

The Science of Exposure Assessment

By

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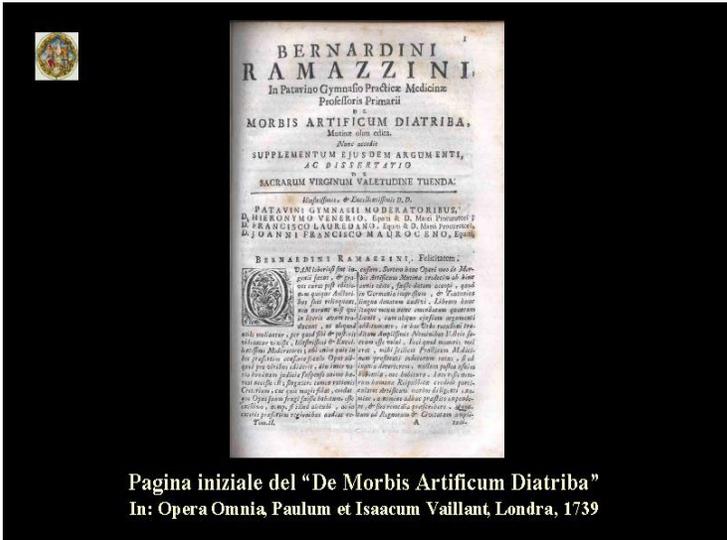
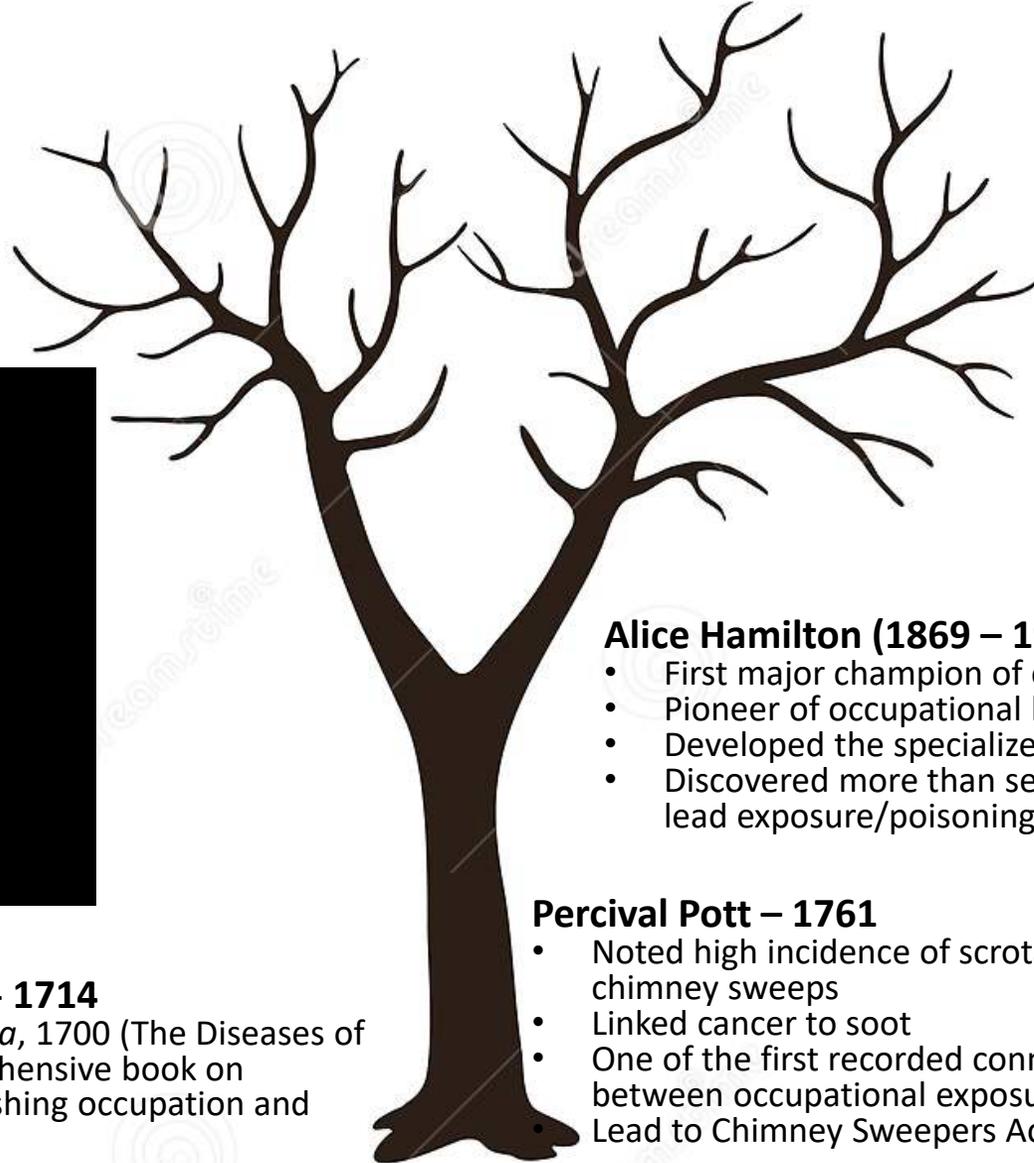
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Outline

- A view of historical evolution
- Responding to evolving complex environmental health problems
- Public health stakes
- Computational exposure science at the cutting edge



Exposure Science's Historical Context



Pagina iniziale del "De Morbis Artificum Diatriba"
In: Opera Omnia, Paulum et Isaacum Vaillant, Londra, 1739

Bernardino Ramazzini, 1633 - 1714

De Morbis Artificum Diatriba, 1700 (The Diseases of Workers) – the first comprehensive book on industrial medicine establishing occupation and disease links



Alice Hamilton (1869 – 1970)

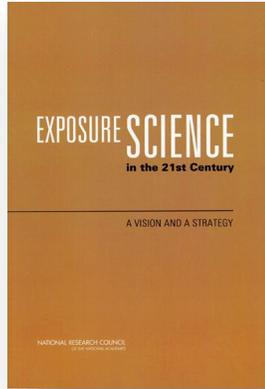
- First major champion of occupational health in the United States
- Pioneer of occupational health, safety and industrial hygiene.
- Developed the specialized field of industrial hygiene in the USA
- Discovered more than seventy industrial processes where workers suffered lead exposure/poisoning

Percival Pott – 1761

- Noted high incidence of scrotal cancers in chimney sweeps
- Linked cancer to soot
- One of the first recorded connections between occupational exposures and cancer.
- Lead to Chimney Sweepers Act of 1788



Becoming a Distinct Science



- NAS 2012 Exposure Science in the 21st Century

- 2005 – Exposome Introduced
**Cancer Epidemiology,
Biomarkers & Prevention**

Complementing the Genome with an "Exposome": The Outstanding Challenge of Environmental Exposure Measurement in Molecular Epidemiology

Christopher Paul Wild

Cancer Epidemiol Biomarkers Prev 2005;14:1847-1850.

- 1990 – ISEA/ISES Founded

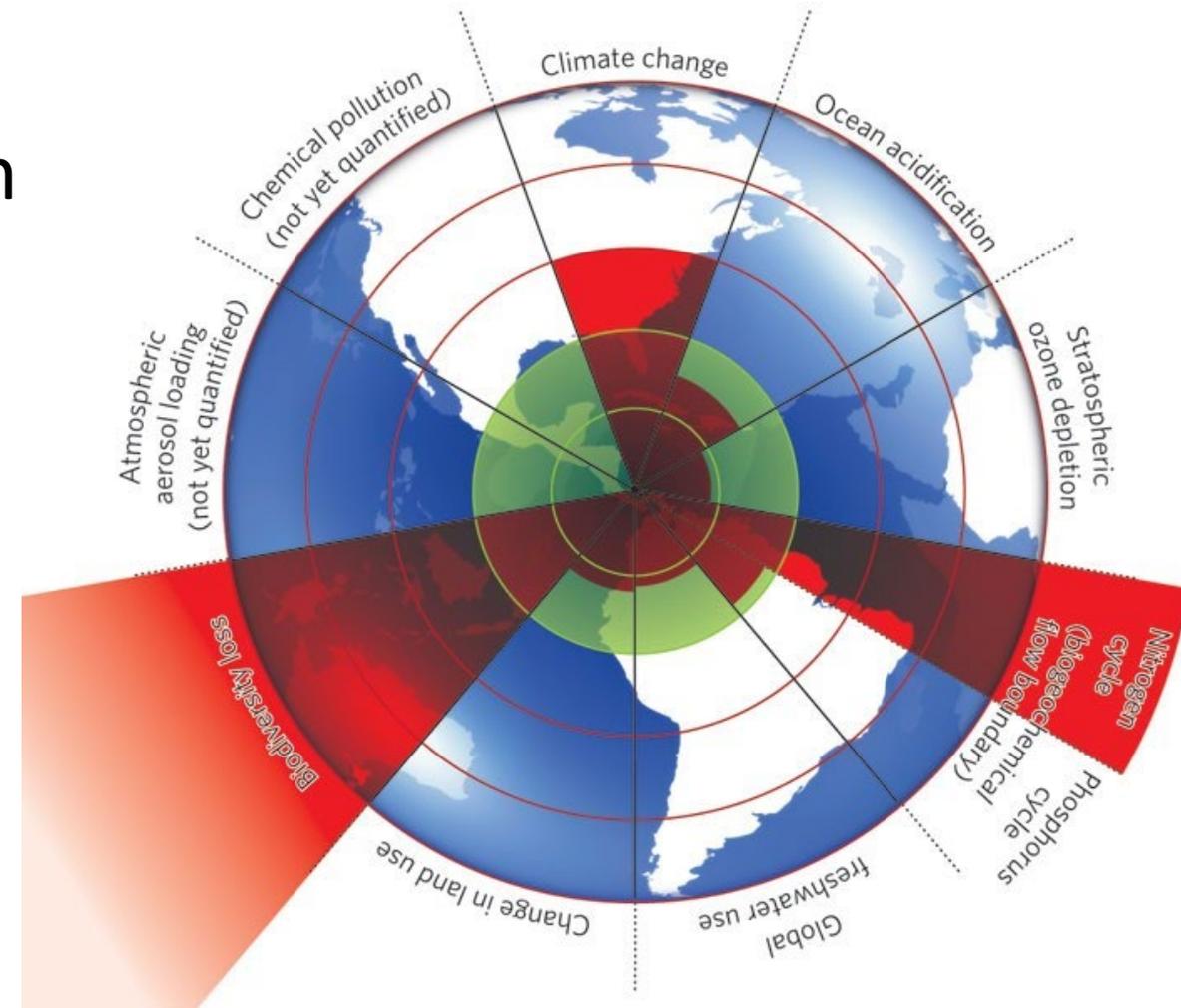


AACR American Association
for Cancer Research



Growing Demand on Exposure Science to Help Address Expanding and Wicked Environmental Health Challenges

