

Development of Mathematical New Approach Methods to Assess Chemical Mixtures

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Overview

<u>Problem</u>

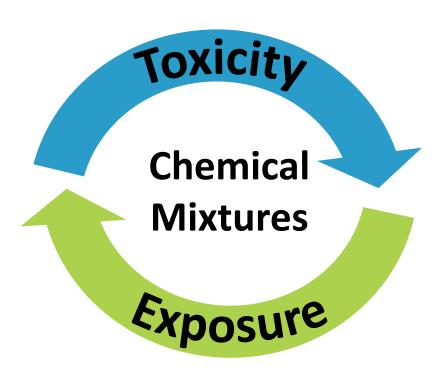
- Real-life chemical exposures are mixtures
- Screening all possible mixture combinations is impractical

Part 1: Mixture Modeling Approach

- Combine single chemical component bioactivity data using mathematical models to predict mixture response
- Evaluate with an experimental binary mixtures dataset

Part 2: MOE Case Study

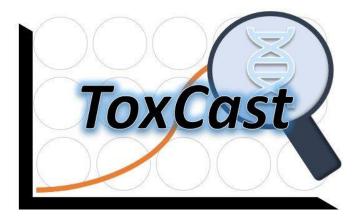
- Examine Margin of Exposure (MOE) for a PFAS subset
- Compare predicted bioactivity from mixture models (Part 1) to human biomonitoring exposure data



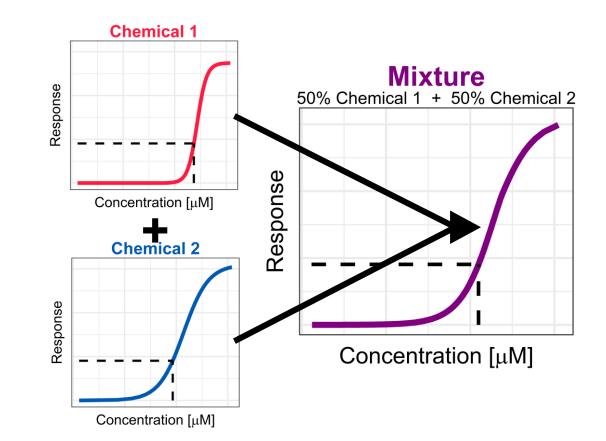
Part 1: Mixture Modeling with Experimental Concentration-Response Data

Project Objectives

1) Use available single chemical screening assay data from the U.S. EPA's Toxicity Forecaster (ToxCast) program

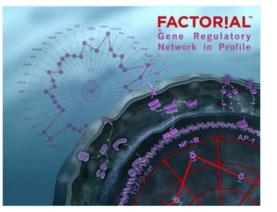


- 2) Predict bioactivity behavior of binary chemical mixtures with mathematical mixture models
- 3) Evaluate performance of predicted model fits compared to experimental data



Test Dataset

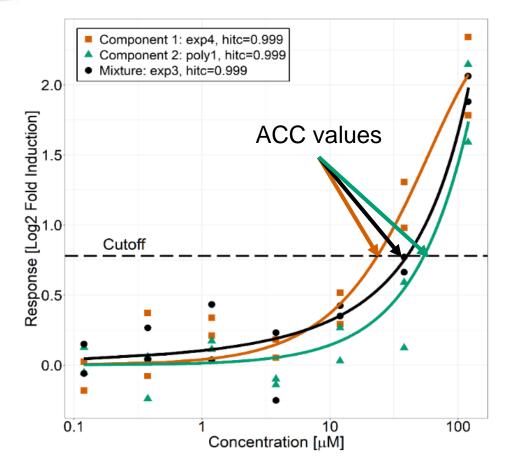
- Selected **21** binary chemical mixtures and their single components for screening
 - Inspired by consumer product ingredients
 - Included single chemical components with legacy
 ToxCast screening data
 - Used fixed concentration ratios (either 1:1 or 1:2)
- Mixtures and additional samples of single chemical components were screened in concentration-response in the Attagene FACTORIALTM platform, profiling **81** transcription factors and nuclear receptor targets
- A filtered subset of **237** active mixture concentration-response curves were considered for this analysis



