

# **Computational Methods to Address Challenges in Chemical Risk Assessment**

**Bio-Seminar in the Department of Electrical & Computer Engineering at Texas A&M  
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# Outline

- Overview of chemical risk assessment
- Examples of key challenges and role of computational methods
  - Risk from complex and varied exposures
  - Addressing population variability
  - Quantifying risk and uncertainty
- Risk assessment as translational science

# Scientific Components of Risk Assessment

## Source-to-Outcome Continuum

Source/stressor formation

Fate & Transport

Environmental concentrations

Exposure

External doses

Toxicokinetics

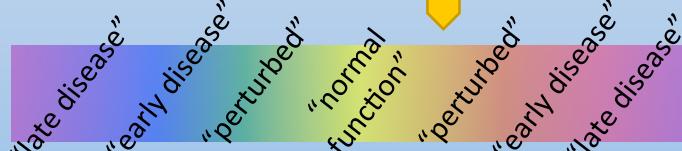
Internal concentrations

Toxicodynamics

Biological response measurements

Systems dynamics

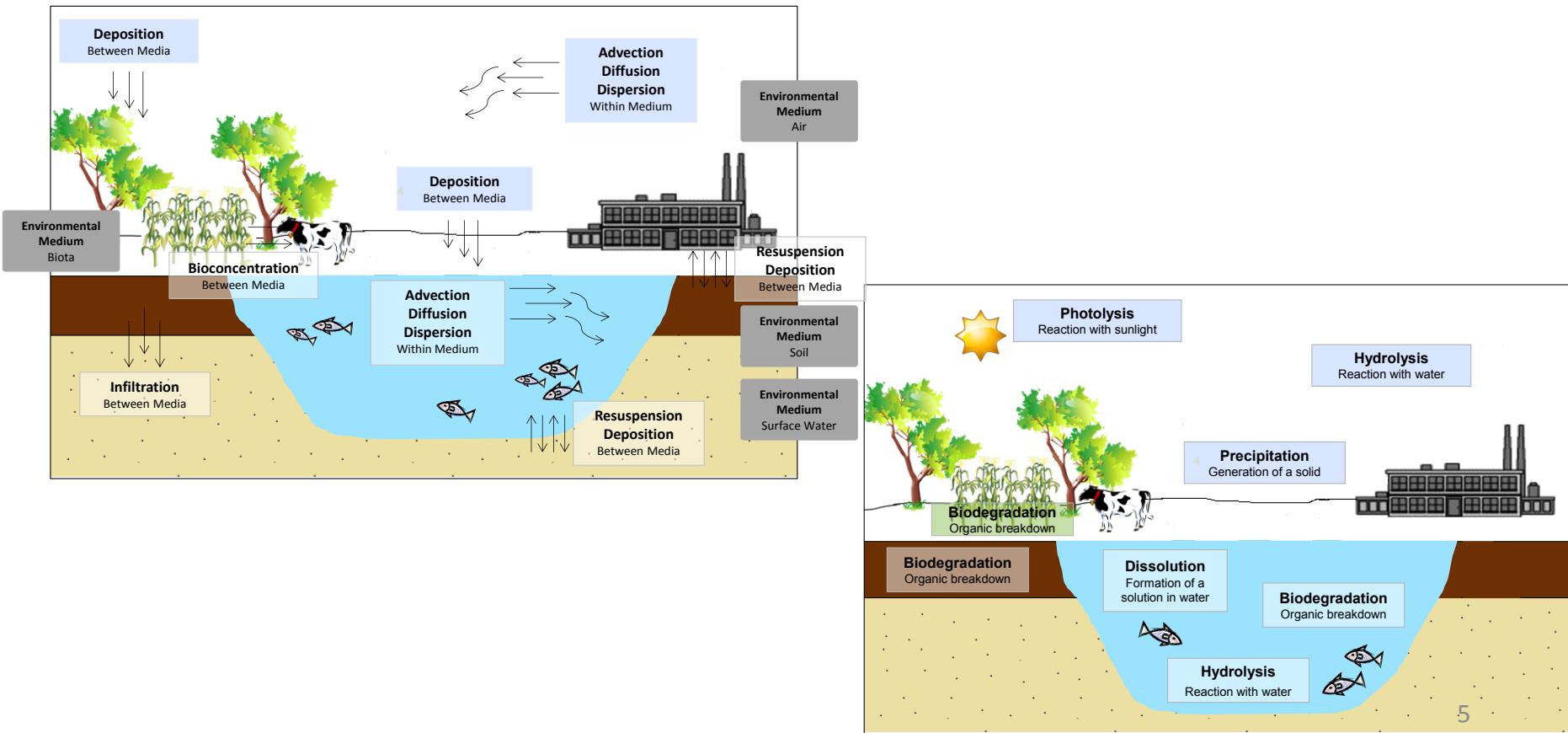
Physiological/health status



## Exposure Assessment



# Transport and transformation of chemicals in the environment



# Exposure modeling

## Storm surge from Hurricane

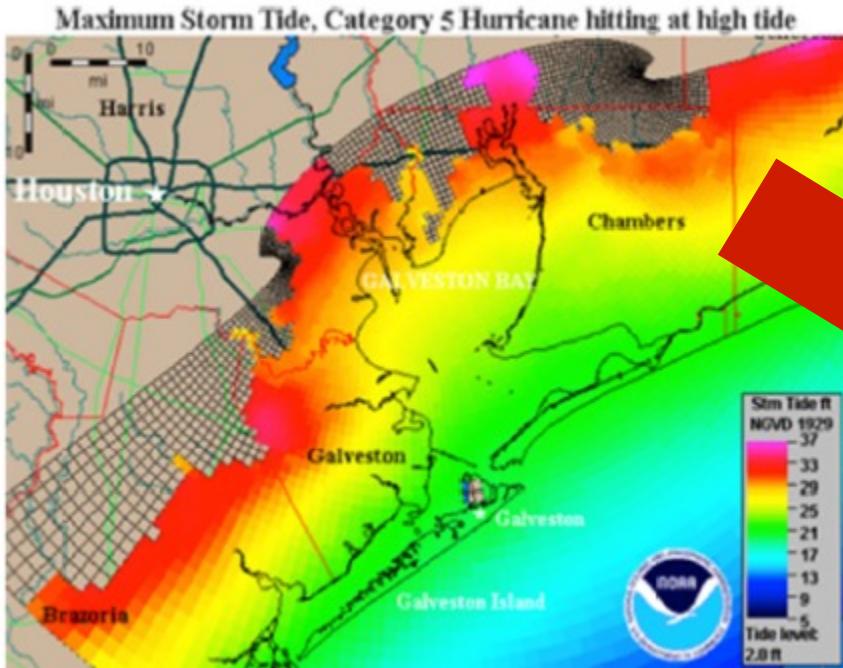
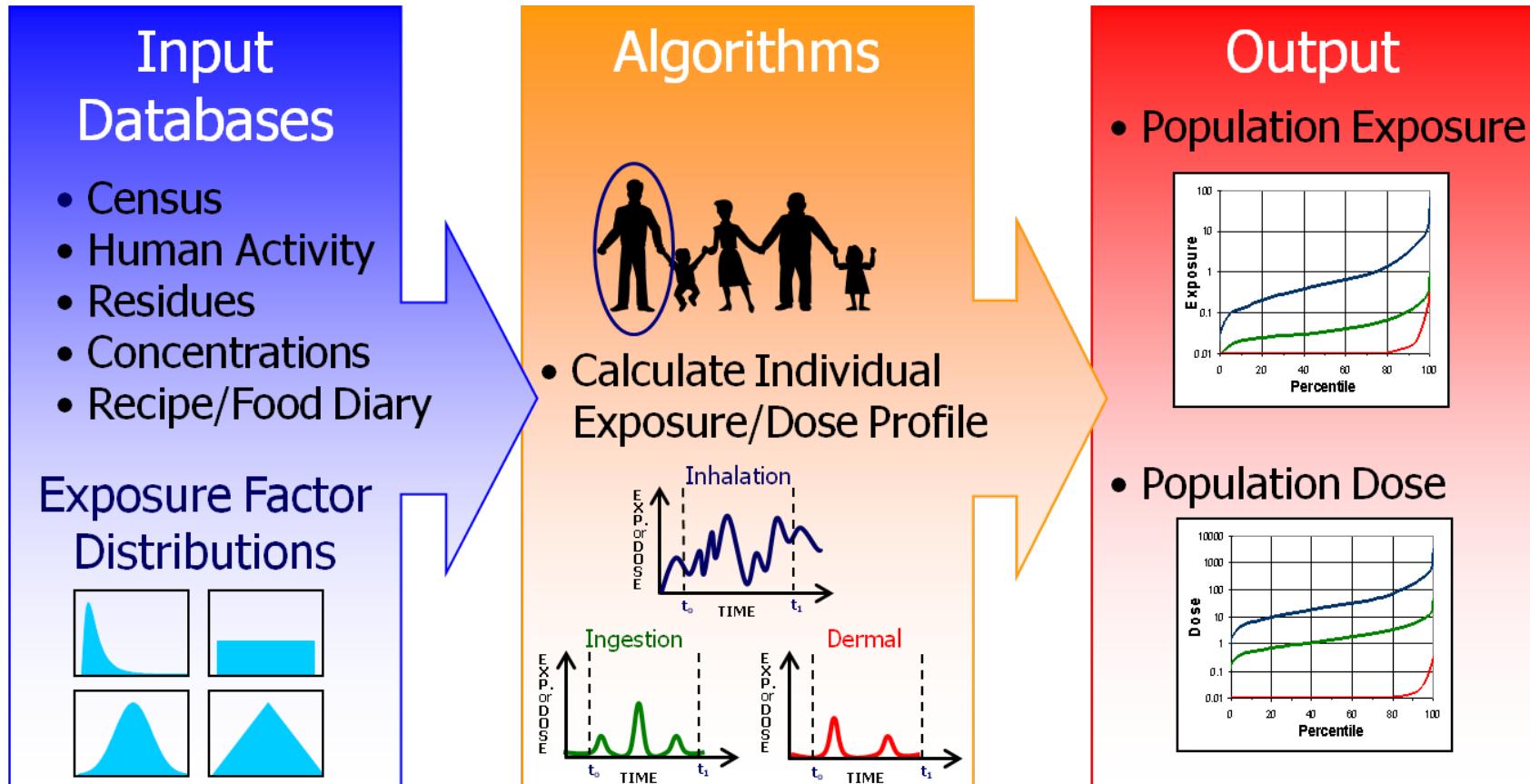


Figure 8. Coastal inundation in a Category 5 hurricane.



Figure 4. Sediments in residential New Orleans post-Katrina (Photo: Geoff Plumlee).

# Estimating Human Exposure in the Population



Source: SAP SHEDS Overview, 7/14/2010

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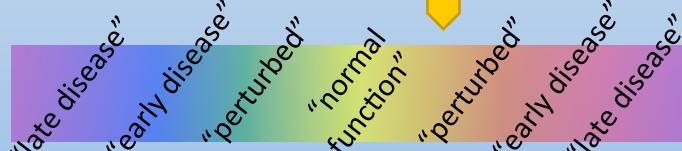
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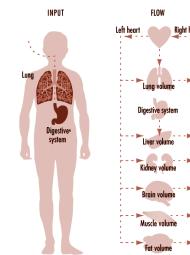
Physiological/health status



## Exposure Assessment



## Pharmacokinetics/Toxicokinetics

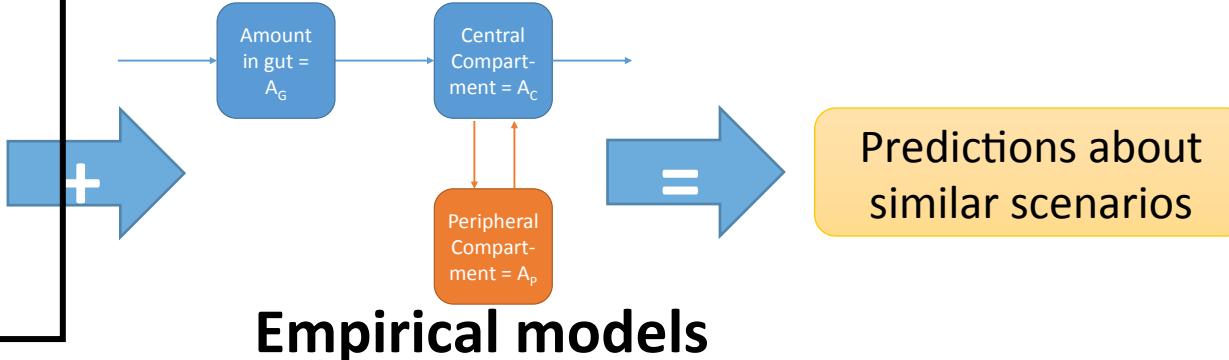
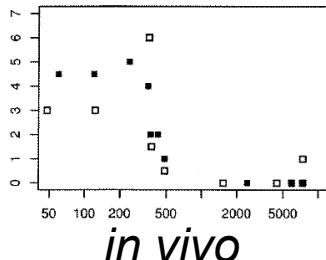