- Climate change
- Population pressure on ecosystem services
- Environmental justice
- Urbanization
- Natural resource depletion
- Loss of biodiversity
- Chemical contaminants
- Habitable planet
- Infectious disease pandemics



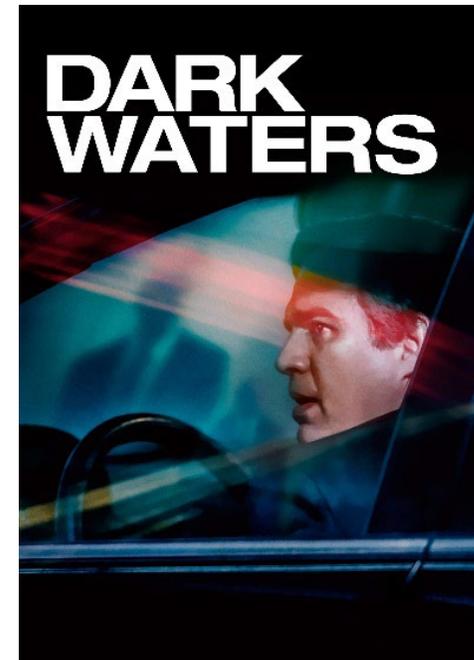
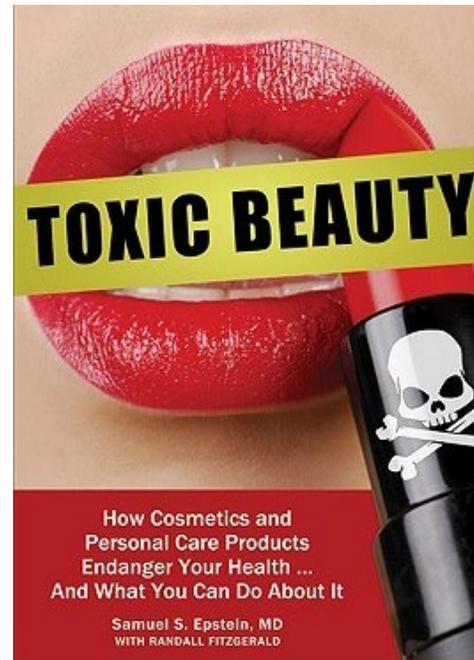
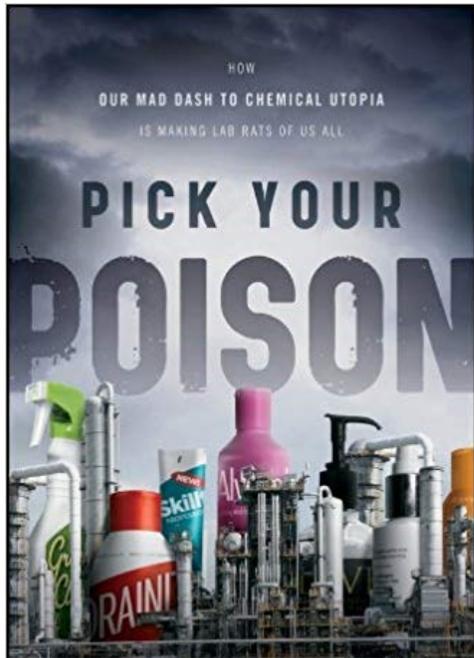
Exposure Science's Expanding Scope/Complexity

Integral Sciences

- Biological Sciences
- Ecological Sciences
- Social Sciences
- Chemistry
- Data Sciences
- Personalized Medicine
- Exposomics
- One Health
- Systems Science



Case Example of Exposure Science and the Challenge of Chemical Exposure



Computational Exposure Science Colleagues

Peter P. Egeghy

Ann M. Richard

Jon Sobus

Elin M. Ulrich

Kristin Isaacs

John Wambaugh

Caroline Ring

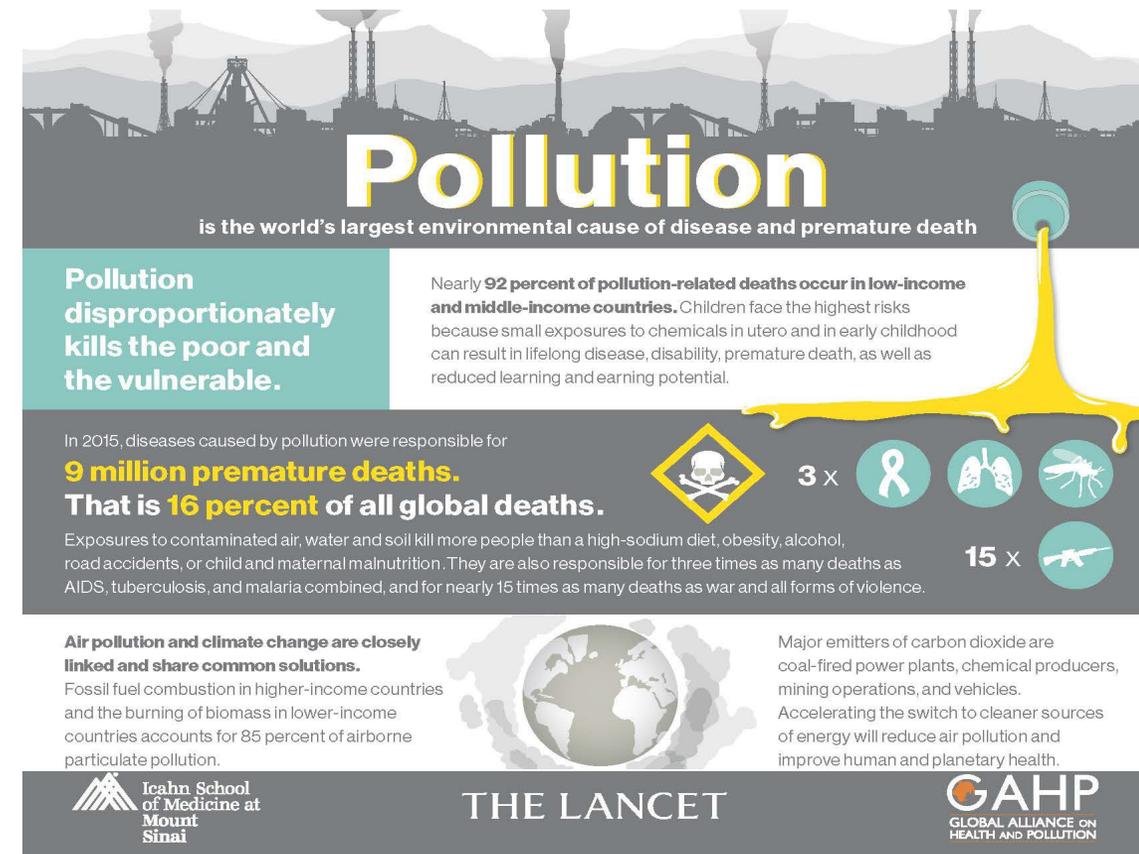
Risa R. Sayre

Antony J. Williams

Russell S. Thomas

Challenge: Chemical Pollution and Public Health

- Pollution is known to be a leading public health threat
- Effects likely underestimated
- Exposure and effects are poorly understood
- A large proportion of the environment-attributed disease is of unknown etiology (Rappaport, 2016)
- Chemical production and release to the environment vastly outpace ability to test and measure