Binary Mixtures (Concentration Ratio) 1) Propylparaben; 2-(Phenylmethylene)octanal (33:67) 2) Propylparaben; 2-(Phenylmethylene)octanal (67:33) 3) Propylparaben; Butylated hydroxytoluene (33:67) 4) Propylparaben; Butylated hydroxytoluene (67:33) 5) 2-(Phenylmethylene)octanal; Butylated hydroxytoluene (33:67) 6) 2-(Phenylmethylene)octanal; Butylated hydroxytoluene (67:33) 7) Citric acid; Butylated hydroxytoluene (33:67) 8) Citric acid; Butylated hydroxytoluene (67:33) 9) Dodecanoic acid; Propylparaben (67:33) 10) Dodecanoic acid; Propylparaben (33:67) 11) 1-Phenyl-3-methyl-5-pyrazolone; Propylparaben (33:67) 12) 1-Phenyl-3-methyl-5-pyrazolone; Propylparaben (67:33) 13) Propylparaben; Bisphenol A (50:50) 14) Propylparaben; Triclosan (50:50) 15) Propylparaben; Rifampicin (50:50) 16) 2-(Phenylmethylene)octanal; Bisphenol A (50:50) 17) 2-(Phenylmethylene)octanal; Triclosan (50:50) 18) 2-(Phenylmethylene)octanal; Pioglitazone hydrochloride (50:50) 19) 2-(Phenylmethylene)octanal; Bexarotene (50:50) 20) Citric acid; Bexarotene (50:50) 21) Butylated hydroxytoluene; Bexarotene (50:50)

Stanfield, Z. et al. Mining of consumer product ingredient and purchasing data to identify potential chemical coexposures. Environmental Health Perspectives. 2021 Jun; 129(6). PMID: <u>34160298</u>

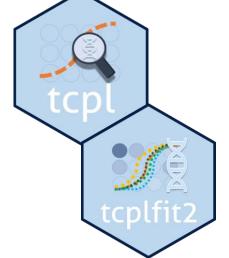
Concentration-Response Data



Mixture: 0.67 2-(Phenylmethylene)octanal + 0.33 Butylated hydroxytoluene **Endpoint:** Nuclear factor (erythroid-derived 2)-like 2 (ATG NRF2 ARE CIS)

- Example Binary Mixture consists of 67% Chemical Component 1 and 33% Chemical Component 2
- Key Point of Departure (POD) used was the Activity Concentration at the Cutoff (ACC) since it allows comparability across endpoints
- Data is available in ToxCast's invitrodb v4.2, fit with ToxCast Pipeline software *tcpl* v3.2.0 and *tcplfit2* v0.1.7.





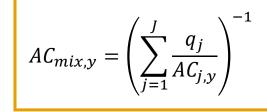
U.S. EPA. 2024. ToxCast from invitrodb_v4.2. Retrieved from https://www.epa.gov/chemical-research/toxicity-forecaster-toxcasttm-data on Oct. 1, 2024.

Mathematical Mixture Models

Extrapolated Concentration Addition (CA)

- Assumes chemicals act in the same way on the same target
- Input = concentrations at given response levels [x(y)]

for $y < 0.7 * top_2$:



Independent Action (IA)

- Assumes chemicals act independently through different pathways to reach an apical response
- Treats responses as probabilities
- Input = response curves for given concentrations [y(x)]

$$E_{mix}(c_{mix}) = 1 - \prod_{j=1}^{J} \left(1 - E_j (q_j \cdot c_{mix}) \right)$$

for $0.7 * top_2 < y < top_1$:

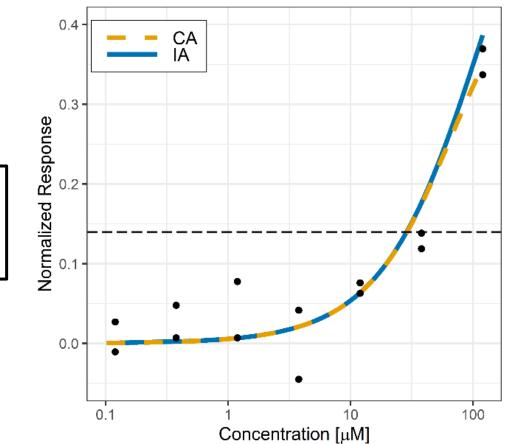
$$AC_{mix} = \frac{AC_{1,y}}{q_1} \cdot (1 - TU_2)$$