# Toxicokinetics = “Fate and transport within the body”

- Exposure alone is not sufficient to elicit toxicity
- Interaction between an exogenous agent and a biological target
  - What is the agent or toxic moiety?
  - How does it get to the biological target?
  - How much of the agent gets there?
  - How long does it stay there?
- **Toxicokinetics** is the study of the movement of chemicals in and out of the body (“what the body does to the chemical”)
  - Absorption
  - Distribution
  - Metabolism
  - Excretion

# For pharmaceuticals – mostly use simple empirical models

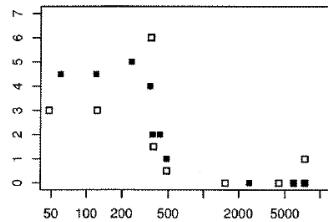
Chemical-specific data



**Empirical models**

# More complex models trade off simplicity for predictive power

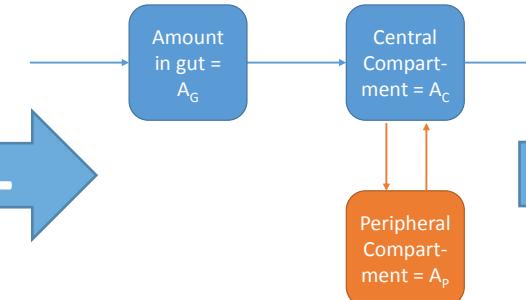
## Chemical-specific data



*in vivo*



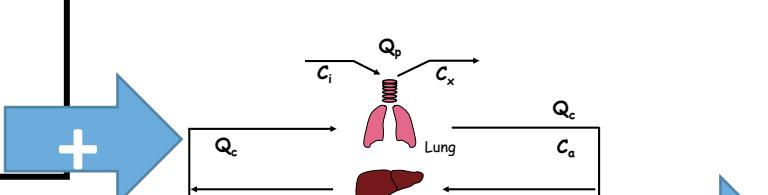
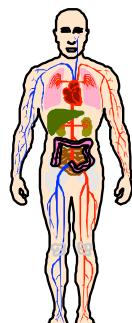
*in vitro*



Predictions about similar scenarios

## Empirical models (simple & quick)

## Physiological Data



Predictions about scenarios with different:

- Exposure routes, durations, levels, patterns
- Species
- Individuals

## PBPK models (complicated & time-consuming)

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Systems dynamics

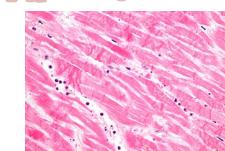
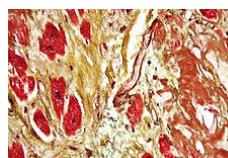
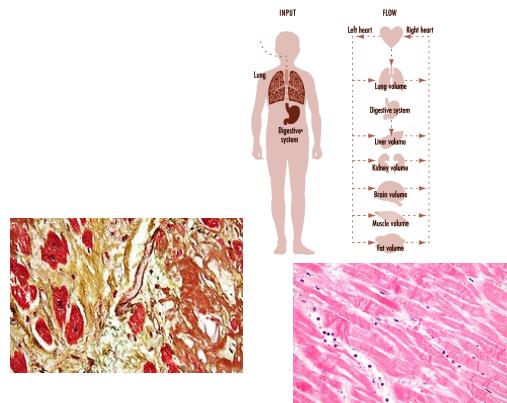
Physiological/health status

“late disease”  
“early disease”  
“perturbed”  
“normal function”  
“perturbed”  
“early disease”  
“late disease”

## Exposure Assessment



## Hazard Identification and Pharmacokinetics/Toxicokinetics/Dose Response Assessment



# Hazard Identification

- Determination of whether a particular chemical is or is not ***causally linked*** to particular health effects
  - Increased ***incidence***
  - Increased ***severity***

**What adverse effects have been observed or are anticipated?**

- Human data
- Laboratory animal data
- In vitro data
- Physical/chemical/molecular property data

**For each adverse effect, what is the evidence that the agent can cause it in humans?**

- Availability of data  
(absence of evidence ≠ evidence of absence)
- Consistency within and across the different types of data.
- Biological plausibility / mechanistic basis

**Recent emphasis has been on applying systematic review methods to evaluate evidence of causality (not discussed further today)**

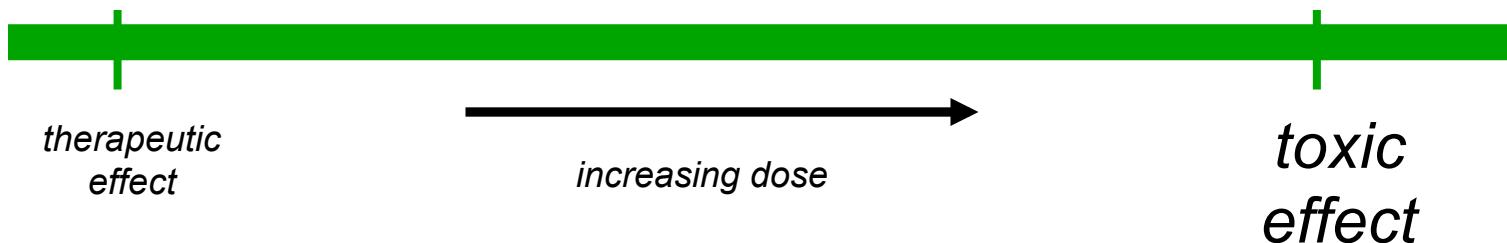
# Dose-Response – Many still ascribe to the principles of Paracelsus...

Paracelsus

(Phillippus Aureolus Theophrastus Bombastus von Hohenheim)

1493-1541

Known as the ‘father of toxicology’. The saying “Dosis facit venenum” (The dose makes the poison) is attributed to him. His actual quote translates “All things are poisons, for there is nothing without poisonous qualities...it is only the dose which makes a thing poison.”

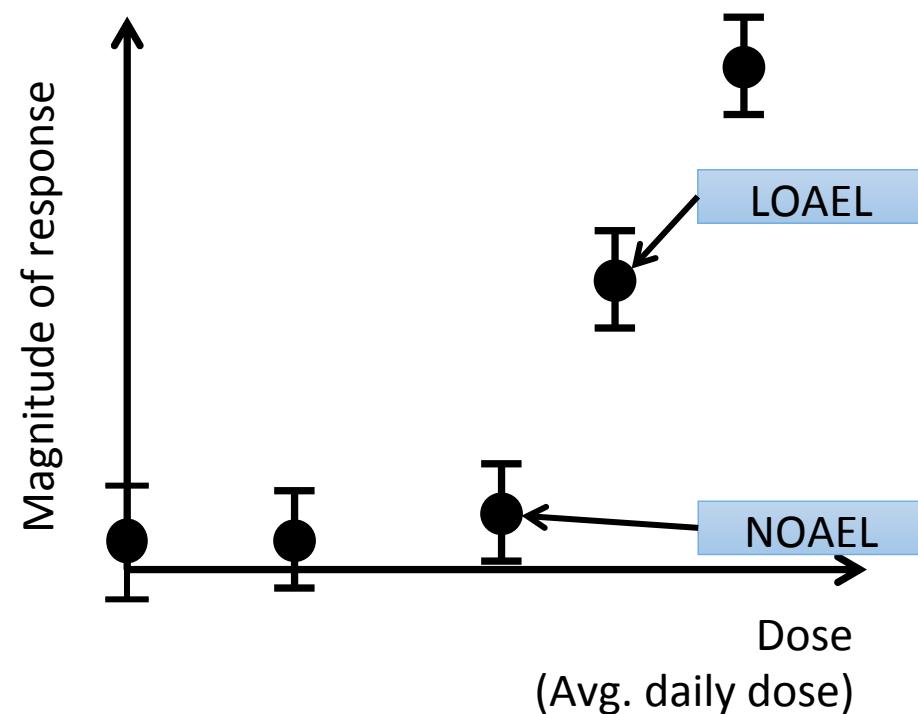


# Traditional interpretation: Existence of a “threshold” below which there are no effects

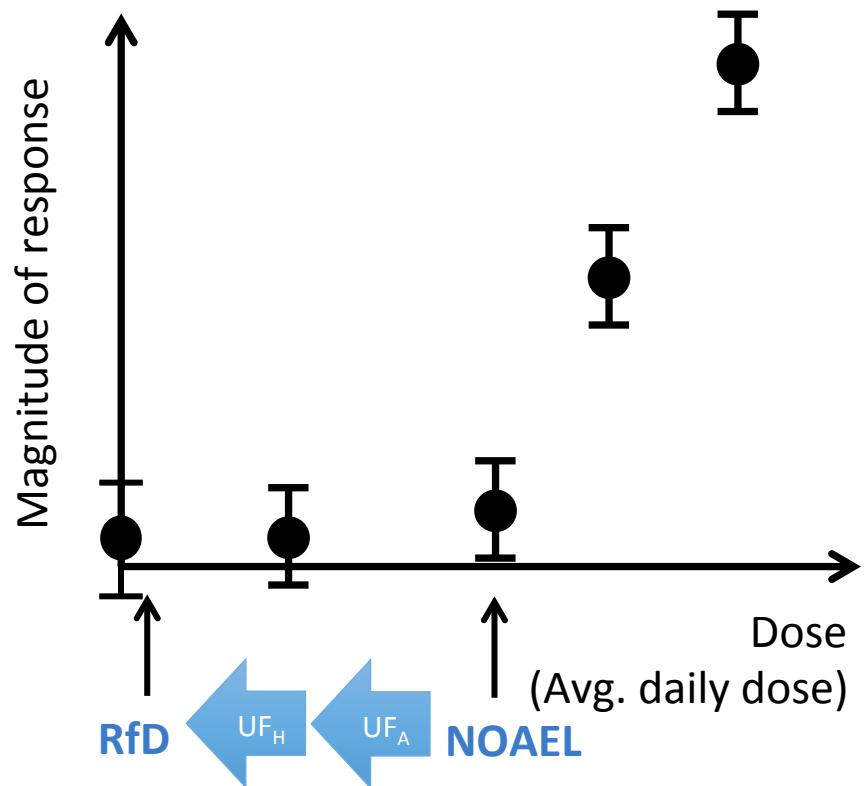
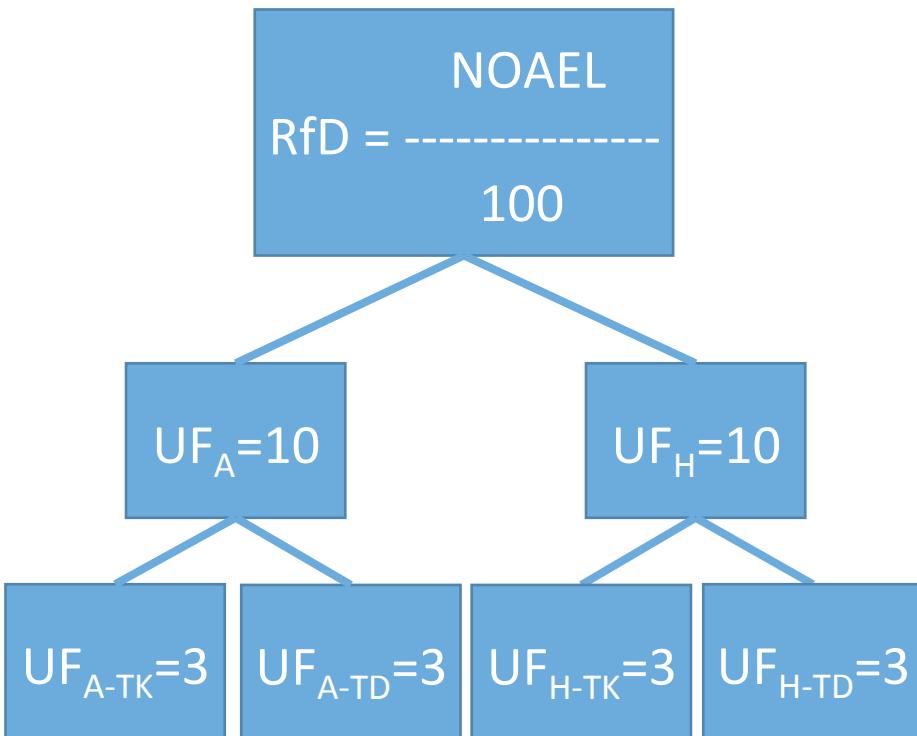
- **NOAEL:** Greatest concentration or amount of a substance, found by experiment or observation, that causes no adverse alteration ...of the target organism distinguishable from those observed in normal (control) organisms of the same species and strain under the same defined conditions of exposure.\*

- Commonly viewed (**incorrectly**) as an experimental dose threshold.

