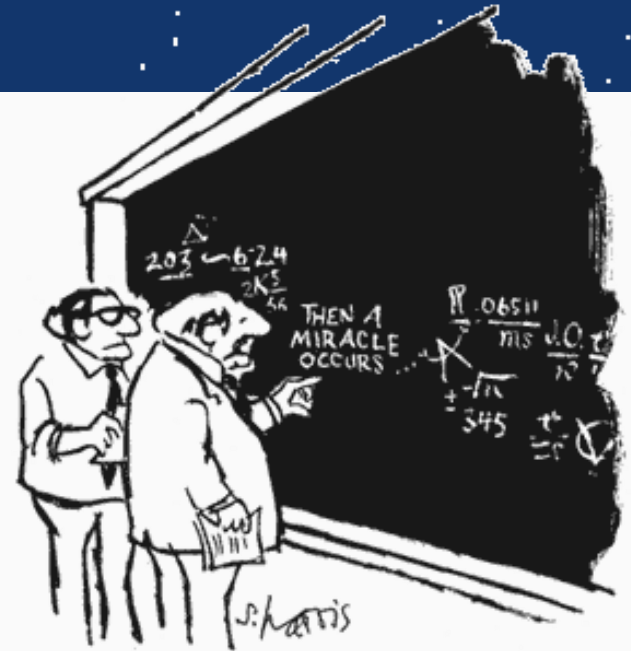
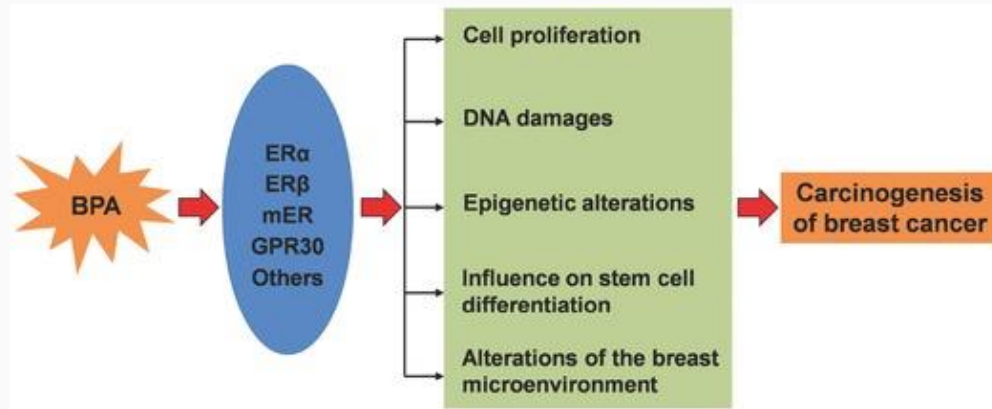
## Unfortunately. . .

- turns out that BPA binds canonical Estrogen receptors but bind Estrogen-Related Receptor Gamma (ERR-Gamma) with higher affinity
- BPS binds ER- $\gamma$  with higher affinity than BPA
- Unlikely to be detected by animal testing
- 1-10 Million tons of BPS are produced per year



Out of the frying pan and into the fire

# BPA - Pathways and Adverse Outcomes



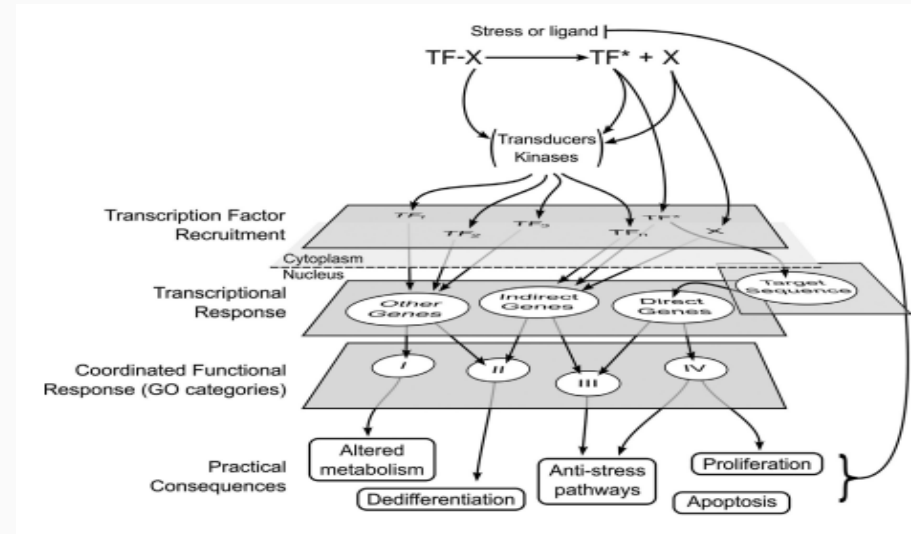
"I THINK YOU SHOULD BE MORE EXPLICIT HERE IN STEP TWO."