Concentration fraction $q_{j} = \frac{c_{j,y}}{\sum_{j=1}^{J} c_{j,y}}$

Model Inputs:

- 1) Data for single components
 - Compare component data from test
 - dataset vs. from legacy ToxCast database
- 2) Concentration fractions of the mixture

0.67 2-(Phenylmethylene)octanal + 0.33 Butylated hydroxytoluene



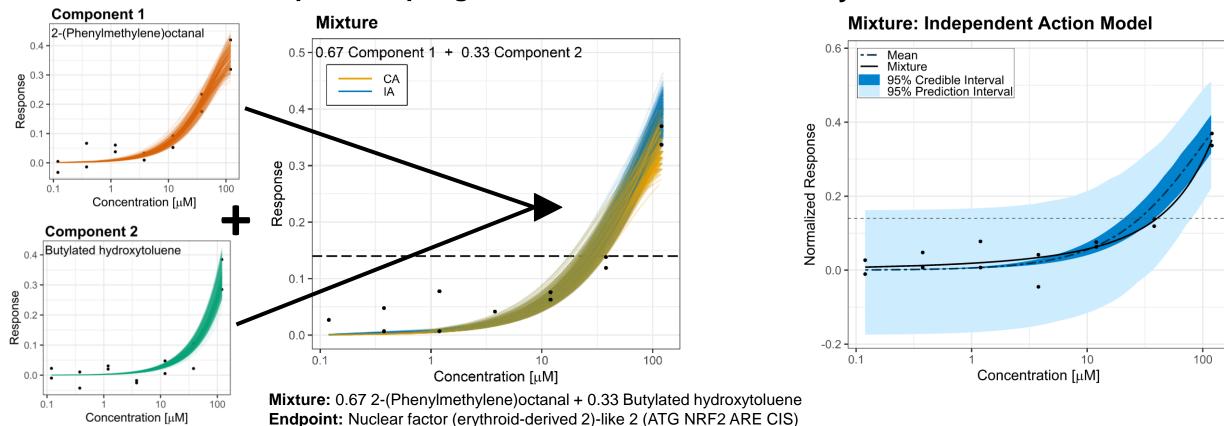
Scholze, Martin et al. "Extending the applicability of the dose addition model to the assessment of chemical mixtures of partial agonists by using a novel toxic unit extrapolation method." *PloS one* vol. 9,2 (2014). Backhaus, Thomas et al. "Predictability of the toxicity of a multiple mixture of dissimilarly acting chemicals to *Vibrio fischeri." Environmental Toxicology and Chemistry* vol. 19,9 (2000).

Capturing Uncertainty

Uncertainty from experimental mixture concentration-response data

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Uncertainty from single chemical concentration-response curves input to the models



Bootstrap Resampling

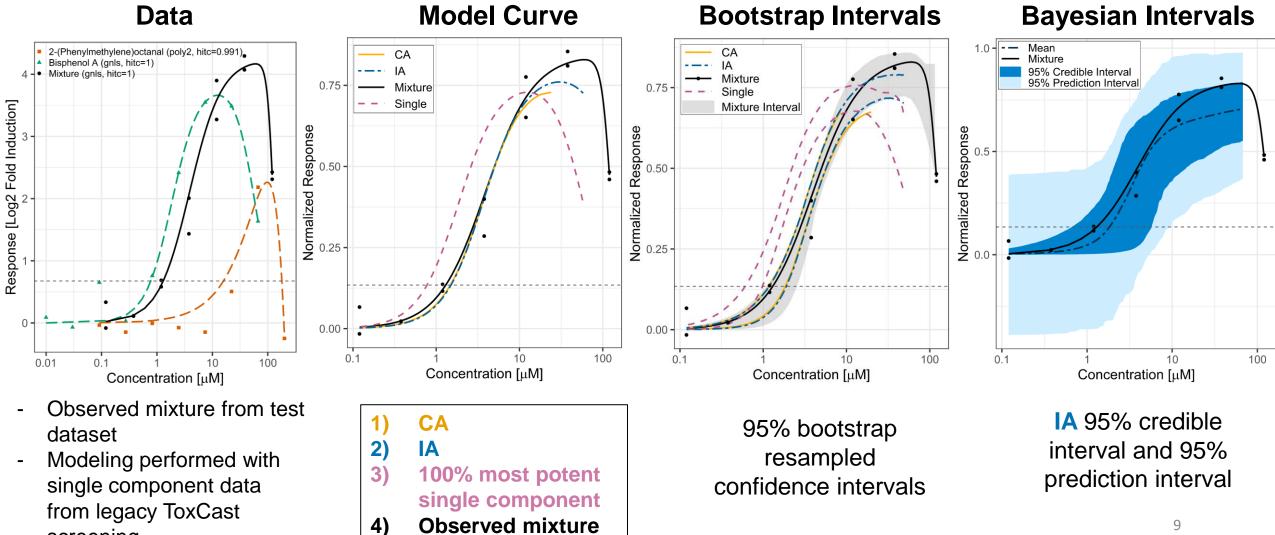
Watt E.D., Judson R.S. (2018) Uncertainty quantification in ToxCast high throughput screening. PLoS ONE 13(7): e0196963.