\*WHO definition



# Implementation: “Safe Human Dose” Established by Use of “Uncertainty” or “Safety” Factors



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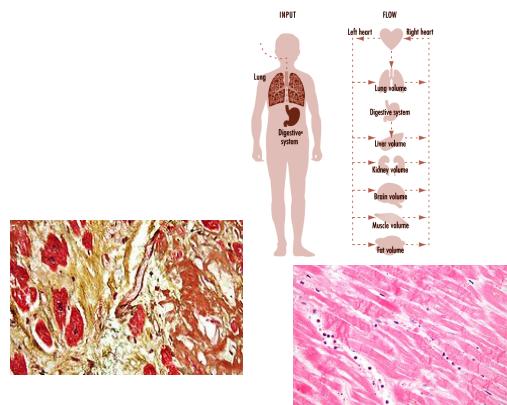
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“normal function”  
“perturbed”  
“early disease”  
“late disease”

## Exposure Assessment

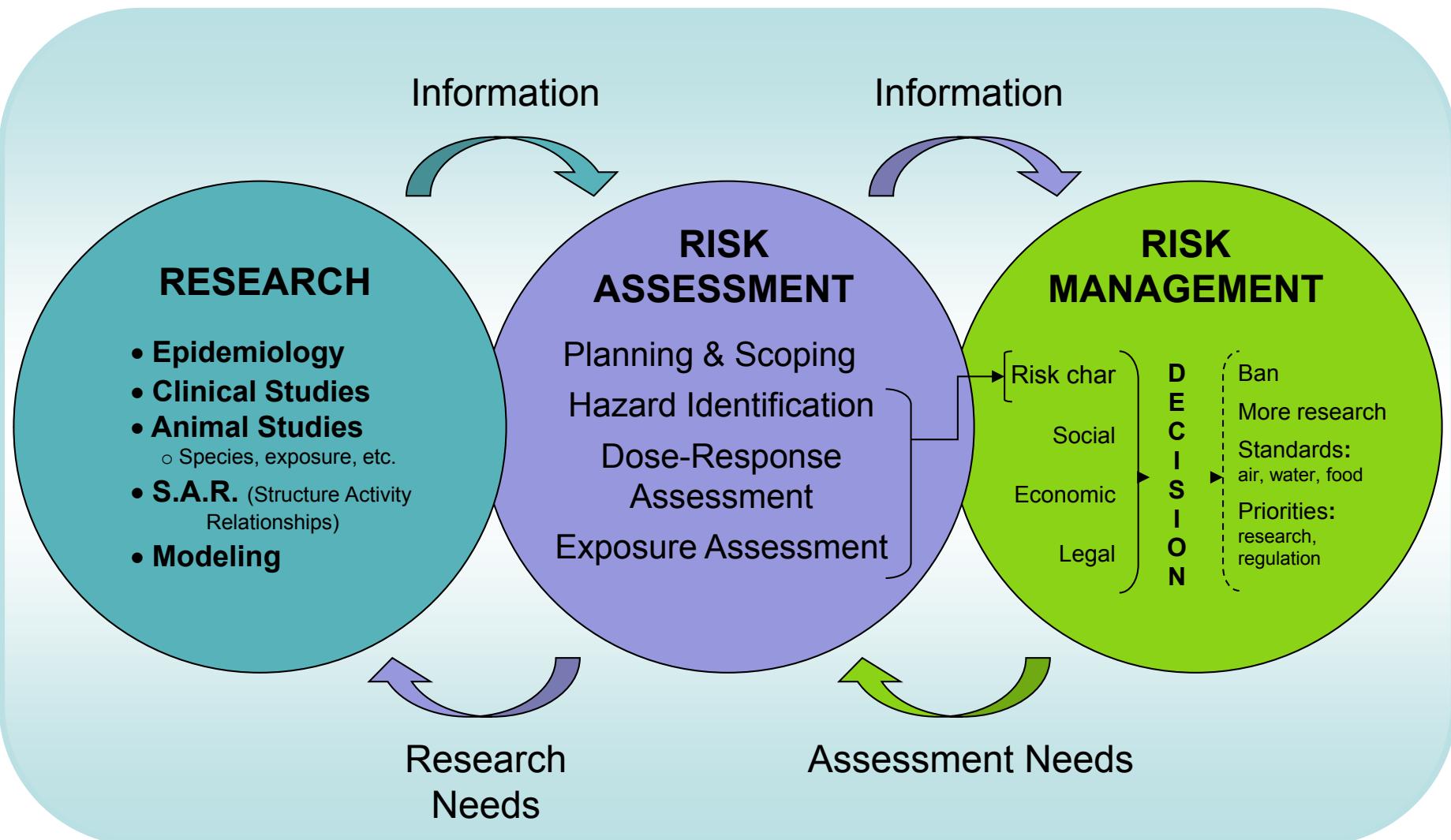


## Hazard Identification and Dose-Response Assessment



Risk  
Characterization

# Risk Assessment in the Context of Research & Decision-Making



# Multi- and trans-disciplinary nature of risk assessment

- Requires data and models from multiple scientific disciplines.
- Requires methods and approaches for integrating diverse information to draw scientific conclusions about risk.
- Requires consideration of not only scientific, but also social, economic, and legal factors in order to inform decisions about managing risk.

# Examples of challenges and computational methods

- Complex and varied exposures with incomplete data on chemical risks
- Incomplete understanding of population variability in susceptibility to chemical risks
- Inadequate quantification of chemicals risk and its uncertainties

# Example Challenge: Exposure assessment for environmental mixtures

## Source-to-Outcome Continuum

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## Storm surge from Hurricane

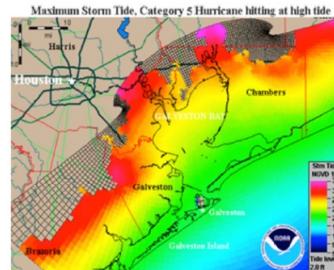


Figure 8. Coastal inundation in a Category 5 hurricane.

## Sediment deposition



Figure 4. Sediments in residential New Orleans post-Katrina (Photo: Geoff Plumlee).

Usual Approach is to perform “targeted” chemical analyses:

“Known known” contaminants

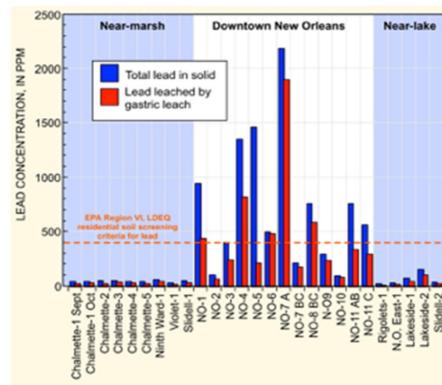


Figure 3. Lead in New Orleans soils after Katrina (1).

“Known unknown” contaminants

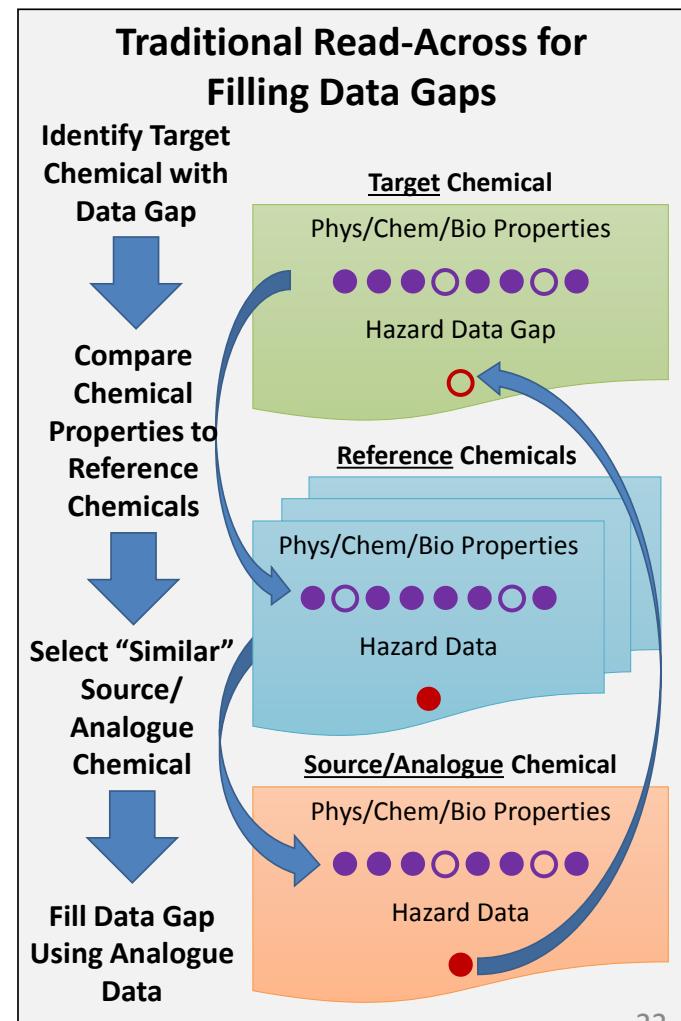
Table 2: Number of compounds per class of Contaminant and their detection limits at GERG Laboratories

Analyte Class Contaminant	Individual Component	Detection Limit
PAH	53	1 ng/g
PCB	209	1 ng/g
Dioxin and Furans	17	0.5 to 2.5 pg/g
Organochlorine Pesticides	20	1 ng/g
Organophosphate Pesticides	4	2 ng/g
Plasticizers	7	1 ug/g
Phenols	3	1 ug/g
Chlorophenols	8	1 ug/g
Nitrophenols	4	1 ug/g
Chlorinated Hydrocarbons	20	1 ug/g
Lead	1	0.4 ug/g
Mercury	1	0.01 ug/g
Cadmium	1	0.1 ug/g
Chromium	1	5 ug/g
Cobalt	1	10 ug/g
Copper	1	0.4 ug/g
Arsenic	1	0.1 ug/g
Zinc	1	0.4 ug/g
Selenium	1	0.10 ug/g
Nickel	1	0.10 ug/g

- How do you prioritize “known unknowns” given limited time and resources?
- What about “unknown unknown” contaminants? <sup>21</sup>

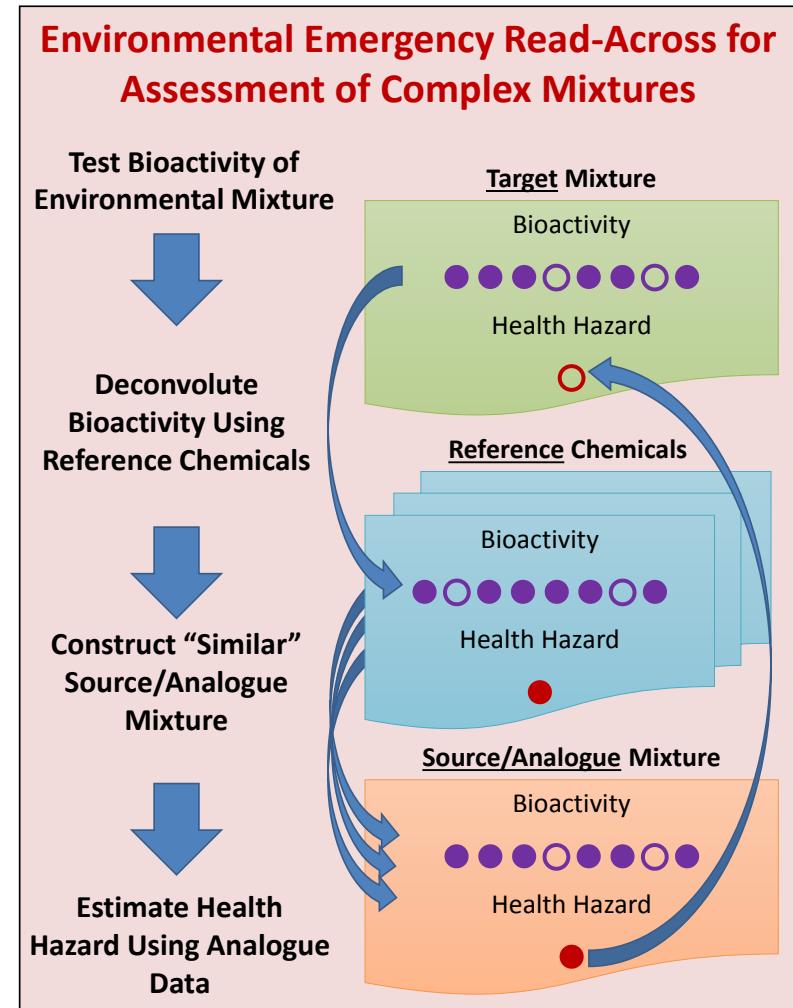
# Proposed solution based on the principle of “read-across.”