Source: Landrigan et al., 2017

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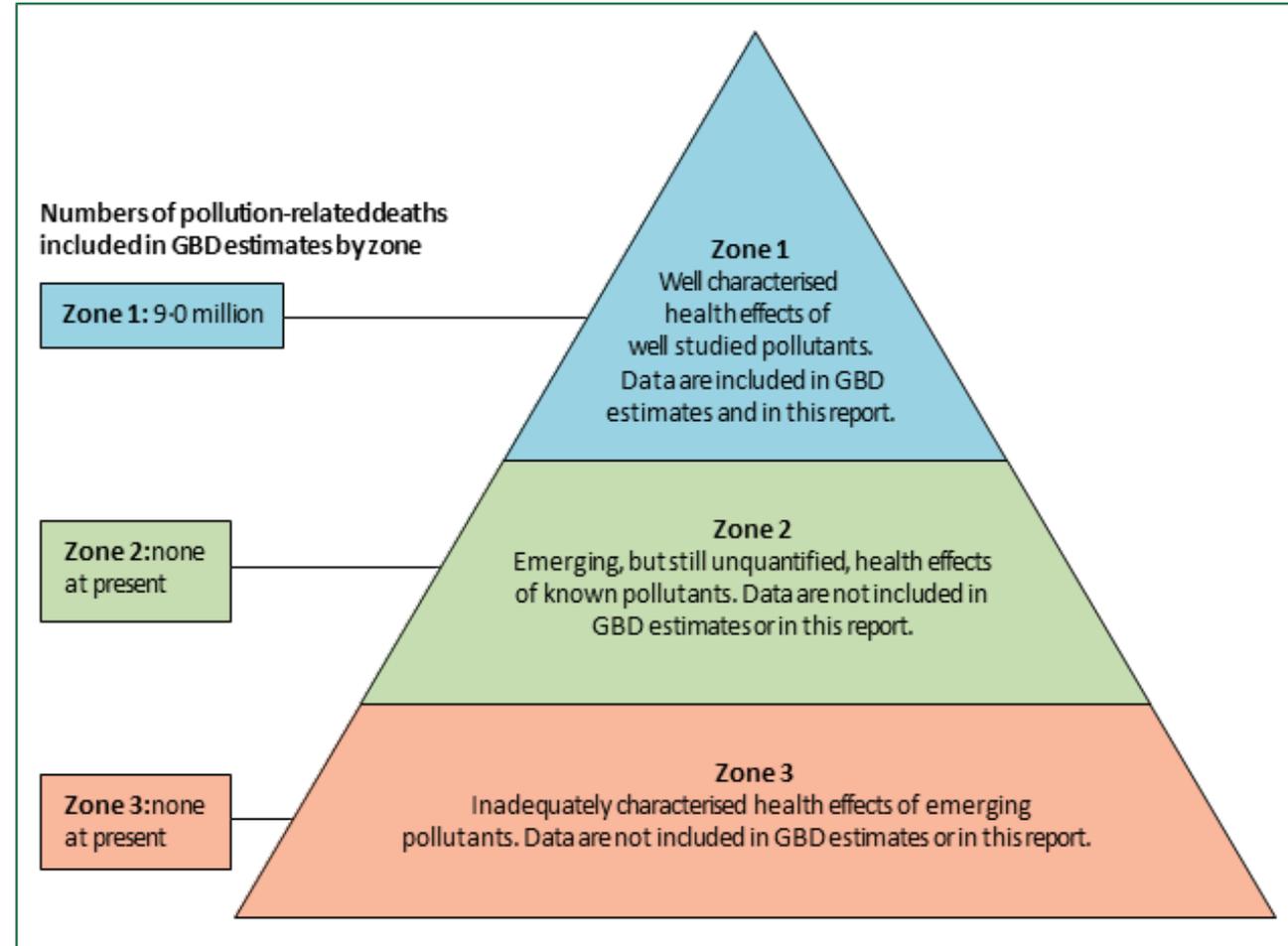


Figure 3: The pollutome

Source: Landrigan et al., 2017

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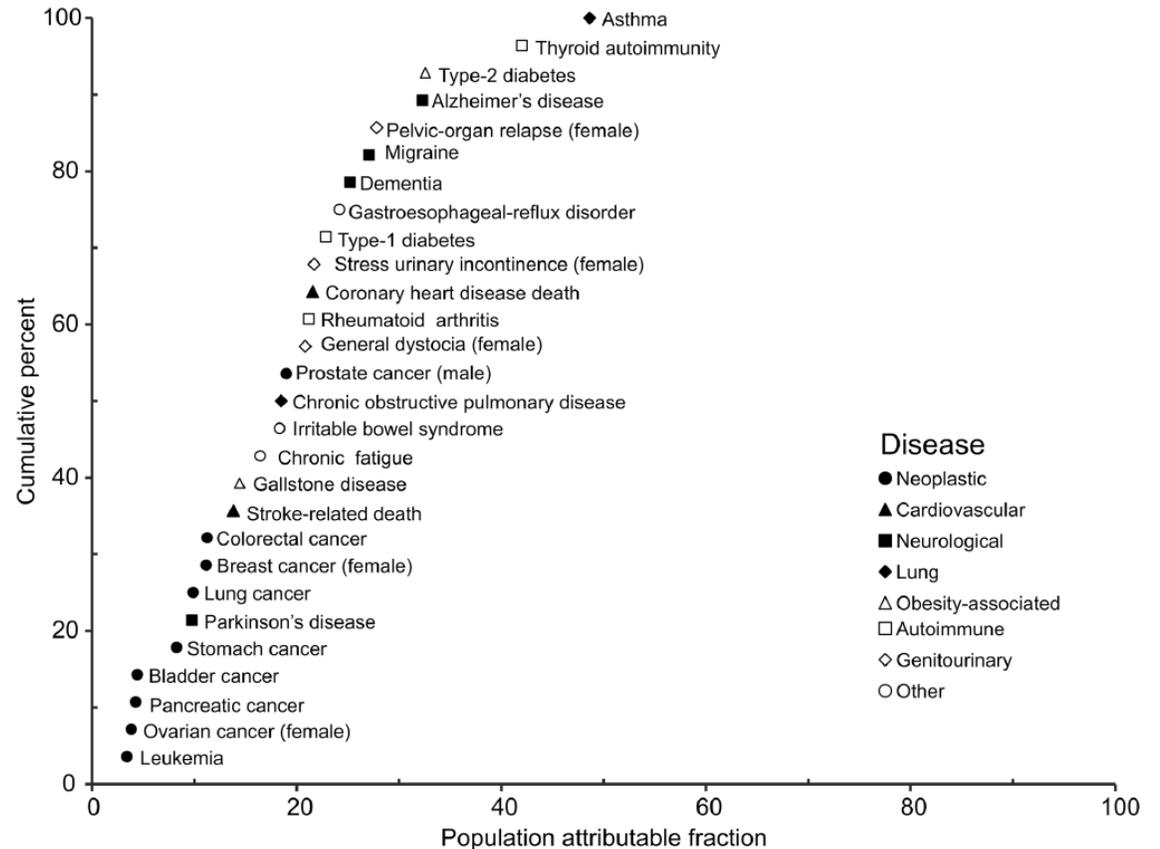
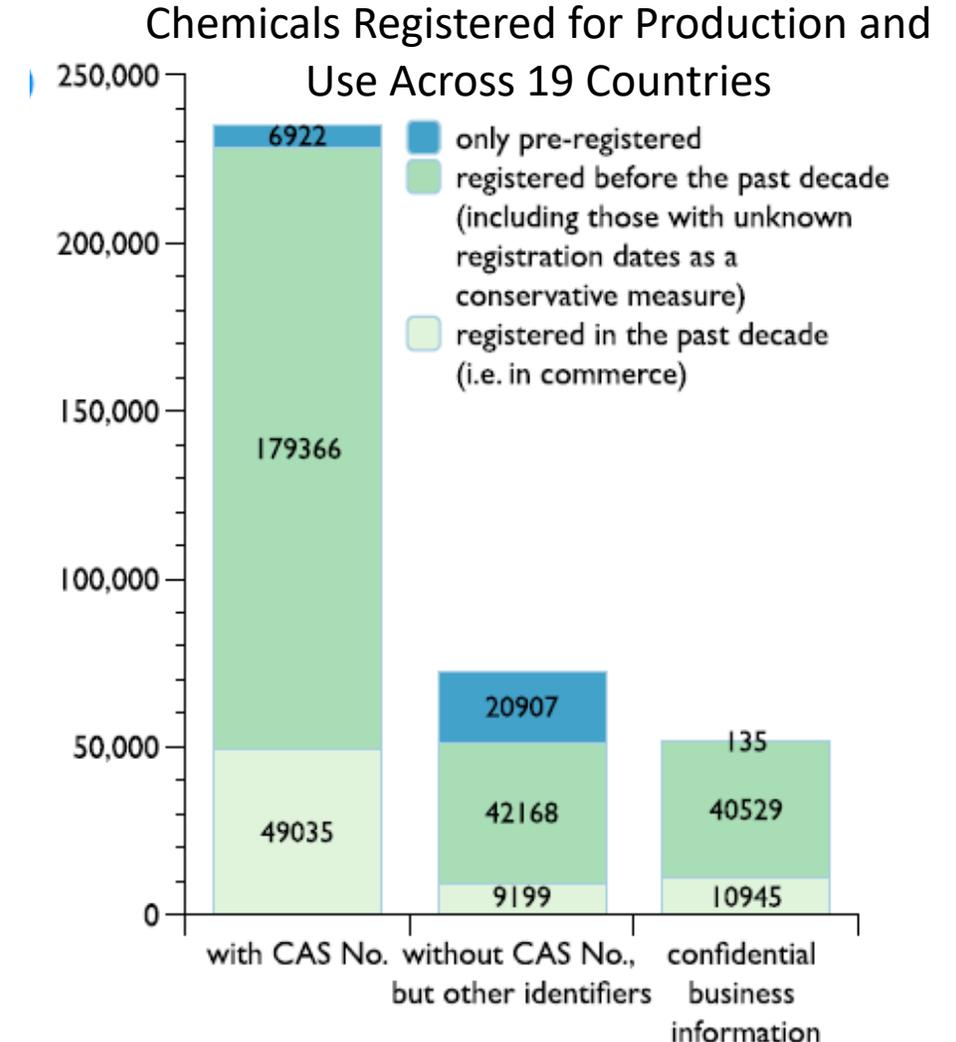


Fig 1. Population attributable fractions (PAFs) for 28 disease phenotypes estimated from studies of monozygotic twins. Sources of data and statistics are summarized in [Table 2](#).