Bayesian Statistical Framework

Example Result

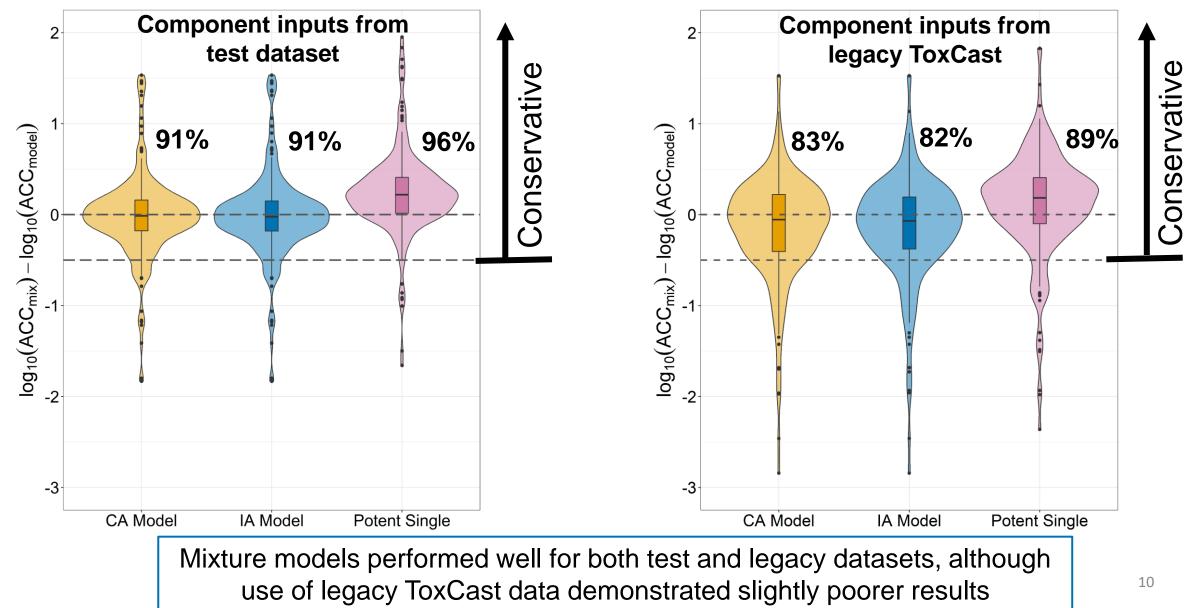
screening

Mixture: 0.5 2-(Phenylmethylene)octanal + 0.5 Bisphenol A **Endpoint:** Pregnane X receptor (ATG PXRE CIS)



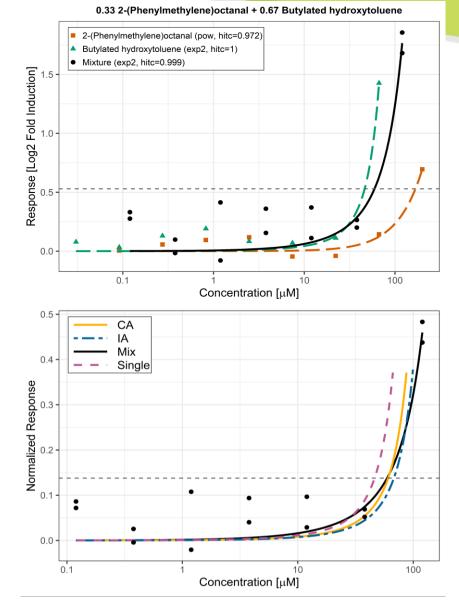
Performance Metrics

Comparison of ACC predictions from each model to the experimentally-derived ACC for the binary mixture