- Hypothesis that data gaps can be filled by “analogy”
- Requires:
  - Data and method to determine “similarity”
  - A “reference” set from which to find “analogue”
- Traditionally based on
  - Qualitative similarity in chemical structure & properties
  - Single reference chemical representative of a “group”



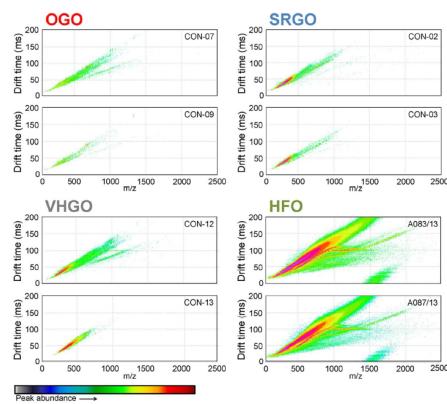
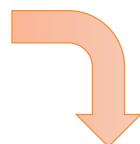
# Proposed solution based on the principle of “read-across.”

- Extend the single chemical approach to environmental mixtures
- Use high-throughput chemical and biological profiling to define “similarity”
- Similarity based on mixture of reference chemicals

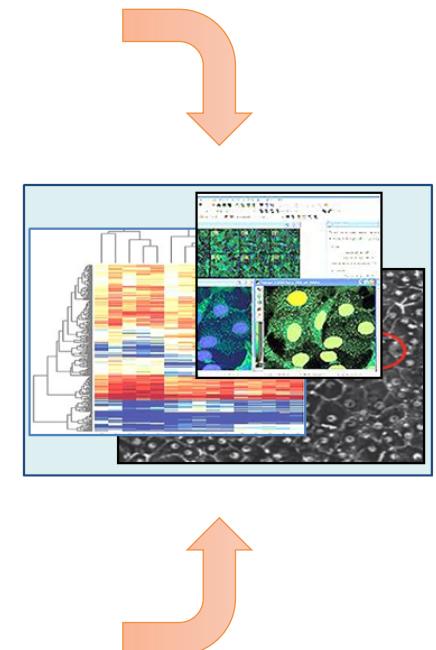
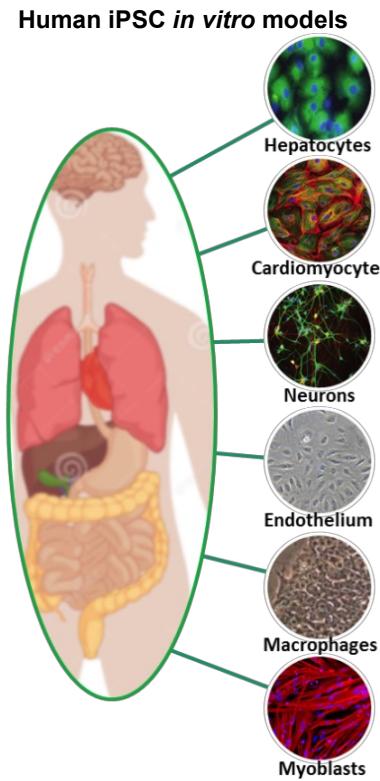


# Data for defining similarity

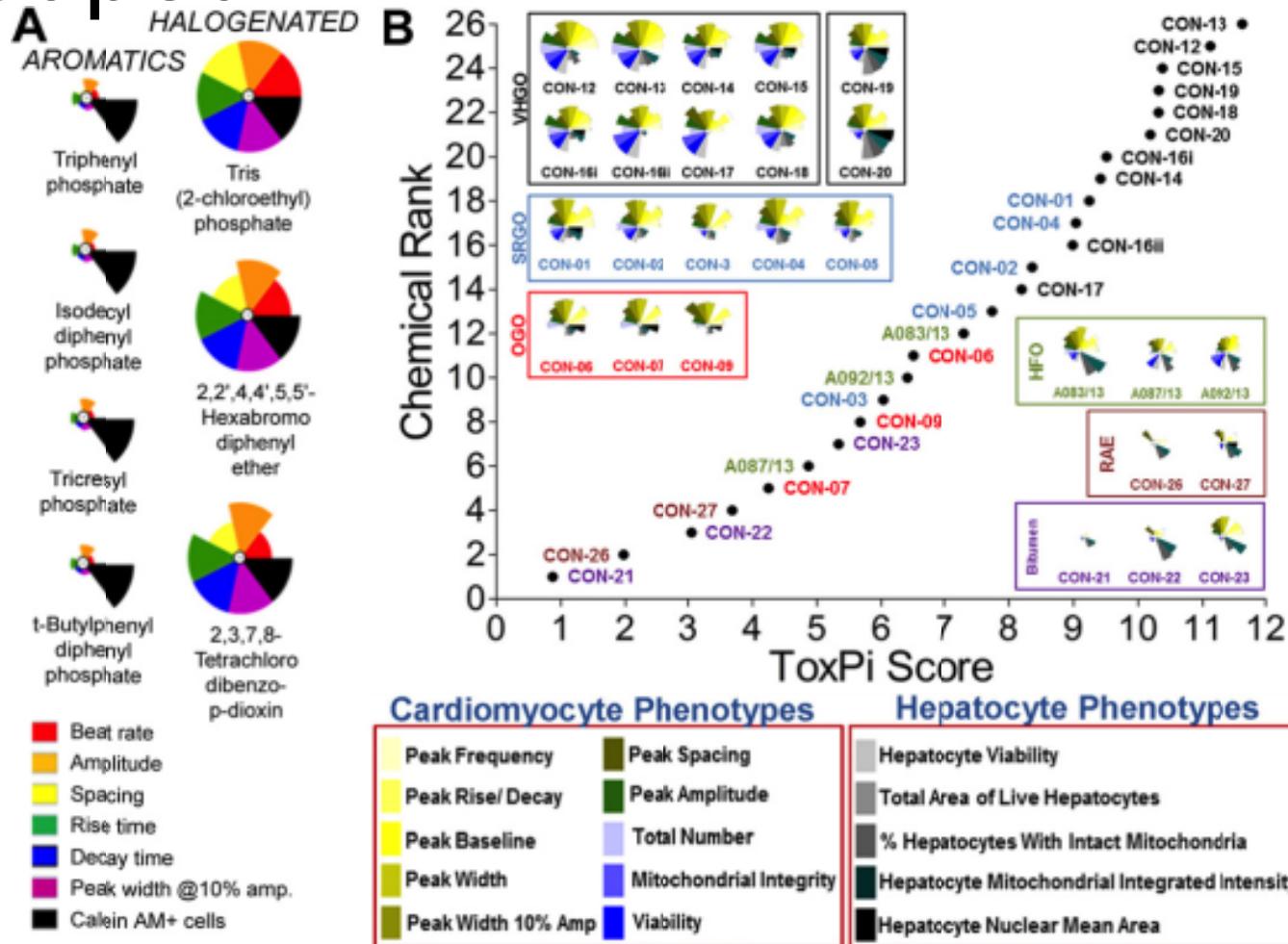
**High dimensional untargeted chemical profiling using Ion Mobility Spectroscopy/Mass Spectrometry**