From: Low-Dose Bisphenol A Exposure: A Seemingly Instigating Carcinogenic Effect on Breast Cancer

[Zhe Wang](#), 1, 2, \* [Huiyu Liu](#), 3, \* and [Sijin Liu](#) 1, \*

# Mapping the Human Toxome

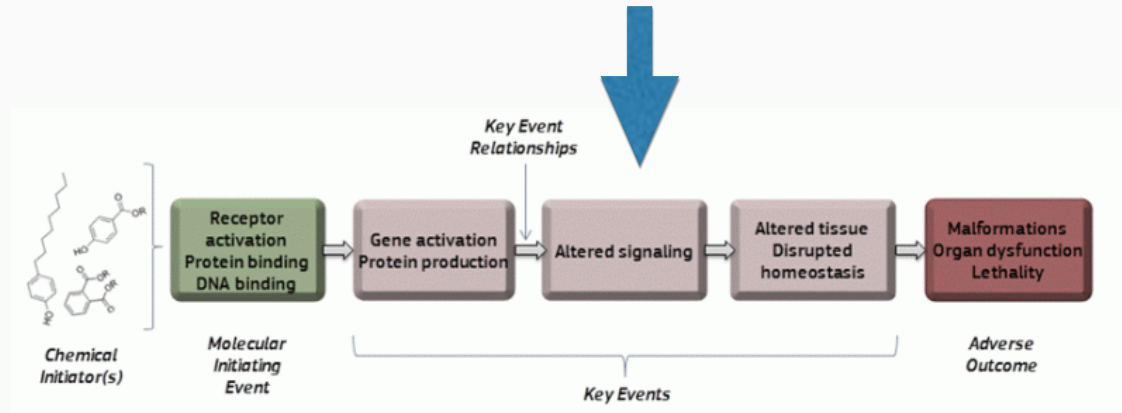
- Pathway of Toxicity is a **molecular definition** of the **cellular processes** shown to mediate **adverse outcomes** of toxicants.
- “Black box” animal models or simple molecular models towards understanding toxicity in it’s full complexity
- A PoT helps to bring toxicology into the “systems biology”



From: Andersen, MA 2013

# Pathway of Toxicity

A **Pathway of Toxicity (PoT)** focuses on the molecular events within a cell - the networks of genes, proteins, or metabolites that are altered as a result of the molecular initiating event.






# Can we use *in vitro* data for a better understanding of molecular mechanisms?

- We used WGCNA and IDEA to extract a genetic regulatory network and possible transcription factors from cells treated with a variety of estrogen and BPA
- Goal: to understand which receptors and transcription factors are driving the biology

Correlation Based Approach  
using a “Guilt by Association  
Approach” to generate  
interaction partners from a  
microarray dataset