Part 1 Summary

- Concentration Addition (CA), Independent Action (IA), and 100% single component models were evaluated for prediction of bioactivity of binary mixtures
- For this chemical set, the CA and IA models performed similarly and captured the mixture behavior, especially at the ACC
 - Models do not include non-additive behaviors
 - Most potent single chemical model was usually conservative, but not as accurate
- Model inputs of single component data from the test dataset versus legacy ToxCast data were compared to predict the same mixtures
 - Both approaches performed well, but using legacy data provided slightly poorer predictions due to more sources of uncertainty



Part 2: Margin of Exposure Case Study

Exposure Dataset

- Examined the 2009 2010 CDC National Health and Nutrition Examination Survey (NHANES) for exposure data
 - Included laboratory blood serum samples from human subjects, specifically polyfluoroalkyl (PFAS) chemical concentrations for 2233 subjects
 - Analyzed via the survey package in R

Chemicals Evaluated:

PFOS = Perfluorooctane sulfonic acid

PFNA = Perfluorononanoic acid

PFOA = Perfluorooctanoic acid

PFHxS = Perfluorohexane sulfonic acid



Centers for Disease Control and Prevention (CDC). National Center for Health Statistics (NCHS). National Health and Nutrition Examination Survey Data. 13 Hyattsville, MD: U.S. Department of Health and Human Services, Centers for Disease Control and Prevention, (2009-2010) https://wwwn.cdc.gov/Nchs/Nhanes/.

MOE Calculation

Margin of Exposure (MOE) for an individual was computed as:

Activity Concentration at the Cutoff (ACC)

Exposure Concentration

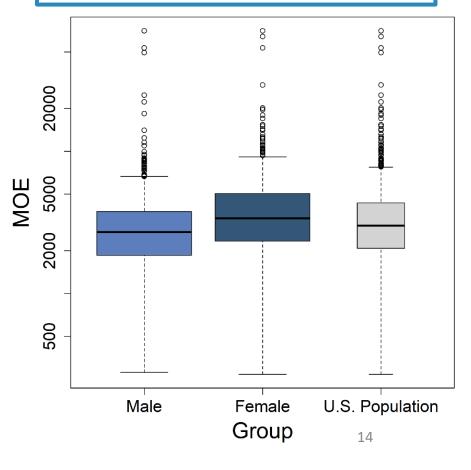
NHANES reported concentration in subject's blood serum [µmol/L]