**High dimensional biological profiling using induced-pluripotent stem cell-derived human tissues**

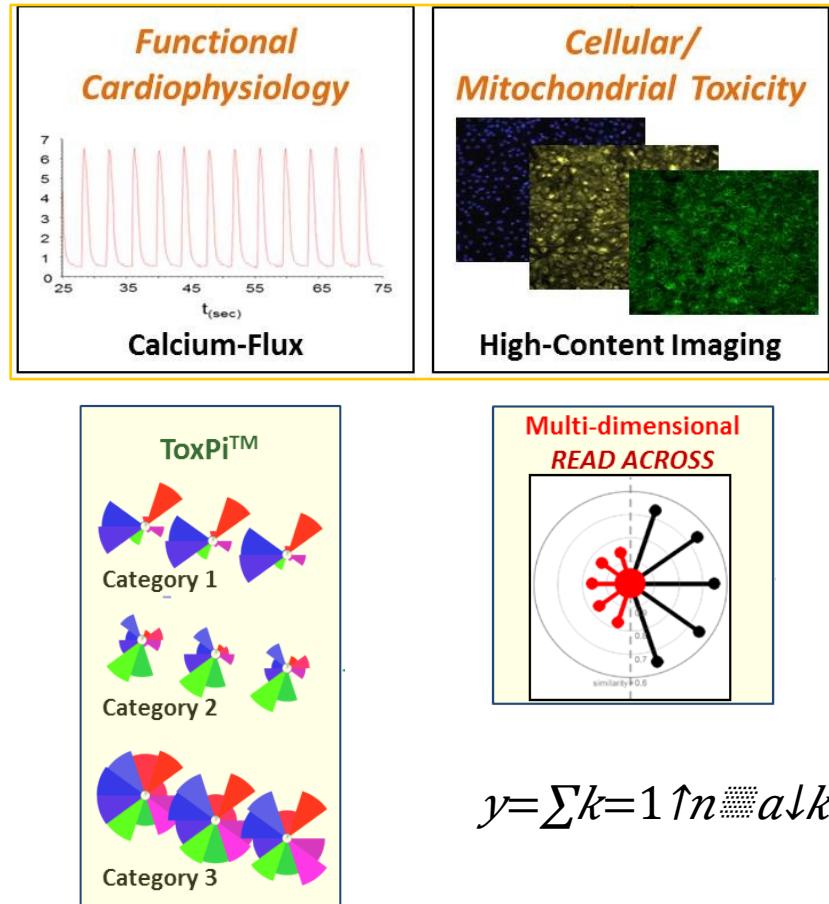


# Proof of principle that individual and complex substances can be grouped



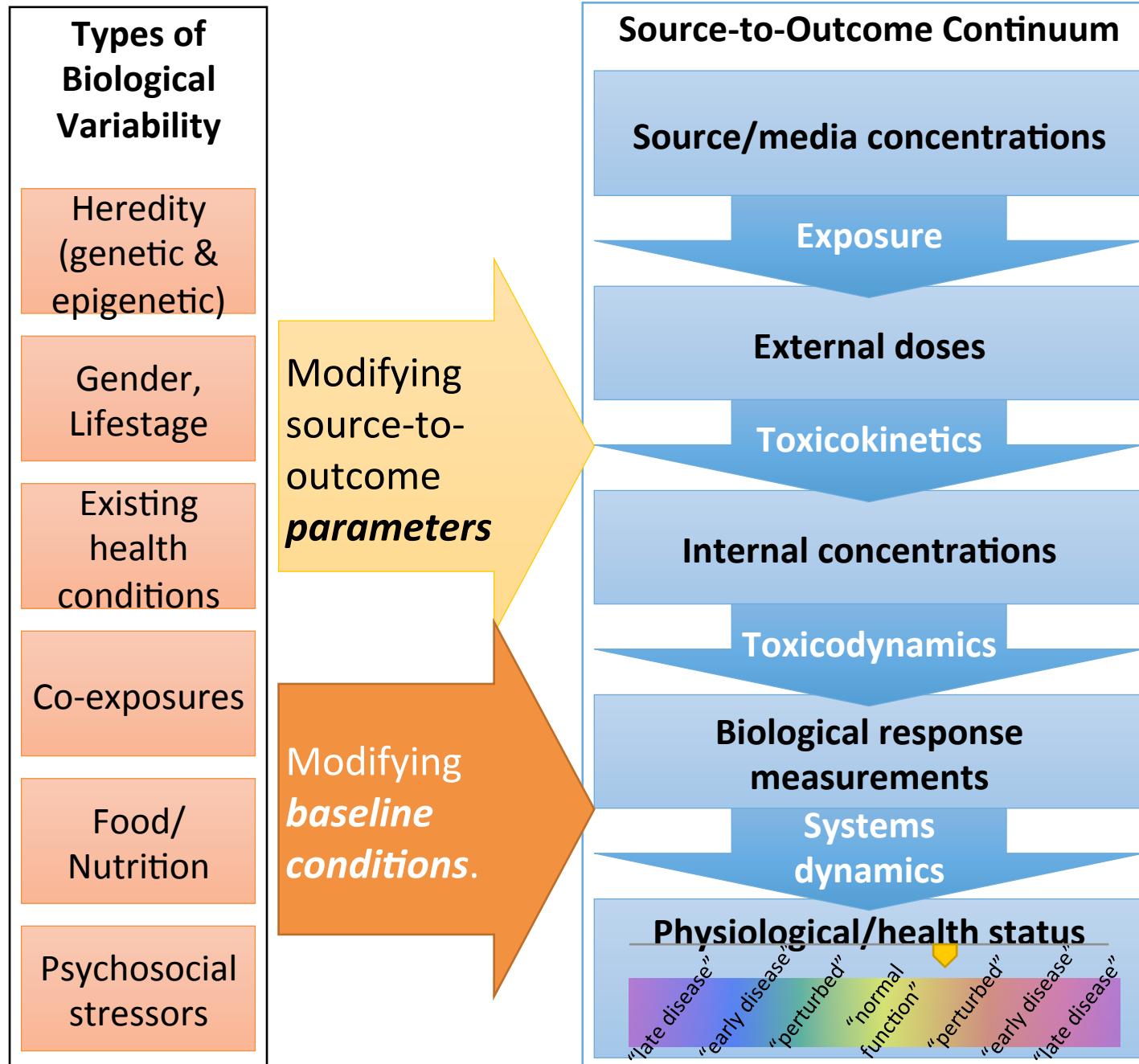
# Computational demands

- Multiple types of high-dimensional data processing
  - Multi-dimensional chemical data
  - Imaging data
  - Time-series data
  - Genomic (gene expression) data
- Multivariate data integration to define “similarity”
- Deconvolution to construct “mixture analogues” using reference chemicals
- For quantifying risk, classification is not enough – need a numerical prediction.



$$y = \sum_{k=1}^n a_k x_k$$

# Example Challenge: Characterizing human variability

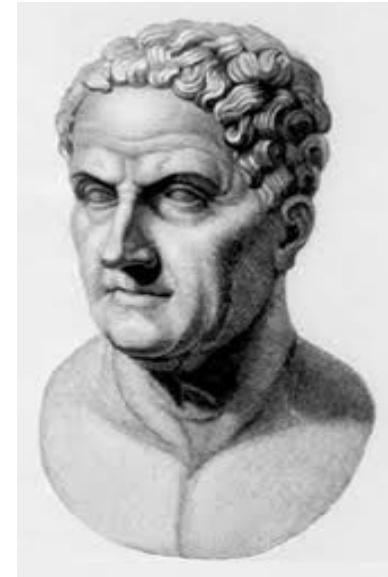
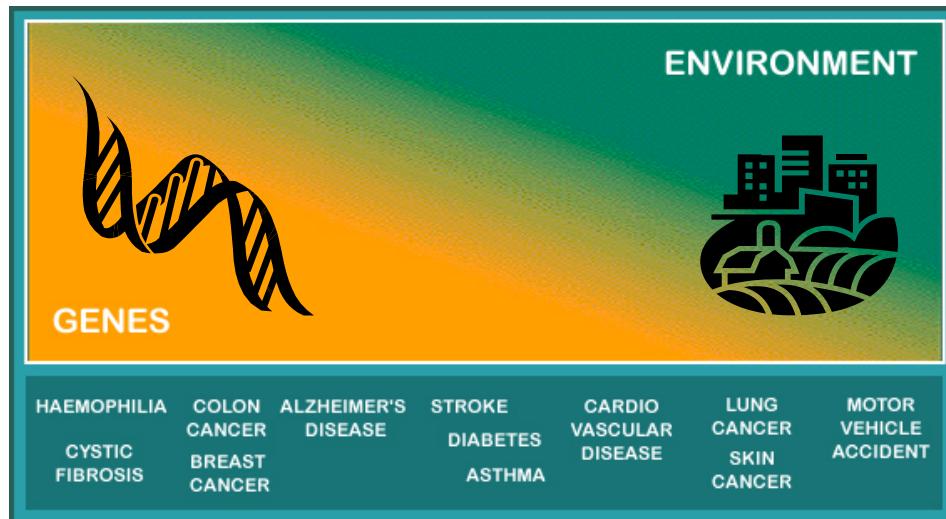


# Example Challenge: Characterizing human variability

*Claudius Galenus (Galen of Pergamum)*

129-217 AD

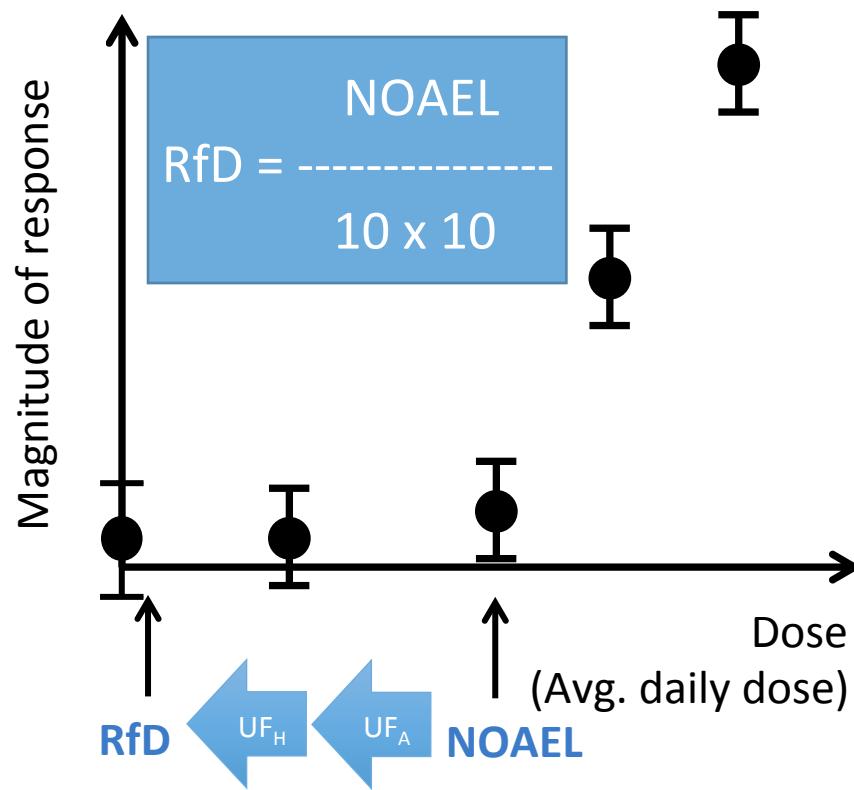
*“But remember throughout that no external cause is efficient without a predisposition of the body itself. Otherwise, external causes which affect one would affect all.”*



Library of Congress

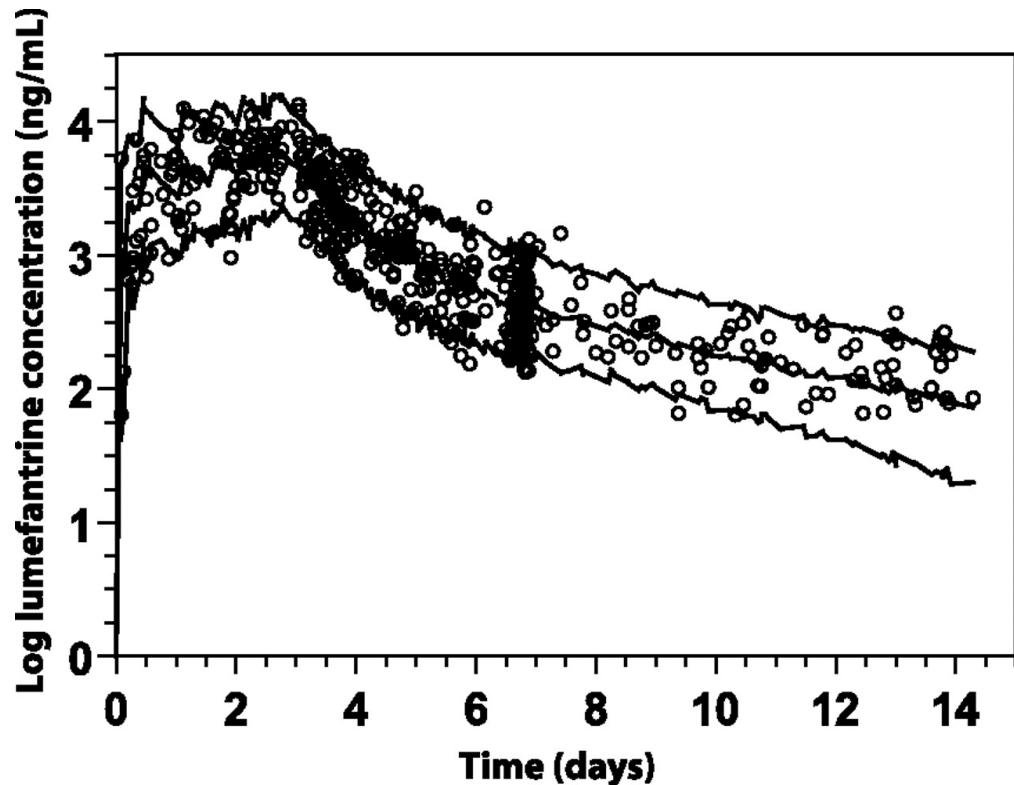
Slide courtesy of D. Threadgill

# Can we do better than dividing by a factor of 10?



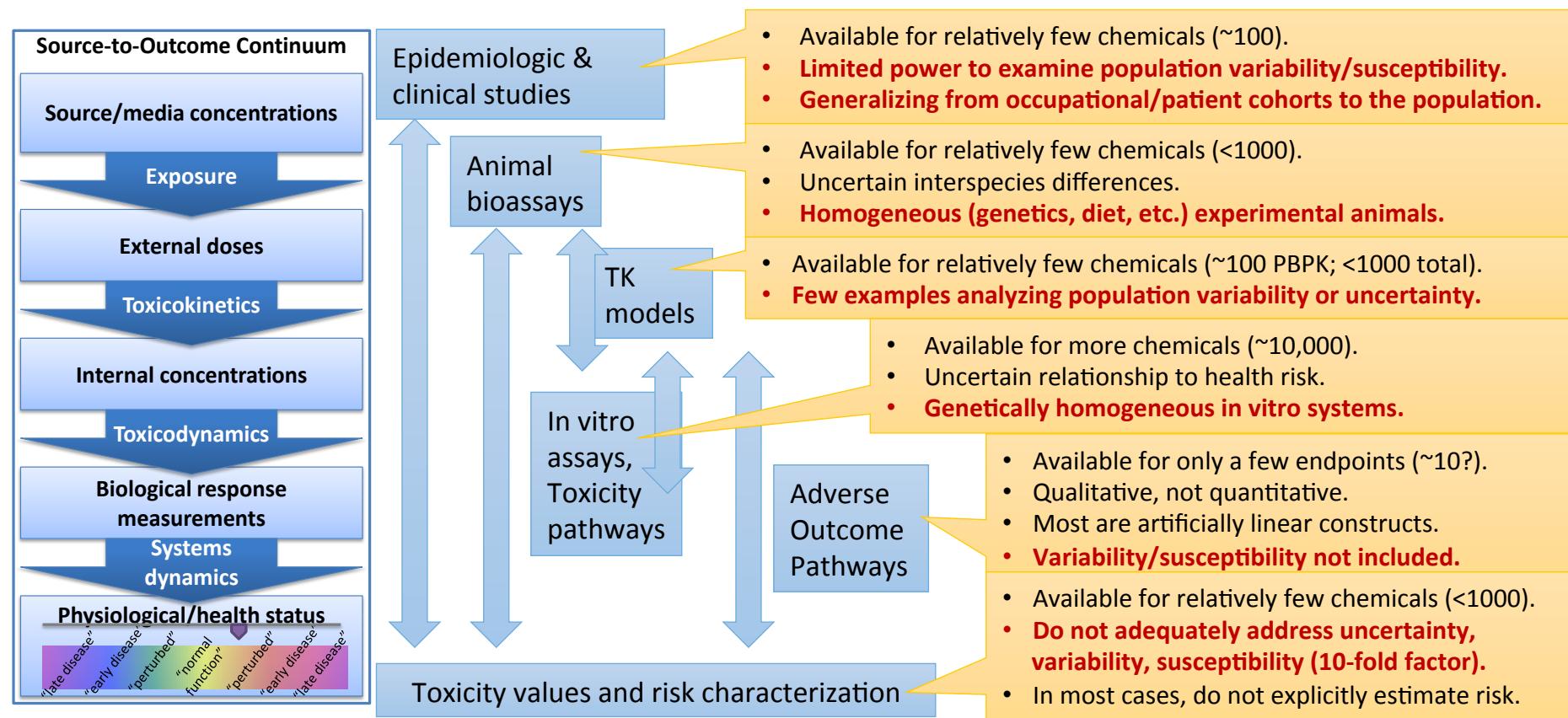
# For pharmaceuticals (and some environmental chemicals), generally have direct, human empirical data

- Long history of methodological development (population PK-PD).
- Both frequentist and Bayesian statistical approaches.
- What can you do in the absence of empirical data?