Challenge: Managing the Enormous Number of Chemicals in Commerce

- Pollution is known to be a leading public health threat
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- **Chemical production and release to the environment vastly outpaces ability to test and measure**
 - 350,000 chemicals and mixtures registered for production and use across 19 countries (Wang et al., 2020)
 - The EPA CompTox Chemicals Dashboard lists 37,143 chemicals within its CPDAT, Chemical and Products Database (https://comptox.epa.gov/dashboard/chemical_lists/CPDA)
 - EPA's DSSTox database currently lists 1.2 million substances of environmental health relevance (Grulke et al., 2019)
 - TSCA lists a total of 86,631 chemicals with about half that number (42,039) identified as currently active in U.S. commerce (February 2022)
 - New assessment approaches are needed



Source: Wang et al. 2020

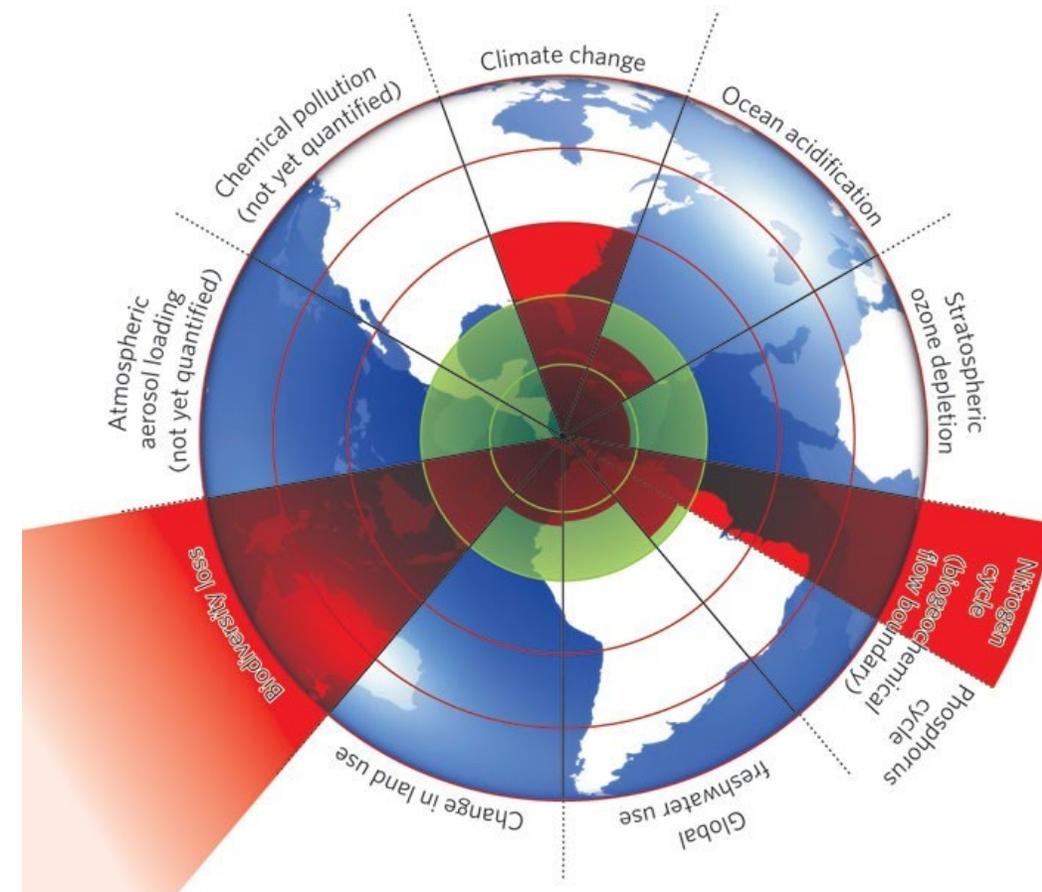
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Challenge: Threat to a Habitable Planet

- Need improved scientific basis for developing effective hazard screening, monitoring and management options that will avoid transgressing planetary boundary (Rockstrom et al., 2009)
- Current chemical management practices do not address this issue and must therefore be complemented with new approaches (Persson et al., 2013)
- Sufficient evidence shows stresses on ecosystem and human health at local to global scales, suggesting that planetary boundary is being transgressed (Diamond et al., 2015)
- Production and releases of novel entities outstrips global capacity for assessment and monitoring (Persson et al., 2022; Cousins et al., 2022)

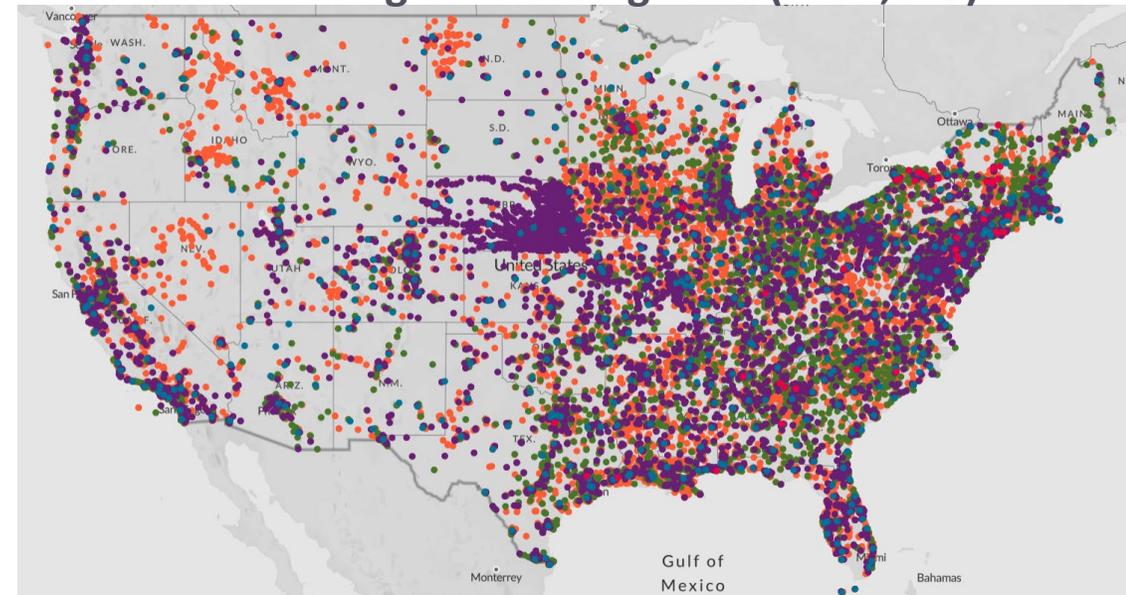


Source: Rockstrom et al., 2009

PFAS A Case-in-Point

- **US Mortality Attributed to PFAS Exposure** (Wen et al., EHP 2023)
 - Analysis based on NHANES 1999 – 2014; 7 PFAS measured in serum
 - Hazard Ratios for mortality:
 - All-cause -- 1.38 (95% CI: 1.07, 1.80)
 - Heart disease – 1.58 (95% CI: 1.05, 2.51)
 - Cancer – 1.70 (95% CI: 1.08, 2.84)
 - PFOS exposure associated annual mortality
 - 1999 – 2015: ~382,000
 - 2015 – 2018: ~69,000
 - Estimated total deaths >6 million
- **Risk to threatened and endangered species**
 - 120 unique PFAS found in over 330 species on every continent except Antarctica (Source: EWG; <https://www.ewg.org/news-insights/news/2023/02/wildlife-warning-more-330-species-contaminated-forever-chemicals>)

Sites known or suspected of making, using or releasing PFAS (n=41,828)



Different colors points represent PFAS known users; suspected users; airports previously required to use AFFF; landfills and waste disposal facilities; and sewage and waste treatment plants.

Source: https://www.ewg.org/interactive-maps/2021_suspected_industrial_discharges_of_pfis/map/

Strengthening Exposure Science

- Provide tools for exposure assessment that keep pace with chemicals in commerce including chemicals designated CBI;
- Make assessment dynamic reflecting dynamic changes in manufacturing, use, behavior, etc.;
- Better account for human behaviors and co-exposures; and
- Strengthen toxicokinetic modeling

Vandenberg *et al.* *Environmental Health* 2023, **21**(Suppl 1):121
<https://doi.org/10.1186/s12940-022-00917-0>

Environmental Health

REVIEW

Open Access

Addressing systemic problems with exposure assessments to protect the public's health



Laura N. Vandenberg^{1*}, Swati D. G. Rayasam², Daniel A. Axelrad³, Deborah H. Bennett⁴, Phil Brown⁵, Courtney C. Carignan⁶, Nicholas Chartres², Miriam L. Diamond^{7,8}, Rashmi Joglekar^{9,10}, Bhavna Shamasunder¹¹, Kristin Shrader-Frechette^{12,13}, Wilma A. Subra¹⁴, Ken Zarker^{15^} and Tracey J. Woodruff²

Developing the Science: Computational Exposure

Recognition

- Exposure science is a complex endeavor that spans chemical, physical, biological, environmental, and social sciences
- The high-throughput need of computation exposure science greatly accentuates the complexity
- Data & models interdependent / highly integrated (NAS, 2012)
- Modeling provides the only practical means to achieve

Defining characteristics

- Predictive
- Rapid
- High-throughput
- Results in closure of knowledge/time gap supporting risk management decisions

Part of larger research effort known as NAMs

(Wambaugh et al. 2019)

The integration of advances in chemistry, computer science, mathematics, statistics, and social and behavioral sciences with new and efficient models and data collection methods to reliably and effectively forecast real-world exposures to natural and anthropogenic chemicals in the environment (Egeghy et al., 2016.)