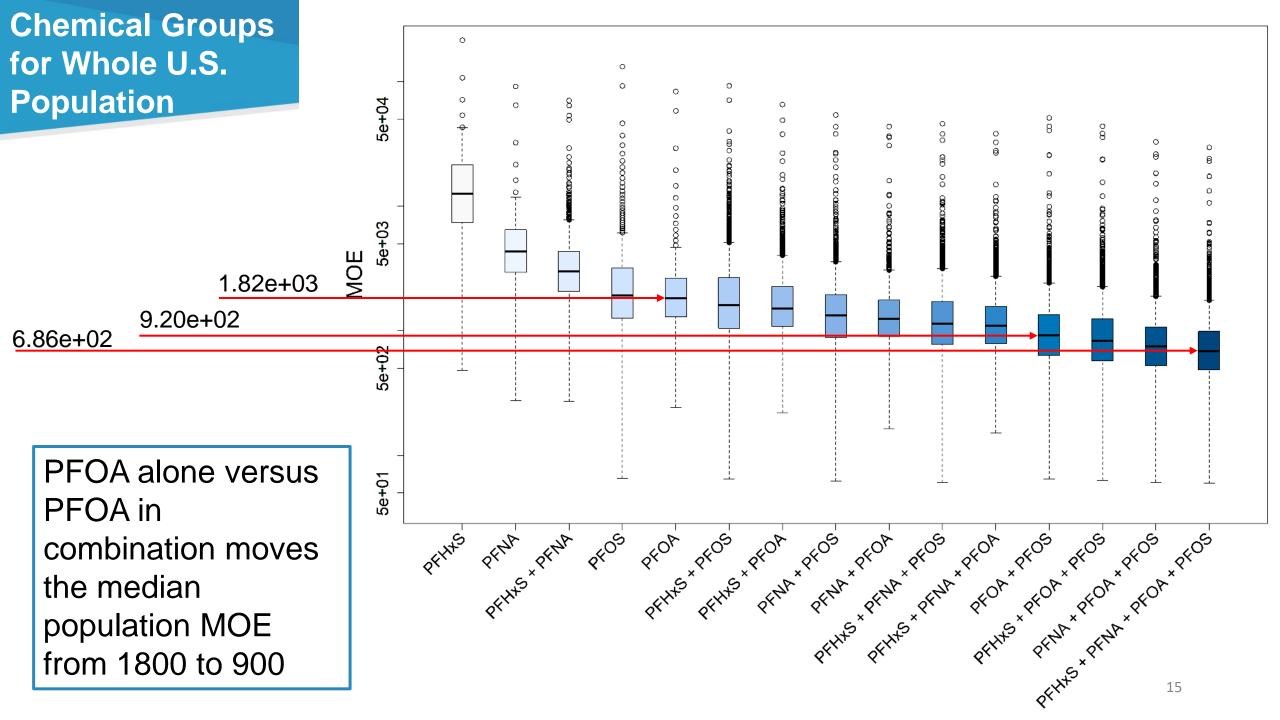
Values may be interpreted as follows:

$$\checkmark$$
 ACC + \uparrow Exposure \rightarrow \checkmark MOE \rightarrow \uparrow Bioactivity-based estimate of risk

Addicks, Gregory C. et al. "Per- and polyfluoroalkyl substances (PFAS) in mixtures show additive effects on transcriptomic points of departure in human liver spheroids", *Toxicological Sciences*, vol. 194, 1, (2023)

Simulated from Concentration Addition mixture model with single component inputs from ToxCast [µmol/L]





MOE Analysis

 MOE's derived considering the bioactivity ACC for one endpoint, peroxisome proliferator-activated receptor-α (ATG_PPARa_TRANS) MOE

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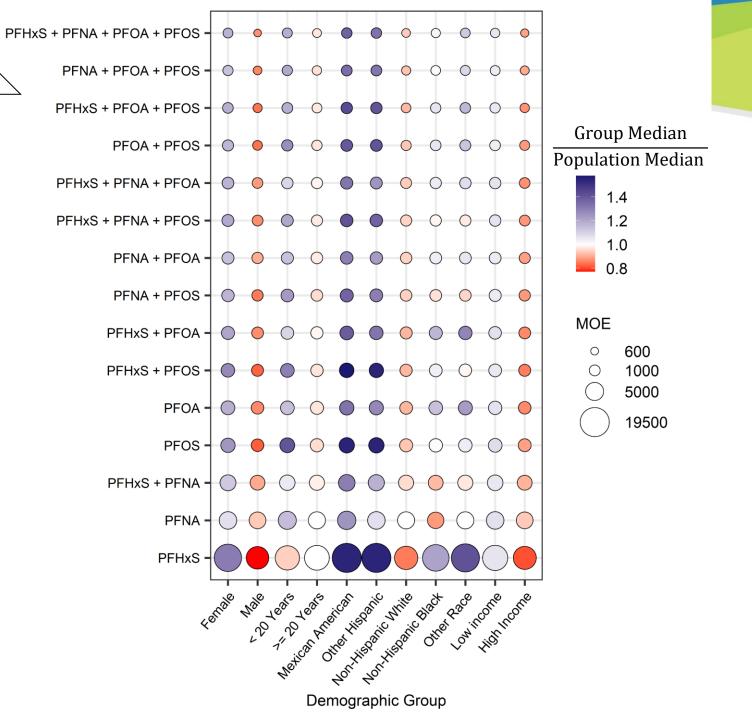
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- Red Circles may indicate where population median MOE may not be protective for that group
- Note: Demographic groups overlap subjects