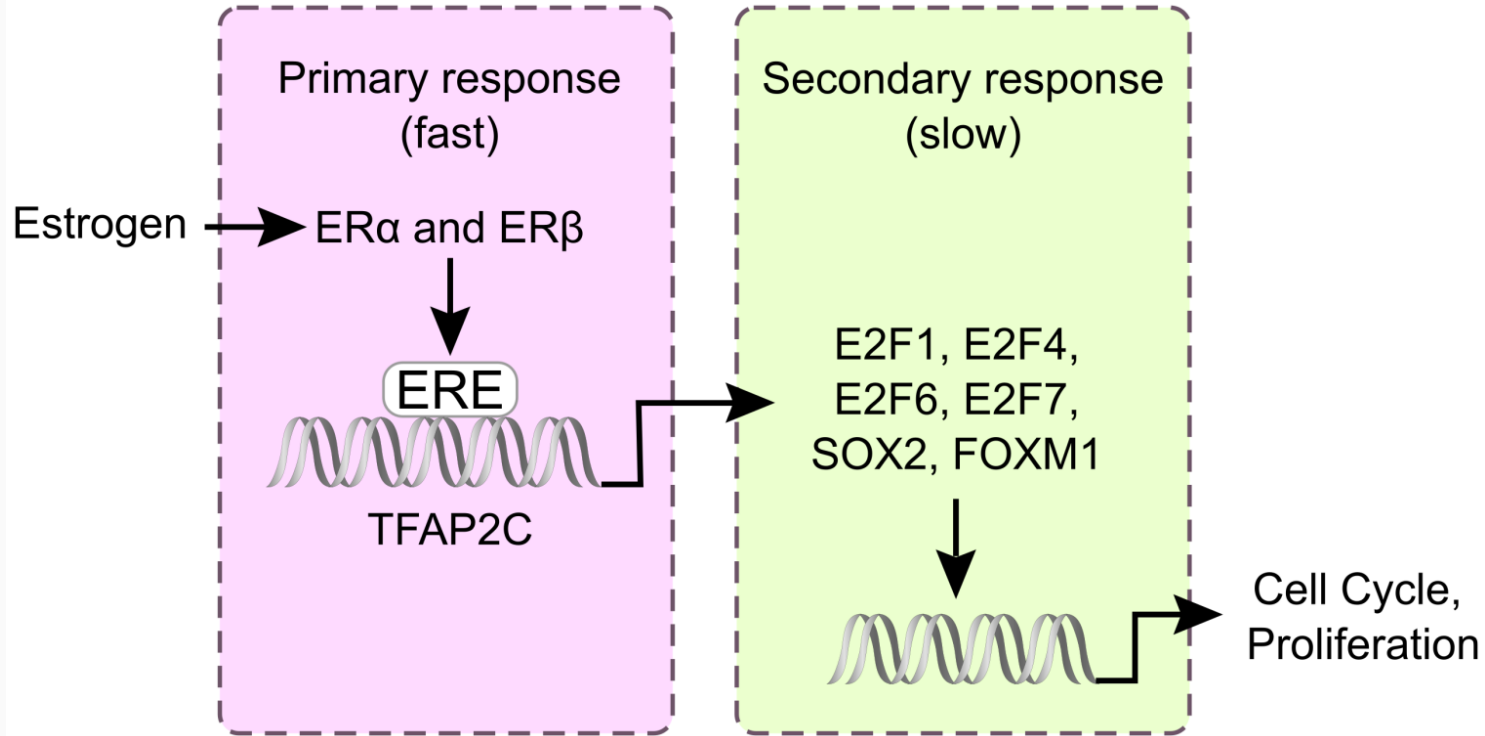
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