Developing the Science: Computational Exposure

Recognition

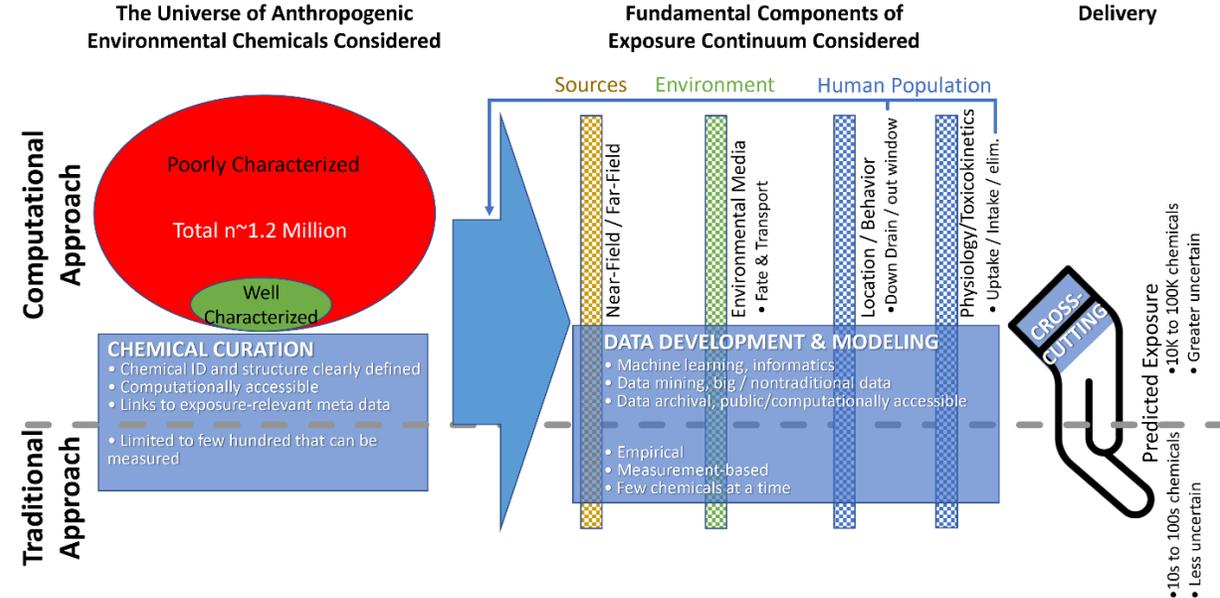
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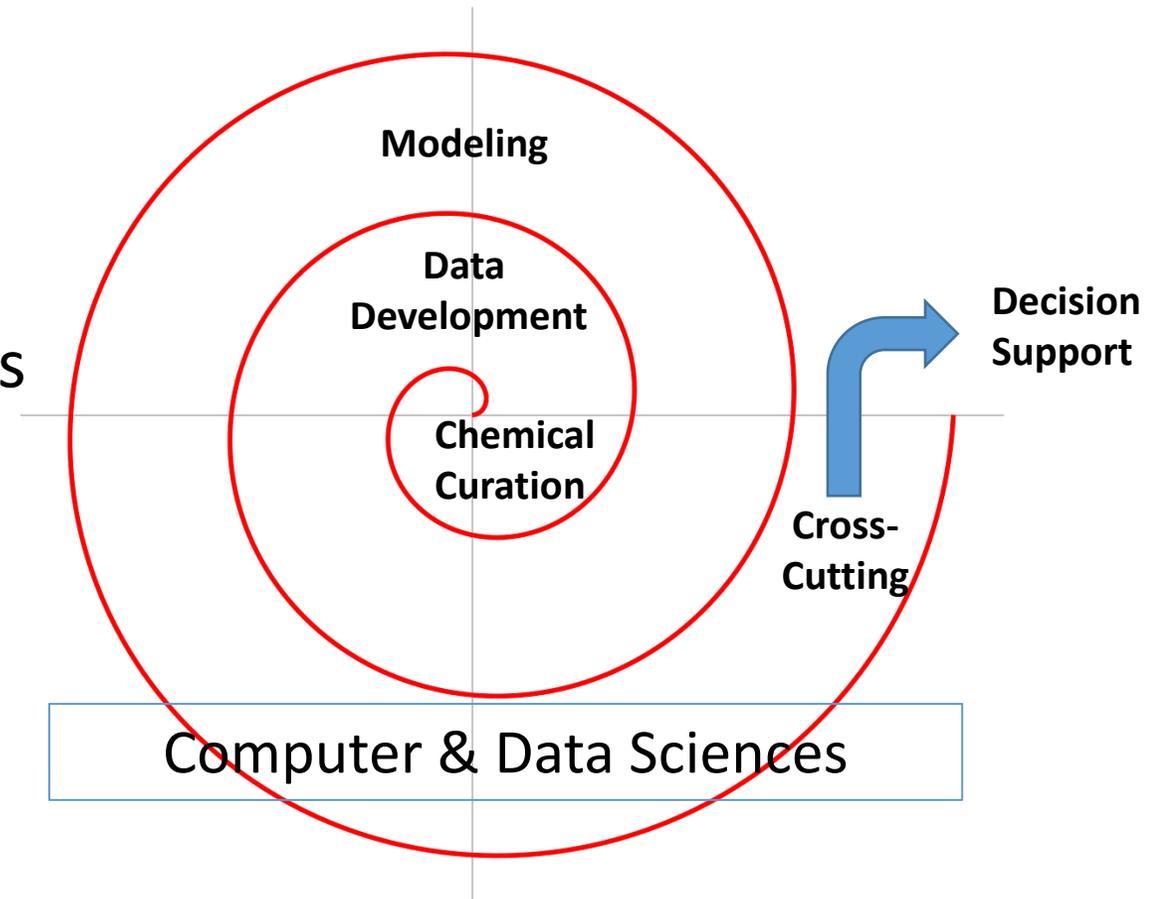
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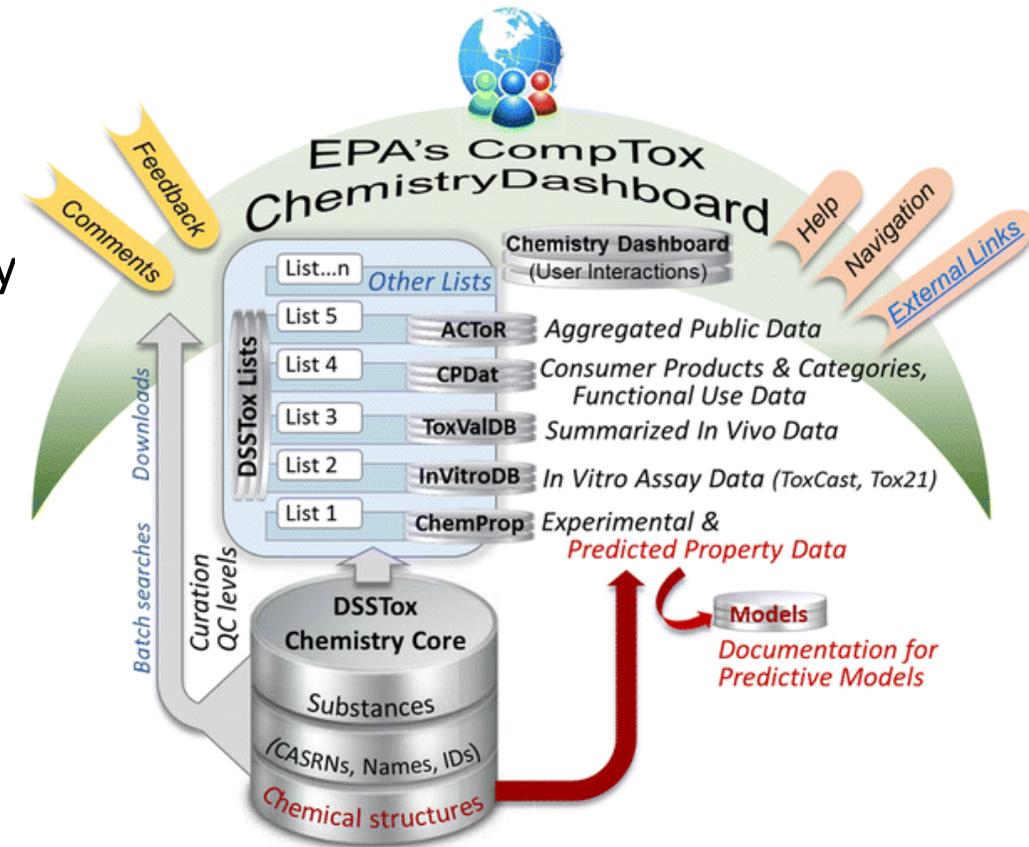
Computational Exposure Research Elements

- Chemical Curation
- Data Development
 - Data curation of public sources
 - Non-targeted Analysis (NTA) methods
- Modeling
- Cross-Cutting
 - Uncertainty
 - Confidence
 - Access



Chemical Curation

- Defining characteristics
 - Most comprehensive accounting of anthropogenic chemical landscape, i.e., 1.2 M chemicals
 - Chemical ID, structure, and meta data rigorously quality assured
 - Accessibility for computational, regulatory, public use
- Foundational resource serving to integrate and enable exposure elements as well as linkages to hazard and risk
- Future plans
 - Continued expansion to include polymers, mixtures, and ambiguously-defined substances
 - Continued expansion based on non-targeted analysis discovery



<https://comptox.epa.gov/dashboard/>

Research Example: Curation of Chemicals in Biosolids

CompTox Chemicals Dashboard Home Search Lists About Tools Submit Comments Search all data

Welcome to the new EPA CompTox Chemicals Dashboard

The new Dashboard is a complete rebuild and is replacing the CompTox Chemicals Dashboard released on July 12th 2020.

LIST: Chemicals in biosolids (2021)

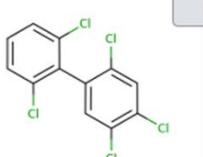
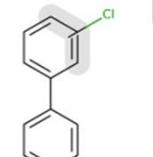
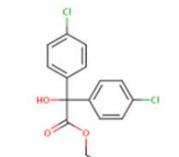
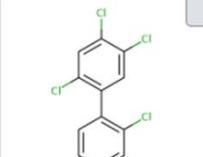
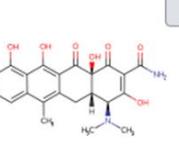
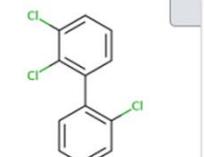
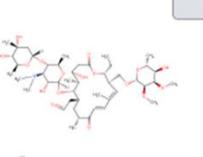
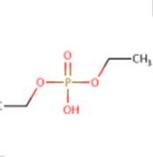
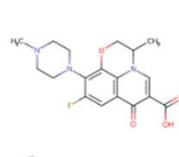
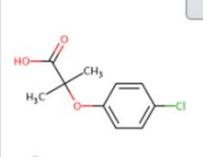
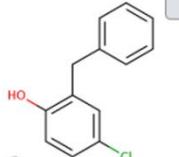
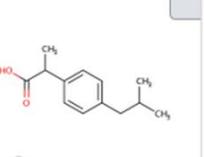
Search for chemical by systematic name, synonym, CAS number, DTXSID or InChIKey