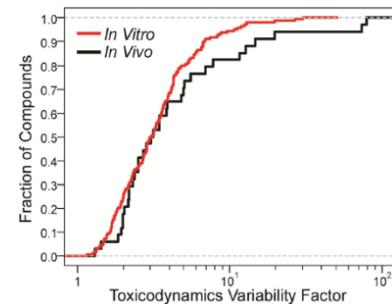
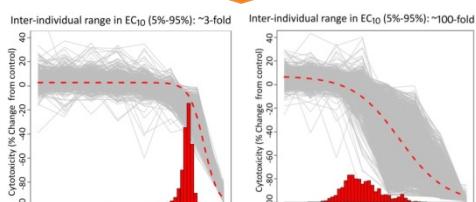
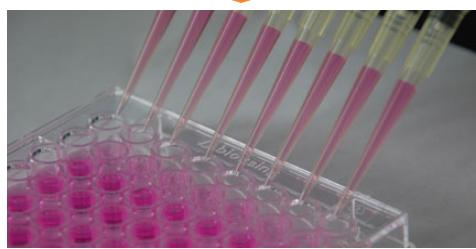
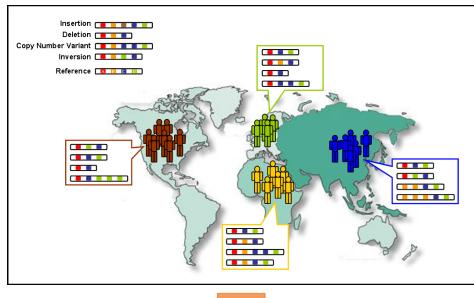
Joel Tarning et al. Antimicrob. Agents Chemother.  
2009;53:3837-3846

# Limitations to characterizing variability for environmental chemicals

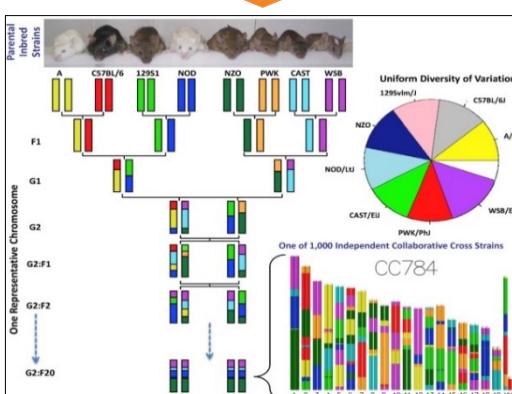
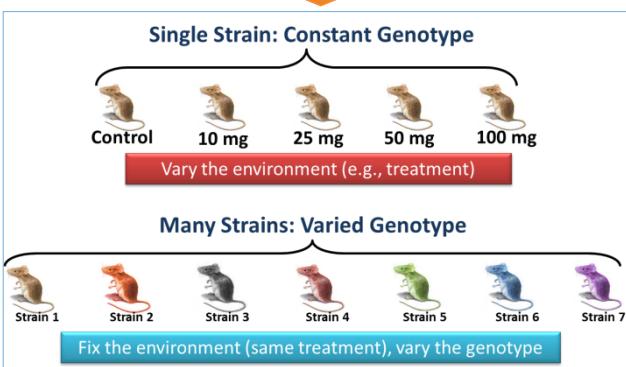
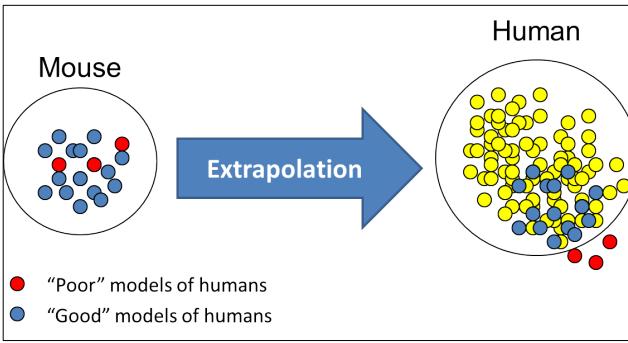


## Possible approaches without direct empirical data

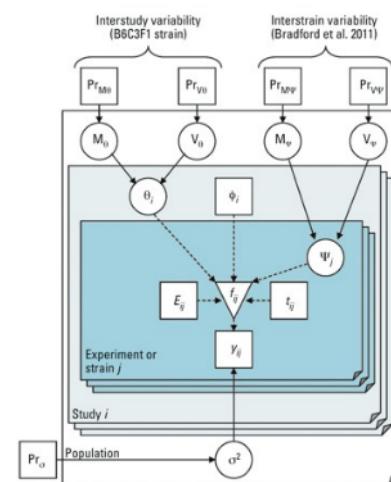
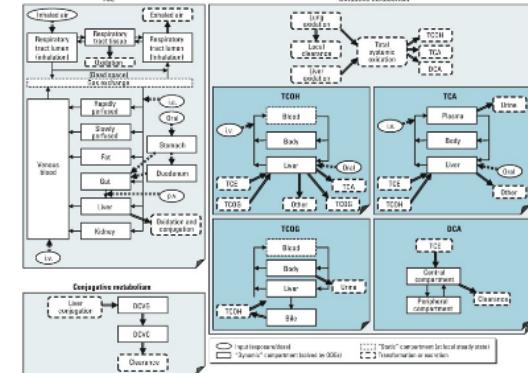
## In vitro data



## In vivo data

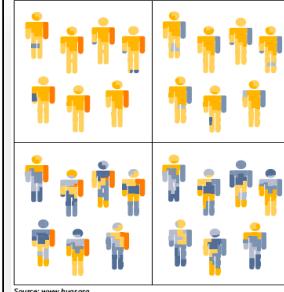
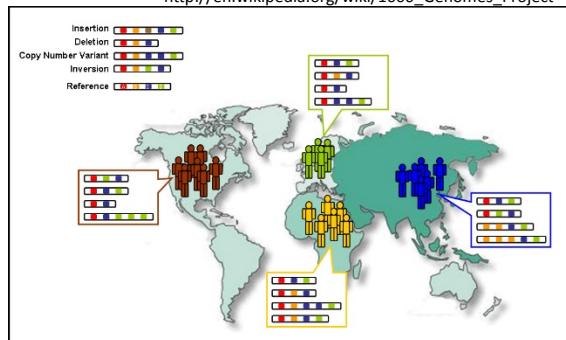


## In silico methods



## Genetically diverse human population

[http://en.wikipedia.org/wiki/1000\\_Genomes\\_Project](http://en.wikipedia.org/wiki/1000_Genomes_Project)



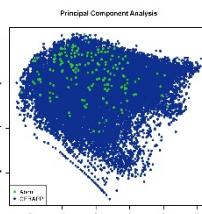
## Genetically defined sample of population

## High throughput *in vitro* model system



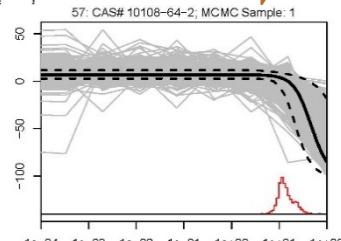
~1000 individuals  
cytotoxicity screening

## Structurally diverse chemical population

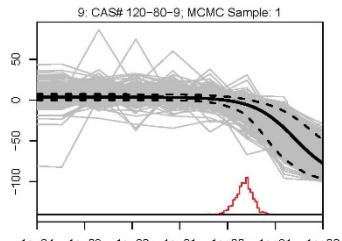


~170 compounds

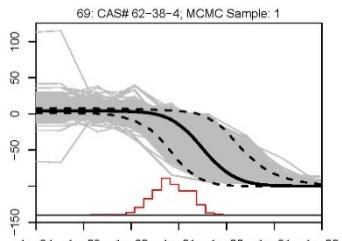
**Chemical-Specific TD Variability Factor ( $TDVF_{01}$ ):**  
The factor estimated to protect up to the most sensitive 1% for  
human toxicodynamic variability for a chemical



Cadmium Chloride  
~2-fold



Catechol  
~3-fold



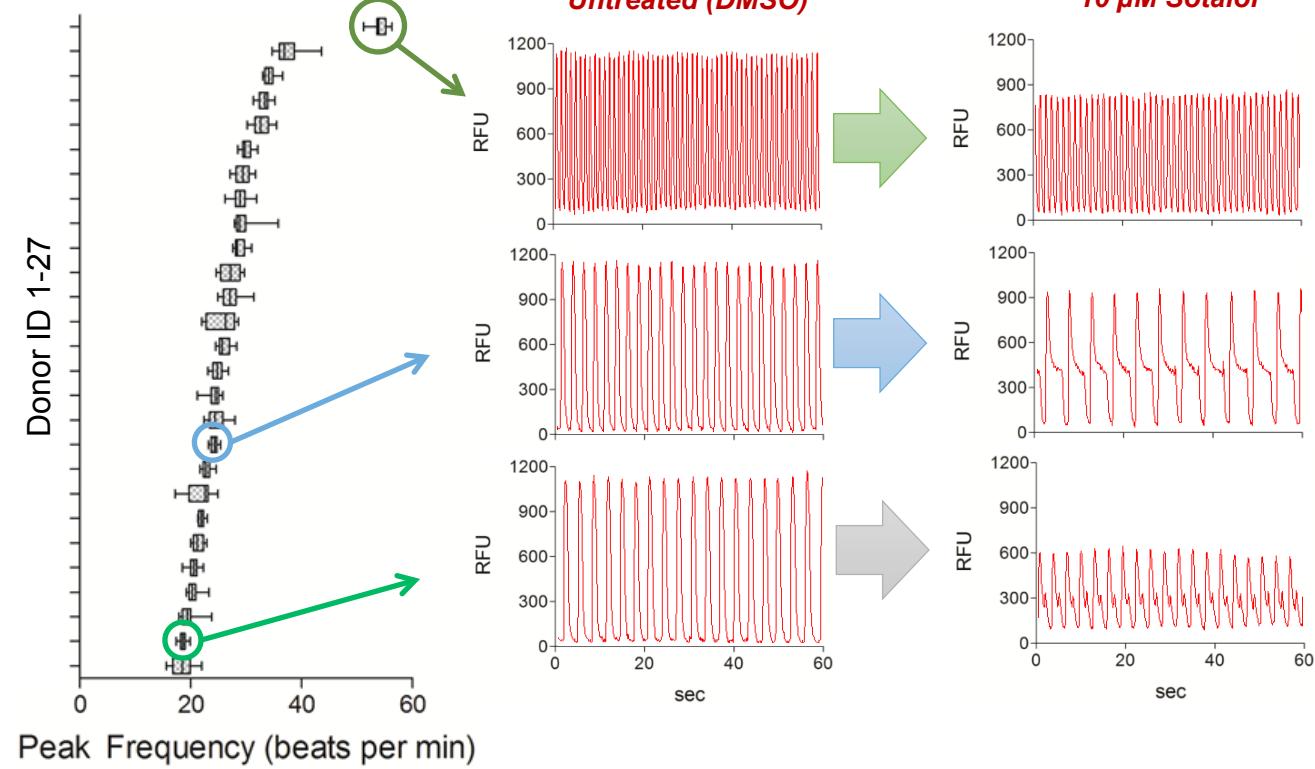
Organic and inorganic mercury compounds  
>8-fold