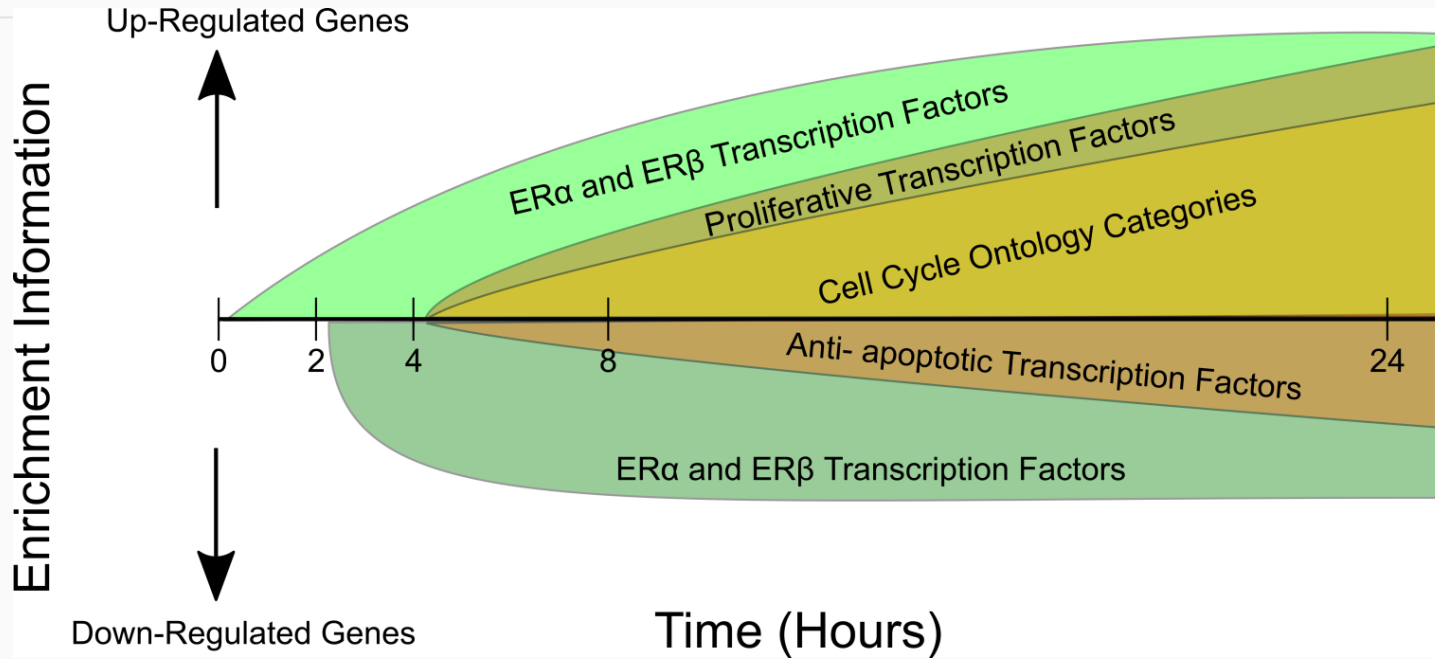