Identifier substring search

List Details

Search Results SEND 726 TO BATCH SEARCH TILE INFO FILTER EXPORT

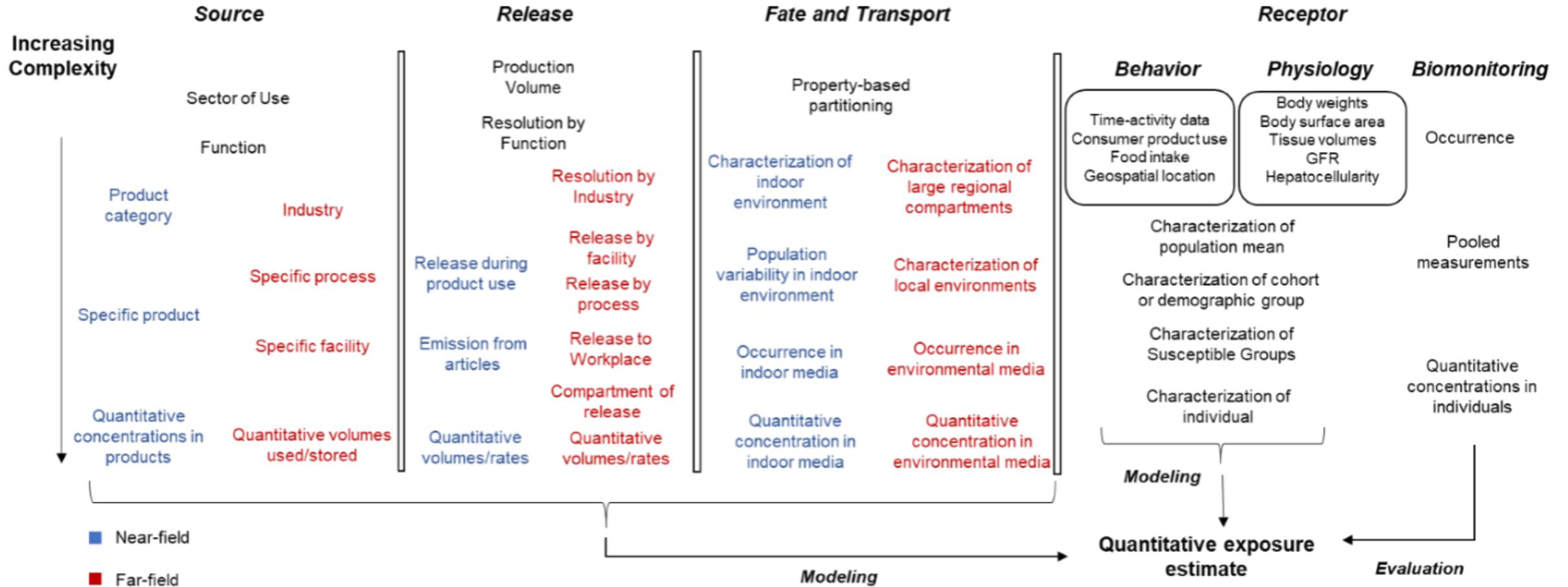
Showing 726 of 726 chemicals

 <p>2,2',4,5,6'-Pentachlorobiphenyl</p> <p>DTXSID : DTXSID10867525 CASRN : 68194-06-9 TOXCAST :</p>	 <p>Monochlorobiphenyls</p> <p>DTXSID : DTXSID401026566 CASRN : 27323-19-6 TOXCAST :</p>	 <p>Chlorobenzilate</p> <p>DTXSID : DTXSID09020209 CASRN : 510-15-6 TOXCAST : 247/919</p>	 <p>2,2',4,5-Tetrachlorobiphenyl</p> <p>DTXSID : DTXSID6074207 CASRN : 70362-47-9 TOXCAST :</p>	 <p>Anhydrotetracycline</p> <p>DTXSID : DTXSID0201016171 CASRN : 1665-56-1 TOXCAST :</p>	 <p>2,2',3-Trichlorobiphenyl</p> <p>DTXSID : DTXSID9073501 CASRN : 98444-79-9 TOXCAST :</p>
 <p>Tylosin</p> <p>DTXSID : DTXSID08439256 CASRN : 1421-69-0 TOXCAST :</p>	 <p>Diethyl hydrogen phosphite</p> <p>DTXSID : DTXSID10444599 CASRN : 599-02-7 TOXCAST :</p>	 <p>Ofloxacin</p> <p>DTXSID : DTXSID09041085 CASRN : 92419-36-1 TOXCAST : 1/285</p>	 <p>Clofibric acid</p> <p>DTXSID : DTXSID10405651 CASRN : 992-09-7 TOXCAST : 11/409</p>	 <p>Clorophene</p> <p>DTXSID : DTXSID08020184 CASRN : 120-92-1 TOXCAST : 981/1250</p>	 <p>Ibuprofen</p> <p>DTXSID : DTXSID05020792 CASRN : 15687-27-1 TOXCAST : 27/650</p>

- Clarity of candidate pollutants relevant to Clean Water Act
- Important implications for human exposure & risk
- Replaces standalone biennial reports
- Provides context for interpreting risk
- List includes 726 chemicals / concentrations for 484

Richman et al., 2022

Data Development Framework



Data Development

- Exposure continuum provides framework
- Consistent with NAS guidance for organization, useability, and access
- Includes advanced
 - Informatics, data-mining, machine learning
 - Data infrastructure for collection, organizing, and integration
- Receptor oriented / near field data especially valuable
- Exposure Databases developed
 - ChemExpoDB , CPDat (Dionisio et al., 2018), MMDB (Isaacs et al., 2022), and CvTdb (Sayre et al., 2020)
- Future research to address occupational settings, chemicals in consumer articles

Example Data Development Research Effort

- Informatics approaches and a custom document management/curation application are being used to obtain and curate chemical use descriptor information from thousands of public documents
- Consumer product ingredient data (composition), functional use information, general chemical use keywords
 - **600K+ products**
 - **>35K unique chemicals**
 - **>45K functional use records (91 unique functions)**
- Extraction and curation of chemical and product data via reproducible script-based methods (including natural language processing) or manual tools
- QA/audit trail tools
- Data released to CompTox Chemicals Dashboard and in bulk as CPDat
- Data also forms basis for Quantitative Structure-Use Relationships (QSURs)

500K+ Raw Public Documents

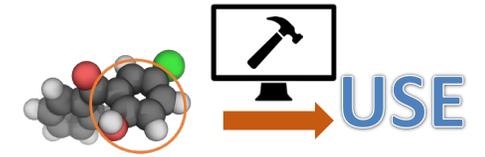
Dionisio et al. Sci Data 5:180125 (2018).

<https://comptox.epa.gov/dashboard/>

Custom Curation Application

CompTox Chemicals Dashboard
Search 1,200,060 Chemicals

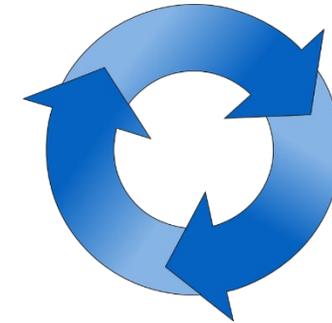
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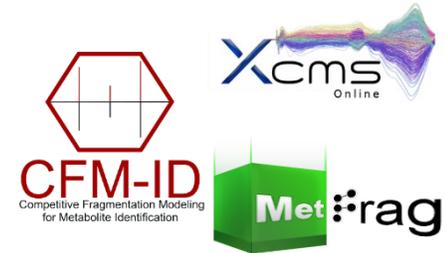
Non-Targeted Analysis (NTA)

- Identifies chemicals without *a priori* knowledge or standards
- Applications → source to dose
 - Source: household (& recycled products)
 - Environmental: house dust & water filters
 - Dose: blood & placenta
- Research focused on obstacles to broad adoption (Ulrich et al., 2019)
 - Methods development
 - Workflows
 - Web resources
 - Enhancing transparency & reproducibility
 - Building communities of practice