Abdo et al., 2015

Chiu et al., 2017

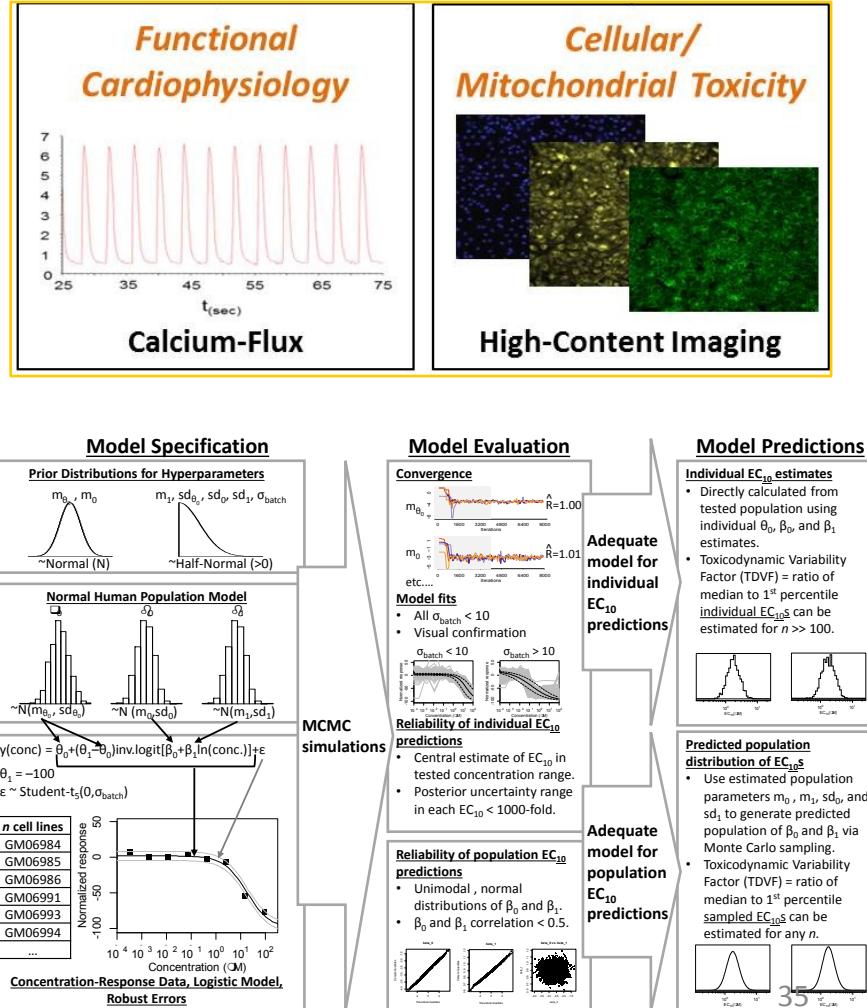
<https://doi.org/10.14573/altex.1608251>

# Population Toxicodynamics for Cardiotoxicity using Cardiomyocytes



# Computational demands

- Pre-processing multiple types of high-dimensional data
  - Imaging data
  - Time-series data
  - Genomic (gene expression) data
- Automated concentration-response modeling
- Distinguishing true population heterogeneity from random errors



# Quantifying risk and uncertainty

## Source-to-Outcome Continuum

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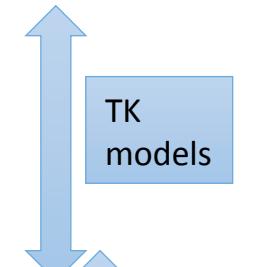
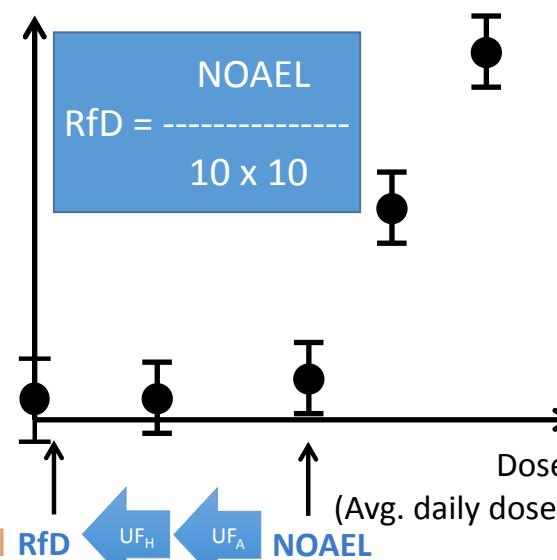
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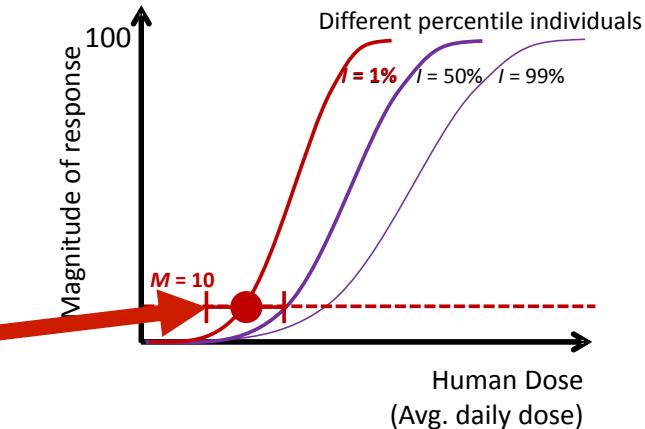
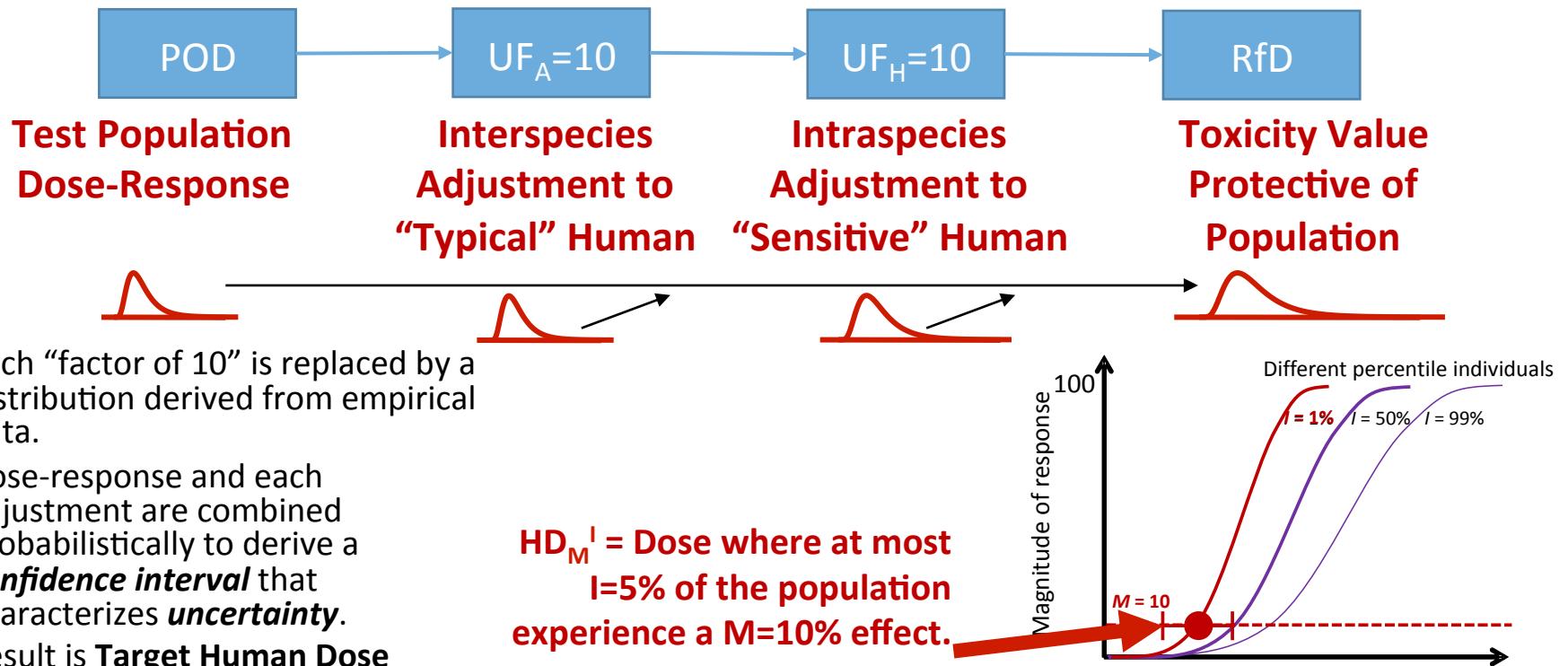
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"normal function"  
"perturbed"  
"early disease"  
"late disease"



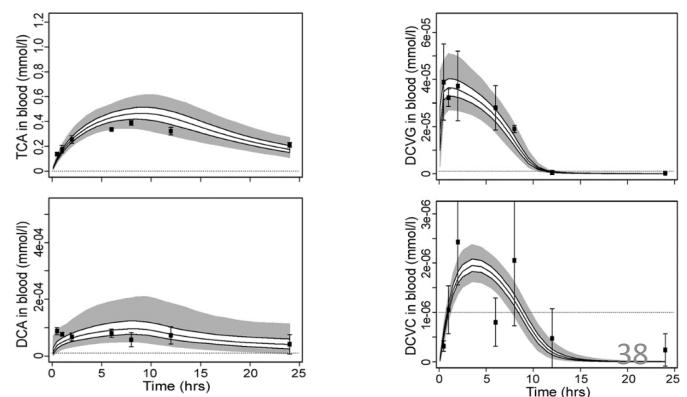
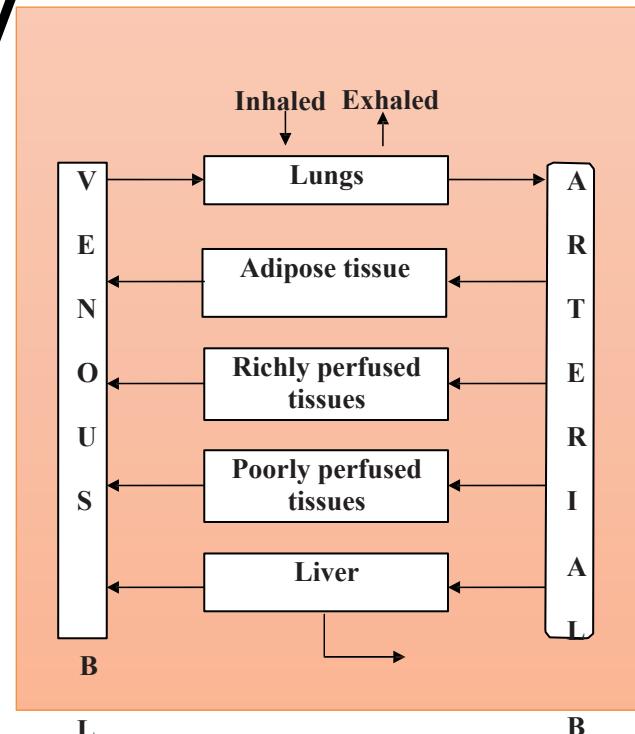
- What is the risk in terms of severity & incidence in the population?
- What are the confidence intervals?