Information Dependent  
Enrichment Analysis  
Uses Rank Order of Gene  
Expression and a Permutation  
Test to Establish Candidate  
Transcription Factors

# ESTROGEN

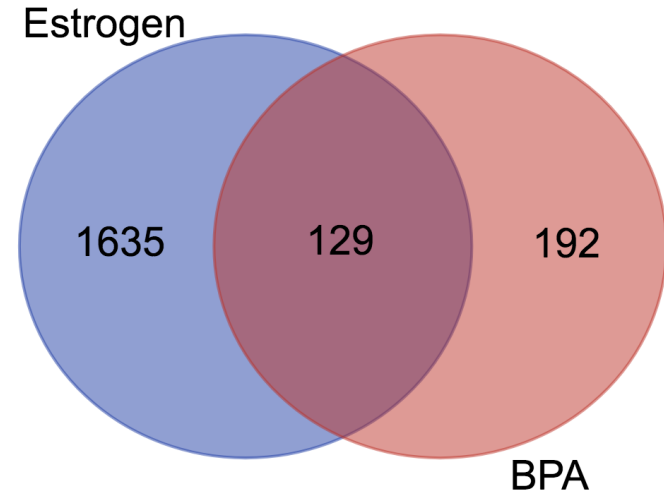


# ESTROGEN

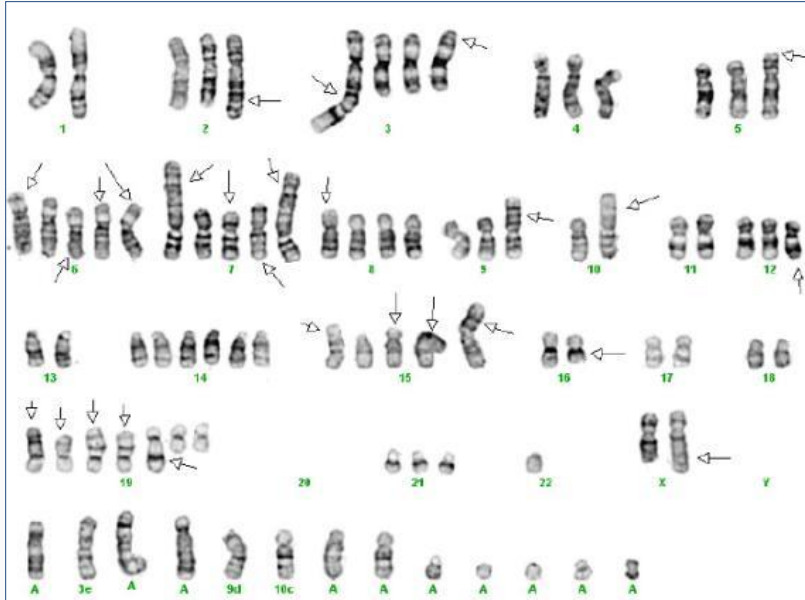


# WGCNA - BPA

- Weak over-representation of canonical estrogen-receptor binding sites; several novel transcription factor unique to BPA
- Just under 10 percent of the genes in the data set were “open reading frames” - meaning they had no official gene name, and no annotations
- Several identified transcription factors (such as FIZ1) are poorly characterized - less than 8 PMIDs



# Use *In Vitro* Data Cautiously



**Karyotyping**

# MCF-7 Does It Really Capture Human Biology?

## *Extent of deviations from normal genome*

<b>Classification</b>	<b>Kilobases</b>	<b>Percentage of genome</b>
Losses	4587603	51.2%
Deletions	667374	7.5%
Amplifications	26904	0.3%
Gains	2587093	28.9%
Normal	871166	9.7%
Centromeres	217339	2.4%
<b>Total Abberations</b>	<b>7868974</b>	<b>87.8%</b>
All Entries	8957479	

**SurePrint G3 ISCA CGH+SNP Microarray Kit, 4x180K**

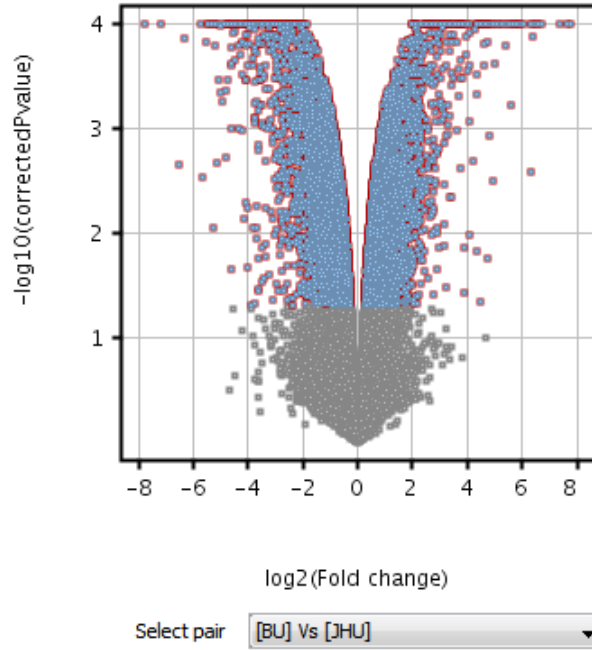
**115234 CGH features.2440 CGH replicate probes, 59647 SNP features**

**reference mapping: caucasian female human reference DNA**

**Kleensang et al., Nature Sci Rep, 2016**

# Use *In Vitro* Data Cautiously

## Comparison of MCF-7 in two laboratories



negative controls, 4h, gene level,  $n = 3$  / group

Same batch from ATCC

Method transfer

Transcriptomics



“The Universe is under no obligation to make sense to you.”

-Neil DeGrasse Tyson



- The mechanism of toxicity is under no obligation to be simple for you -
  - Not necessarily mediated by a single receptor
  - Receptors may have different effects in different cells/organs
  - Dose-response may vary enormously depending on context - oxidative and metabolic state, epigenetic background, etc. . . . and your animal or *in vitro* model may be misleading
  - We need to stop “looking under the lampshade for the key, since that is where the light is” - we have to acknowledge how much of cell biology is unknown

# Mapping Mechanisms of Toxicology is Just a Big Data Problem

- For some endpoints, big data is already paying off - we have good models to guide molecular design and test quickly and without animals - skin sensitization, mutagenicity, etc.
- For more complicated endpoints, the map is still incomplete, but. . .
- The number of ways a chemical can interact with a cell *is finite* - it is “mappable”
- *In vitro* technologies are key, but we need to vastly expand our data sets and improve the way we approach our data sets, and respect the limits of our understanding