Analytical Instruments



Comp. Tools & Workflows



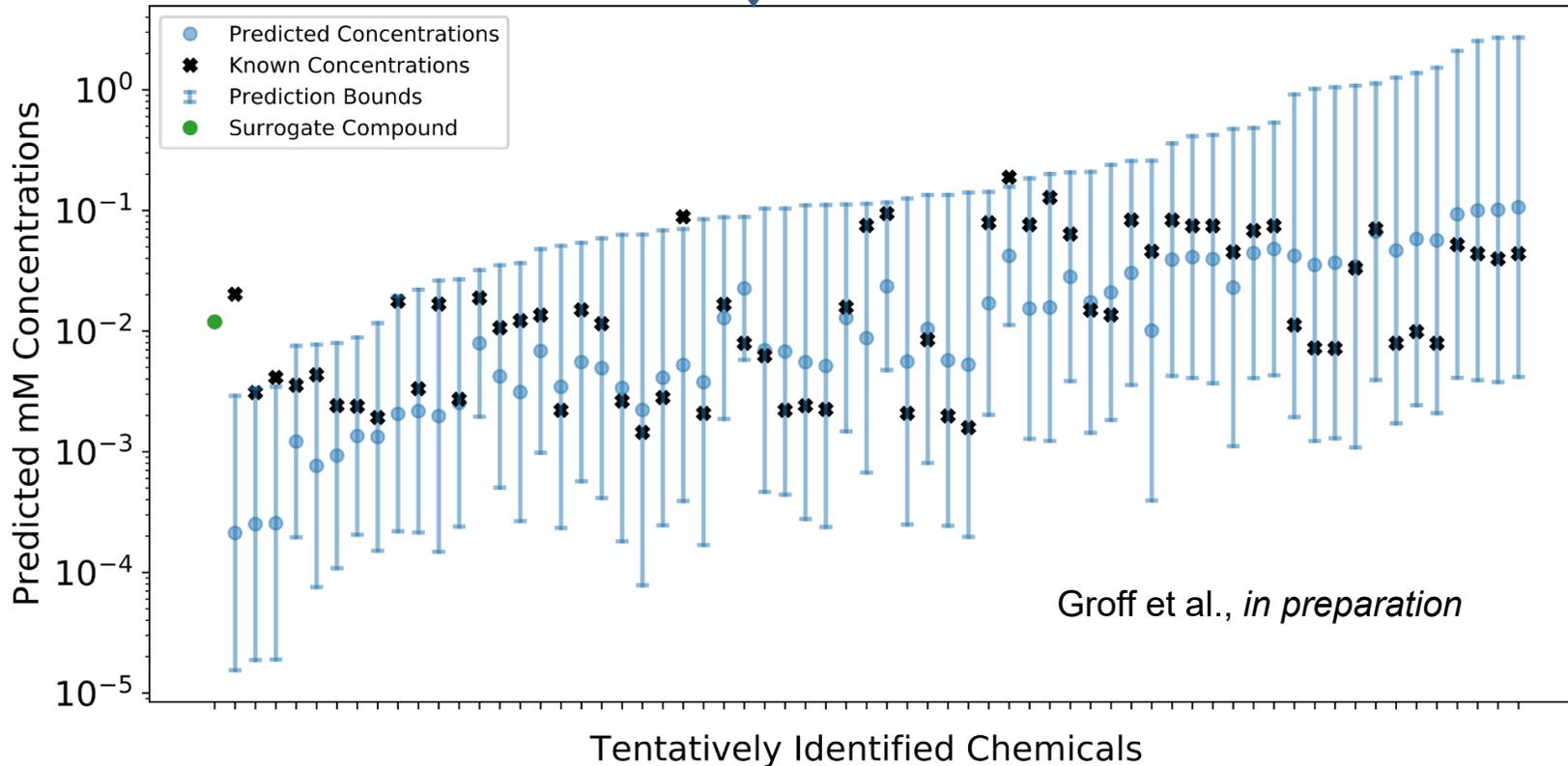
Chemical Database



Example NTA Research



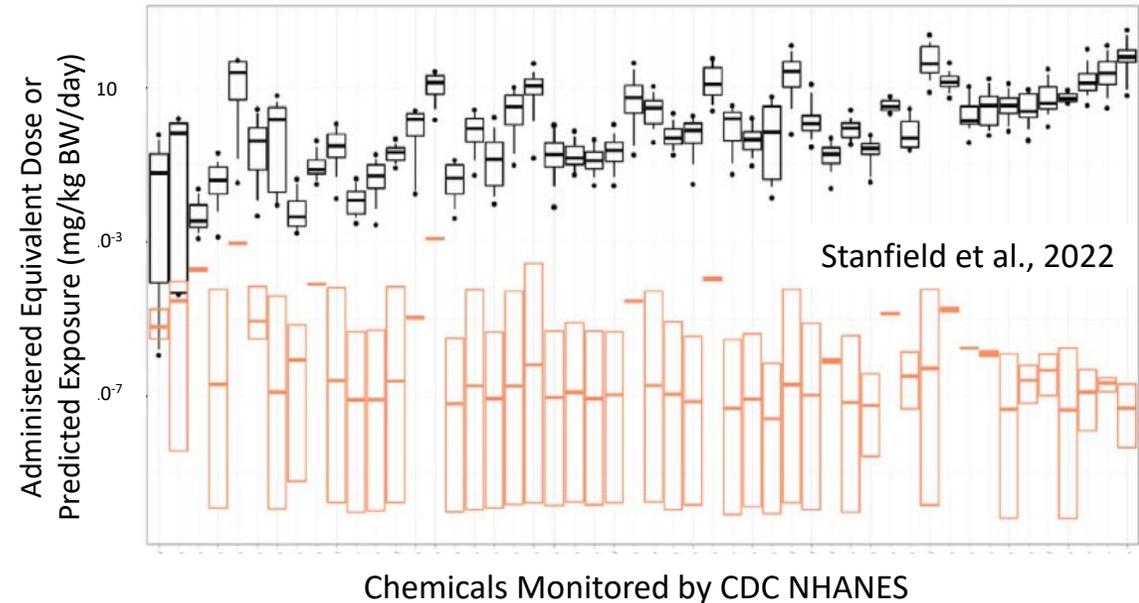
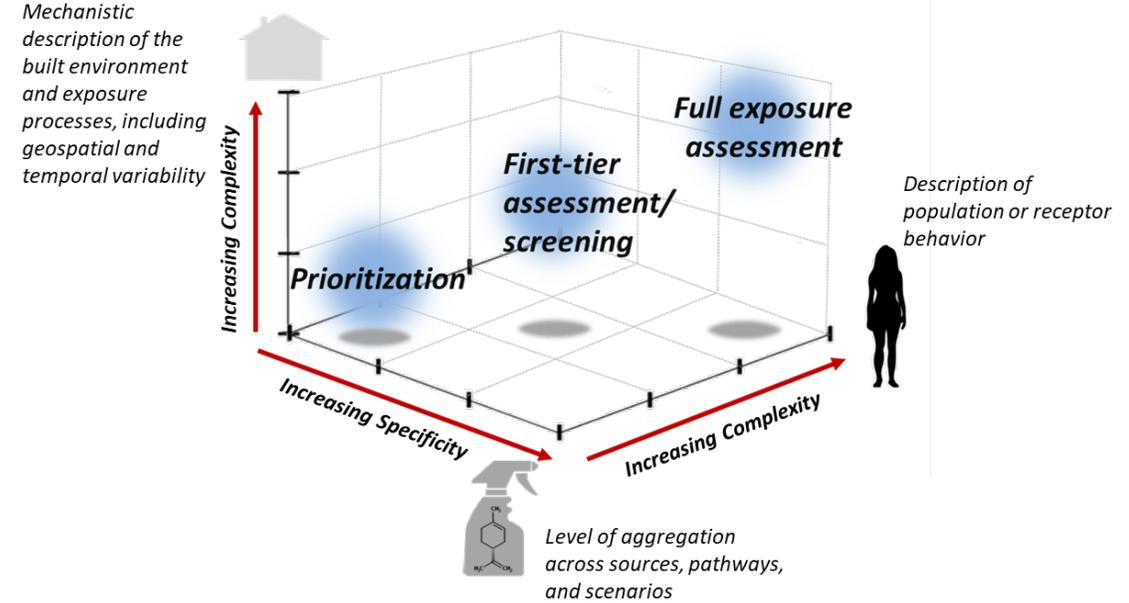
qNTA Proof-of-Concept



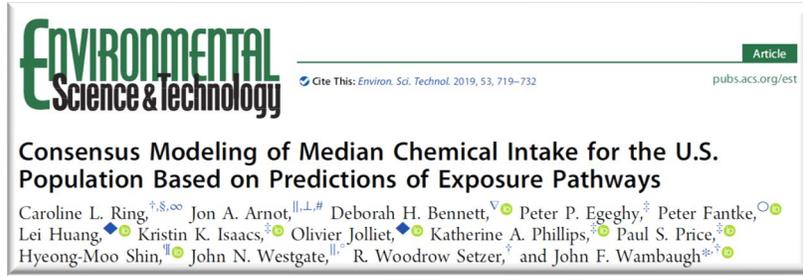
- Analysis of Brita filter extracts via GC-HRMS
- Single surrogate selected and applied to all identified analytes
- Concentration estimates can be above or below true value
- Prediction intervals used to bound concentration estimates
- % prediction intervals shown; Can use 95%, 99%, 99.9%, etc.
- Tentatively identified compounds ranked by upper-bound estimates
- **Upper-bound estimates compared to level-of-interest to set priorities**

Modeling

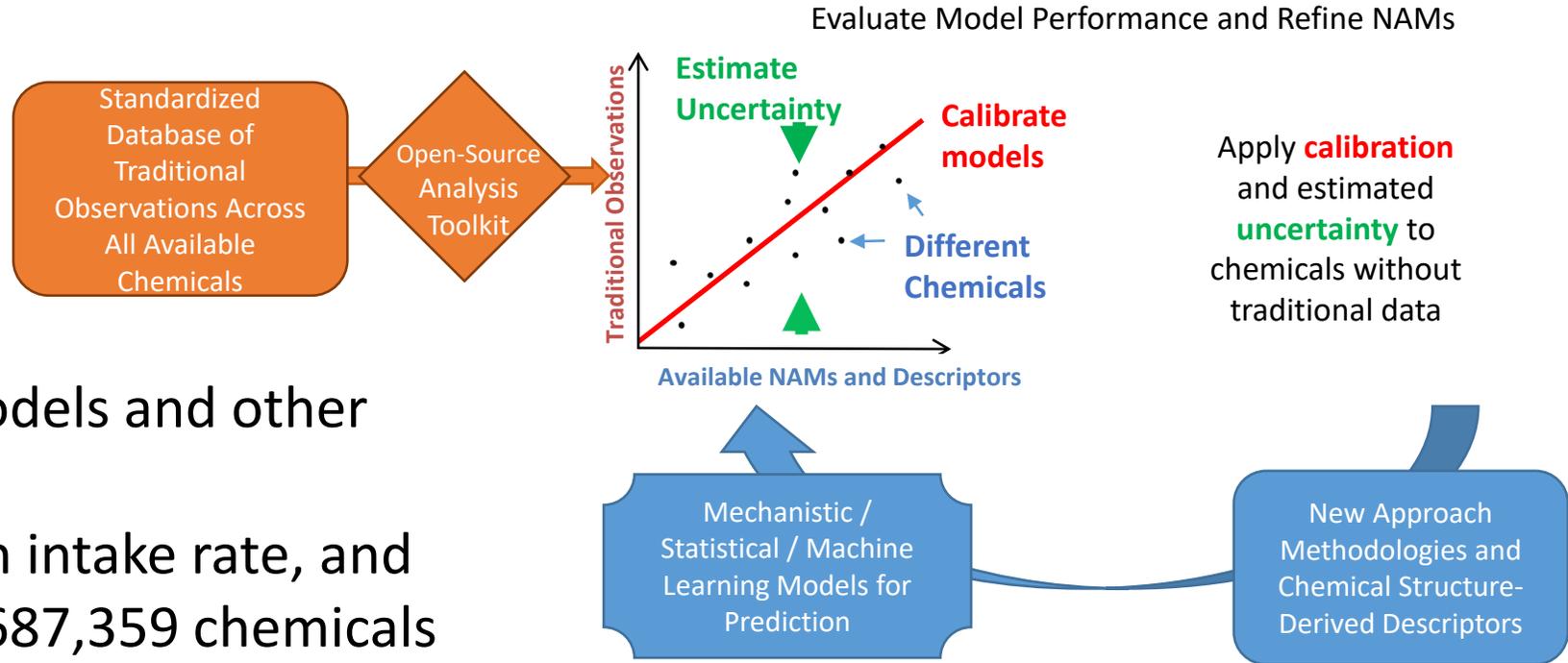
- Source-to-receptor exposure modeling is the ultimate application and integration of computational exposure elements
- Systematic Empirical Evaluation of Models (SEEM) is a consensus modeling approach for intake rate (mg/kg/day) (Wambaugh et al., 2013)
- NHANES biomonitoring key: Uncertainty quantified and balanced with data inputs / model parameterization
- Intake rates have been estimated for hundreds of thousands chemicals – mostly data poor
- Complementary toxicokinetics modeling is a key consideration allowing linkage between exposure estimates and *in vitro* bioactivity data (Breen et al., 2021)