# Application of Probabilistic Approaches to Quantify Risk and Uncertainty



# TK models: approaches to quantify uncertainty

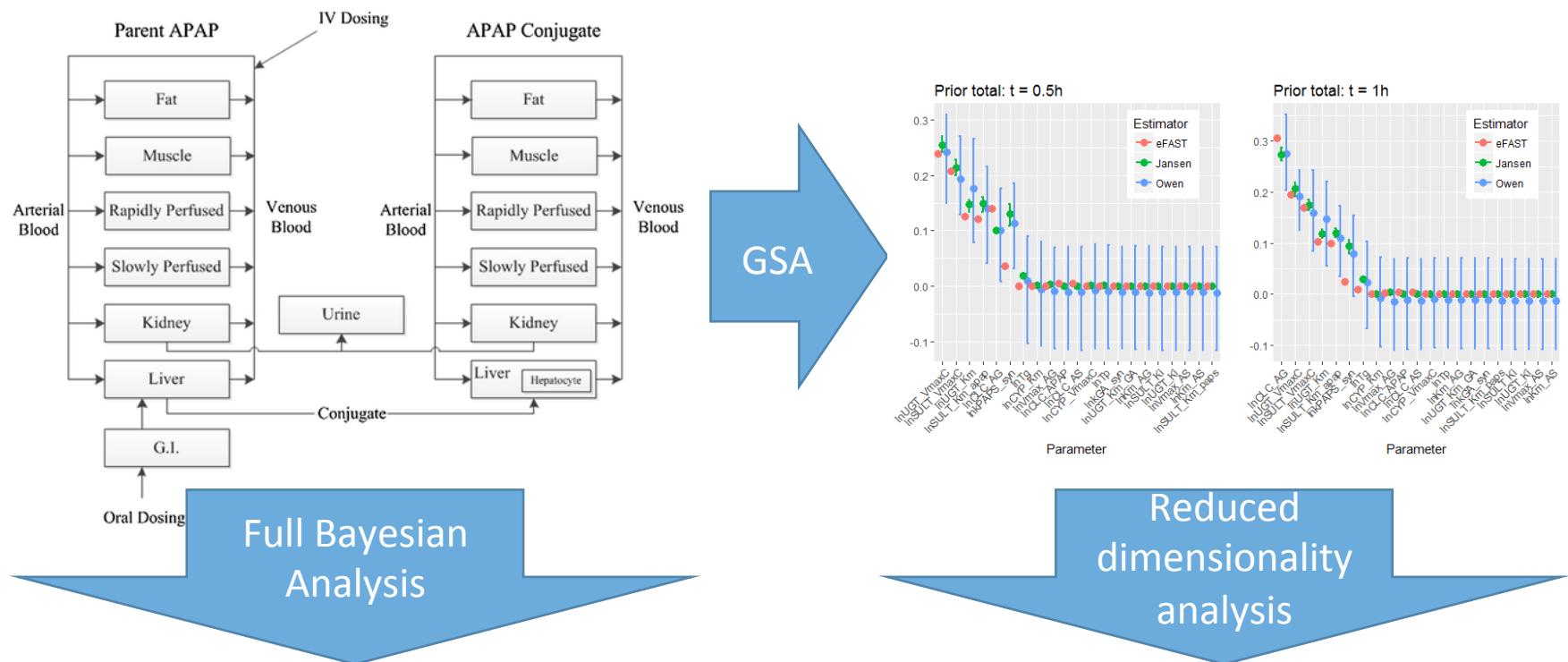
- Physiologically-based pharmacokinetic models are needed in the absence of empirical data
- Many parameters, each with uncertainty and population variability
- Models are not uniquely identifiable from direct observations
- Two approaches have been used:
  1. **Fit** all parameters using Bayesian approach
  2. **Fix** all but a small subset of parameters at nominal values, and fit the remaining using a frequentist approach
- #1 can be computationally prohibitive, whereas #2 can lead to biased results.



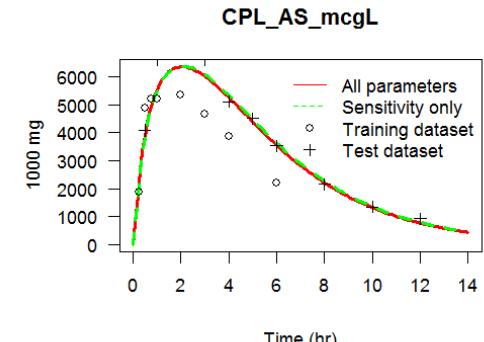
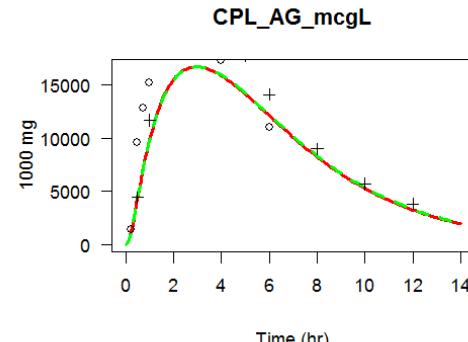
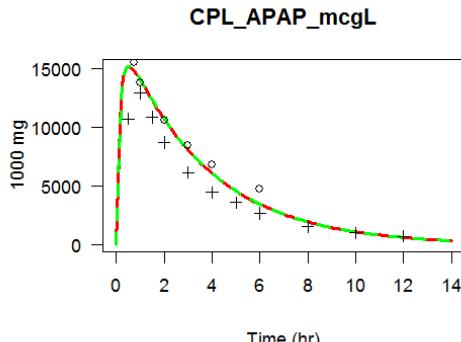
# Reducing dimensionality with global sensitivity analyses

- **Hypothesis:** Can reduce dimensionality of Bayesian analysis by fixing “low sensitivity” parameters at nominal values without introducing significant bias.
- Test hypothesis by comparing reduced model results with those of full Bayesian analysis (“gold standard”).
- Need **global** rather than **local** sensitivity analysis because of potential nonlinearities across parameter space
- **Sobol indices:** Reduction in output variance if the input parameter were known exactly
  - First order term measures direct effect
  - Interaction term measures effects combined with other parameters
  - Multiple algorithms for calculating indices

# Preliminary results



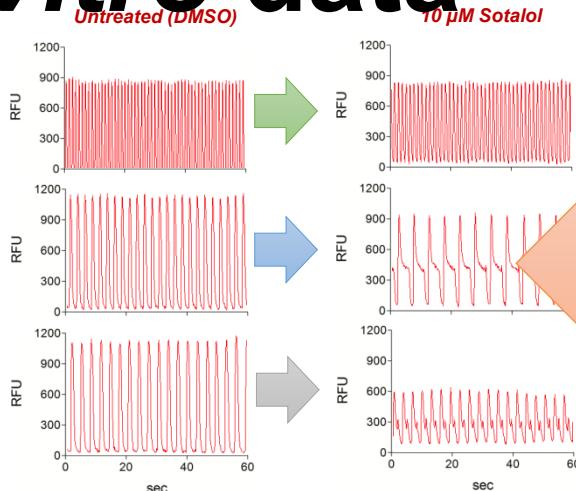
21 parameters, 19 hr simulation time



Results nearly indistinguishable

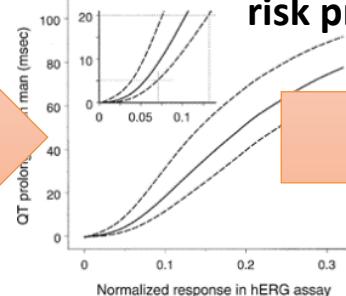
# Predicting population risk from *in vitro* data

Untreated (DMSO)



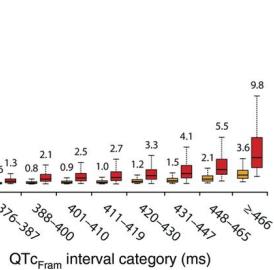
10  $\mu$ M Sotalol

Calibration  
to human  
clinical data



Biomarker-  
based  
population  
risk prediction

Men, age 50–70 years



## Molecular

Chemical blocks ion channel

**Chemical**

**Molecular level**

**DRUG**

Source: hERGAPDbase  
hERG

## Cellular

Action potential prolonged

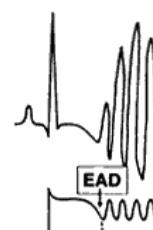
**Cellular level**

QT interval  
prolonged

**Tissue level**  
(Heart ECG)

## Tissue

Increased likelihood  
of torsades de  
pointes



## Organism

Increased  
likelihood of  
myocardial  
infarction



Increased  
likelihood of  
death



Inter-individual variability (TK, TD)

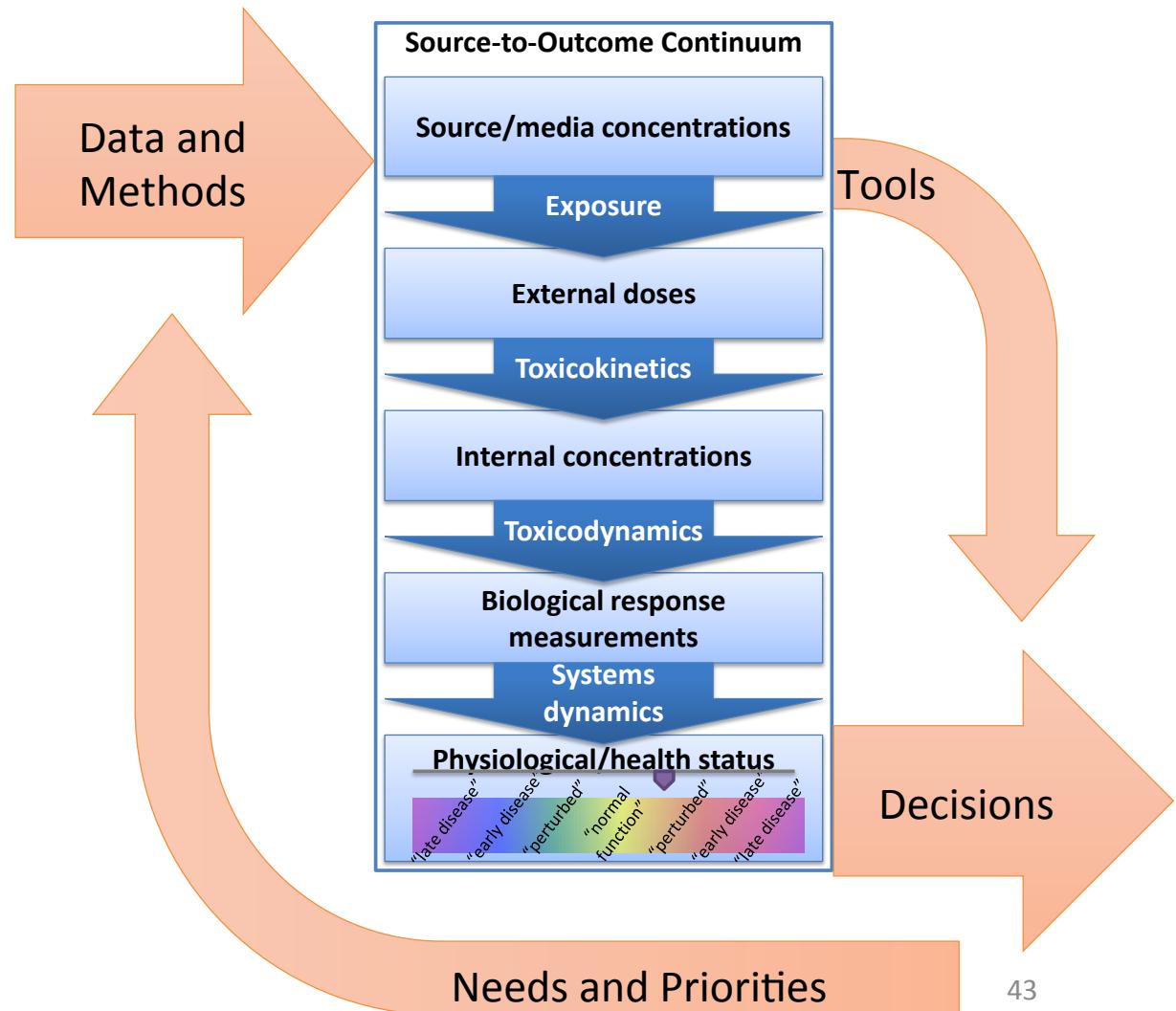
Variability and stochasticity from other stressors/risk factors

# Computational demands

- Monte Carlo simulation
- Bayesian estimation  
using Markov Chain  
Monte Carlo
- Stiff ODE solvers
- Global sensitivity  
analyses

# Risk assessment is an inherently *translational* science

- Requires integration of data from multiple sources across the source-to-outcome continuum.
- Aimed ultimately at supporting decisions, not testing hypotheses.
- Additional challenges involve moving from a researching methods to developing tools



# Echoes Prof. Dougherty's seminar “Modern engineering as a translational science”....



## Science and Action

- **Arturo Rosenblueth and Norbert Wiener:** “The intention and the result of a scientific inquiry is to obtain an understanding and a control of some part of the universe.”
  - For them, science and translational science are inextricably linked, the ultimate purpose of acquiring scientific knowledge being to translate that knowledge into action.

1/23/2017  
1/23/2017

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## Benefits of a Translational System

- A translational mathematical system provides guides.
  - Guide the scientist in building a fruitfully applicable model
  - Guide the engineer in studying costs and benefits of action
  - Guide the technologist in devising devices or treatments.
- In a properly functioning relationship, the scientist does not hand the engineer a set of data and ask the engineer to find something in it; instead, assuming a translational goal, the enterprise should be guided by the goal and this goal should already have led a carefully designed experiment.

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# Summary of computational demands of chemical risk assessment

Multiple types of high-dimensional data processing

- Multi-dimensional chemical data
- Imaging data
- Time-series data
- Genomic (gene expression) data
- Multivariate data integration to define “similarity”
- Deconvolution to construct “mixture analogues” using reference chemicals
- Automated concentration-response modeling

- Distinguishing true population heterogeneity from random errors
- For quantifying risk, methods for classification are not enough – need a numerical prediction.
- Monte Carlo simulation
- Bayesian estimation using Markov Chain Monte Carlo
- Stiff ODE solvers
- Global sensitivity analyses

Opportunities for students/postdocs: Chemical Risk Assessment suffers from lack of expertise in both developing and applying computational methods.