Example Research



- Consensus of 12 different models and other exposure predictors
- Exposure pathway(s), median intake rate, and credible interval for each of 687,359 chemicals with structures available from the CompTox Chemicals Dashboard are estimated
 - 30% low probability for exposure via any of the four pathways
 - 95% confidence that the median intake rate is below 1 $\mu\text{g}/\text{kg BW}/\text{day}$ for 474,572 compounds



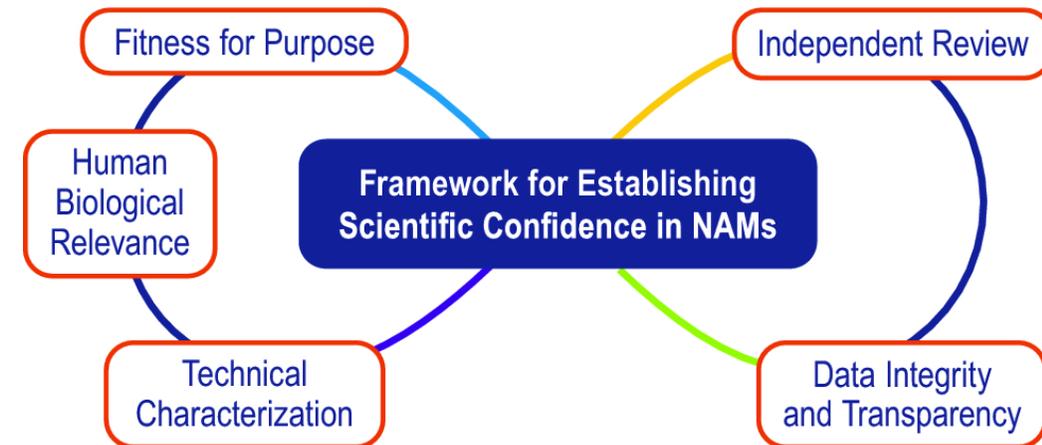
Cross-Cutting

Ensure methods and approaches produce results:

- Accessible
- Fit for purpose
- Effectively communicated
- Transparent
- Reproducible
- Inform health-protective risk assessment
- Uncertainty & variability captured
- Garner confidence

Consistent with NAS (2017) Using 21st Century Science to Improve Risk-Related Evaluations

Analyses should be carried out in transparent and replicable ways to ensure credibility and to enhance review and acceptance of findings for decision-making. Open data access might be critical for ensuring transparency.



Summary

- Exposure science is rapidly evolving from its industrial hygiene roots
 - Expanding scientific scope
 - Responding to wicked/complex environmental challenges
- Chemical exposure presented as case example
 - Habitable planet boundary exceeded
 - Number of chemicals in commerce vastly out-paces traditional approaches to assessment
- EPA developing computational exposure science strategies to address challenge
- Integrated elements to computational exposure
 - Chemical curation
 - Data development / non-targeted analysis
 - Modeling
 - Build confidence
- Delivering
 - Rapid exposure estimates at scale of chemicals in commerce
 - Data and model estimates that are transparent, accessible, quantified uncertainty/variability
 - Integrated workflow with hazard for high throughput risk estimation



References

- Breen M et al. High-throughput PBTK models for *in vitro* to *in vivo* extrapolation. *Expert Opin Drug Metab Toxicol*. 2021 Aug;17(8):903-921. doi: 10.1080/17425255.2021.1935867. Epub 2021 Jun 15. PMID: 34056988.
- Cousins IT, et al. Outside the Safe Operating Space of a New Planetary Boundary for Per- and Polyfluoroalkyl Substances (PFAS). *Environ Sci Technol*. 2022 Aug 16;56(16):11172-11179. doi: 10.1021/acs.est.2c02765. Epub 2022 Aug 2. PMID: 35916421;
- Dionisio KL et al., The Chemical and Products Database, a resource for exposure-relevant data on chemicals in consumer products. *Sci Data*. 2018 Jul 10;5:180125. doi: 10.1038/sdata.2018.125. PMID: 29989593; PMCID: PMC6038847.
- Groff LC et al. Uncertainty estimation strategies for quantitative non-targeted analysis. *Anal Bioanal Chem*. 2022 Jul;414(17):4919-4933. doi: 10.1007/s00216-022-04118-z. Epub 2022 Jun 14. PMID: 35699740.
- Isaacs KK, et al. A harmonized chemical monitoring database for support of exposure assessments. *Sci Data*. 2022 Jun 16;9(1):314.
- Landrigan PJ et al.: The *Lancet* Commission on pollution and health. *The Lancet*. 2017 [http://dx.doi.org/10.1016/S0140-6736\(17\)32345-0](http://dx.doi.org/10.1016/S0140-6736(17)32345-0)
- Lowe CN et al. "Chemical Characterization of Recycled Consumer Products Using Suspect Screening Analysis." *Environmental science & technology* 55.16 (2021): 11375-11387. NAS (2012) The National Academies Press. <https://doi.org/10.17226/13507>.
- Newton SR et al. "Suspect screening and non-targeted analysis of drinking water using point-of-use filters." *Environmental pollution* 234 (2018): 297-306.
- Persson LM et al. Confronting unknown planetary boundary threats from chemical pollution. *Environ Sci Technol*. 2013 Nov 19;47(22):12619-22. doi: 10.1021/es402501c. Epub 2013 Sep 16. PMID: 23980998.
- Persson L et al. Outside the Safe Operating Space of the Planetary Boundary for Novel Entities. *Environ Sci Technol*. 2022 Feb 1;56(3):1510-1521. doi: 10.1021/acs.est.1c04158. Epub 2022 Jan 18. PMID: 35038861
- Phillips KA (2018). Suspect screening analysis of chemicals in consumer products. *Environmental science & technology*, 52(5), 3125-3135.
- Rager JE et al. "Review of the environmental prenatal exposome and its relationship to maternal and fetal health." *Reproductive Toxicology* 98 (2020): 1-12.
- Rappaport SM. Genetic Factors Are Not the Major Causes of Chronic Diseases. *PLoS One*. 2016 Apr 22;11(4)
- Richman T et al. Curation of a list of chemicals in biosolids from EPA National Sewage Sludge Surveys & Biennial Review Reports. *Sci Data* 9, 180 (2022). <https://doi.org/10.1038/s41597-022-01267-9>
- Ring CL et al. (2019). Consensus modeling of median chemical intake for the US population based on predictions of exposure pathways. *Environmental science & technology*, 53(2), 719-732.
- Rockström J et al. A safe operating space for humanity. *Nature*. 2009 Sep 24;461(7263):472-5. doi: 10.1038/461472a. PMID: 19779433.
- Sayre RR et al. Database of pharmacokinetic time-series data and parameters for 144 environmental chemicals. *Sci Data* 7, 122 (2020). <https://doi.org/10.1038/s41597-020-0455-1>
- Stanfield Z et al. Bayesian inference of chemical exposures from NHANES urine biomonitoring data. *J Expo Sci Environ Epidemiol*. 2022 Aug 17. doi: 10.1038/s41370-022-00459-0. Epub ahead of print. PMID: 35978002.
- Ulrich EM et al. EPA's non-targeted analysis collaborative trial (ENTACT): genesis, design, and initial findings. *Anal Bioanal Chem*. 2019 Feb;411(4):853-866. doi: 10.1007/s00216-018-1435-6. Epub 2018 Dec 6. PMID: 30519961; PMCID
- van der Zalm AJ et al. A framework for establishing scientific confidence in new approach methodologies. *Arch Toxicol*. 2022 Aug 20. doi: 10.1007/s00204-022-03365-4. Epub ahead of print. PMID: 35987941.
- Wambaugh JF et al. 2019 New approach methodologies for exposure science, *Current Opinion in Toxicology*, Volume 15, 2019, Pages 76-92, ISSN 2468-2020, <https://doi.org/10.1016/j.cotox.2019.07.001>.
- Wambaugh JF et al. High-throughput models for exposure-based chemical prioritization in the ExpoCast project. *Environ Sci Technol*. 2013 Aug 6;47(15):8479-88. doi: 10.1021/es400482g. Epub 2013 Jul 11. PMID: 23758710.
- Wang Z et al. Toward a Global Understanding of Chemical Pollution: A First Comprehensive Analysis of National and Regional Chemical Inventories. *Environ Sci Technol*. 2020 Mar 3;54(5):2575-2584. doi: 10.1021/acs.est.9b06379. Epub 2020 Feb 14. PMID: 31968937.

Thank you!

QUESTIONS / COMMENTS