Legend Circle Size: Median MOE Color: Ratio of demographic group median to population median



Demographic Trends

 MOEs derived considering the bioactivity ACC for one endpoint, peroxisome proliferator-activated receptor-α (ATG_PPARa_TRANS) MOE

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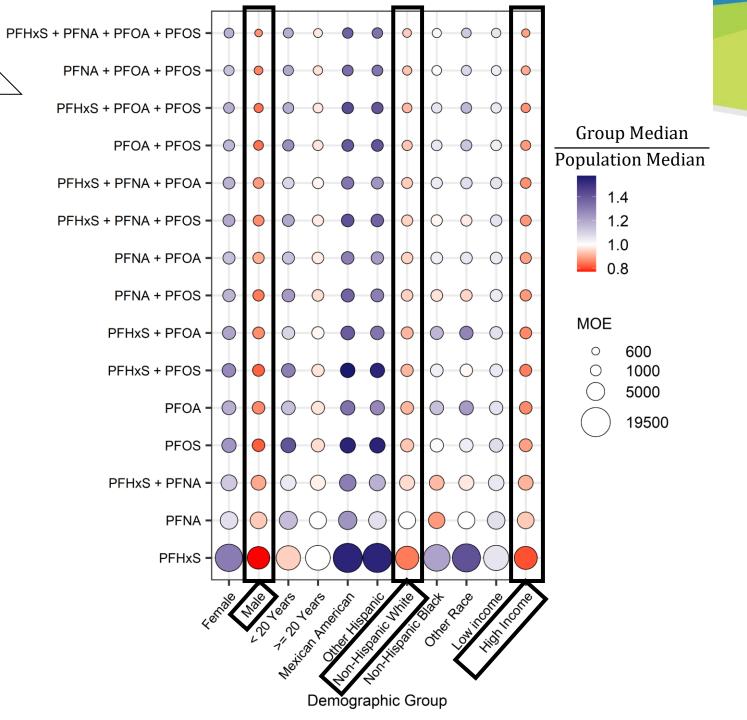
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Legend Circle Size: Median MOE Color: Ratio of demographic group median to population median



Chemical Groups

- **PFOA and PFOS have** lower MOE values
- Beyond single chemicals, the MOE values are similar

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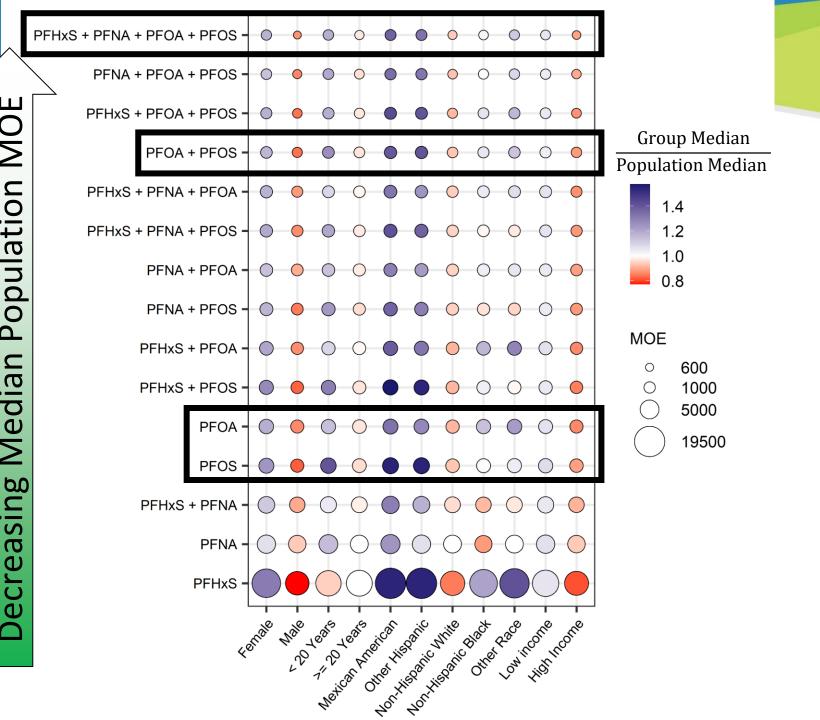
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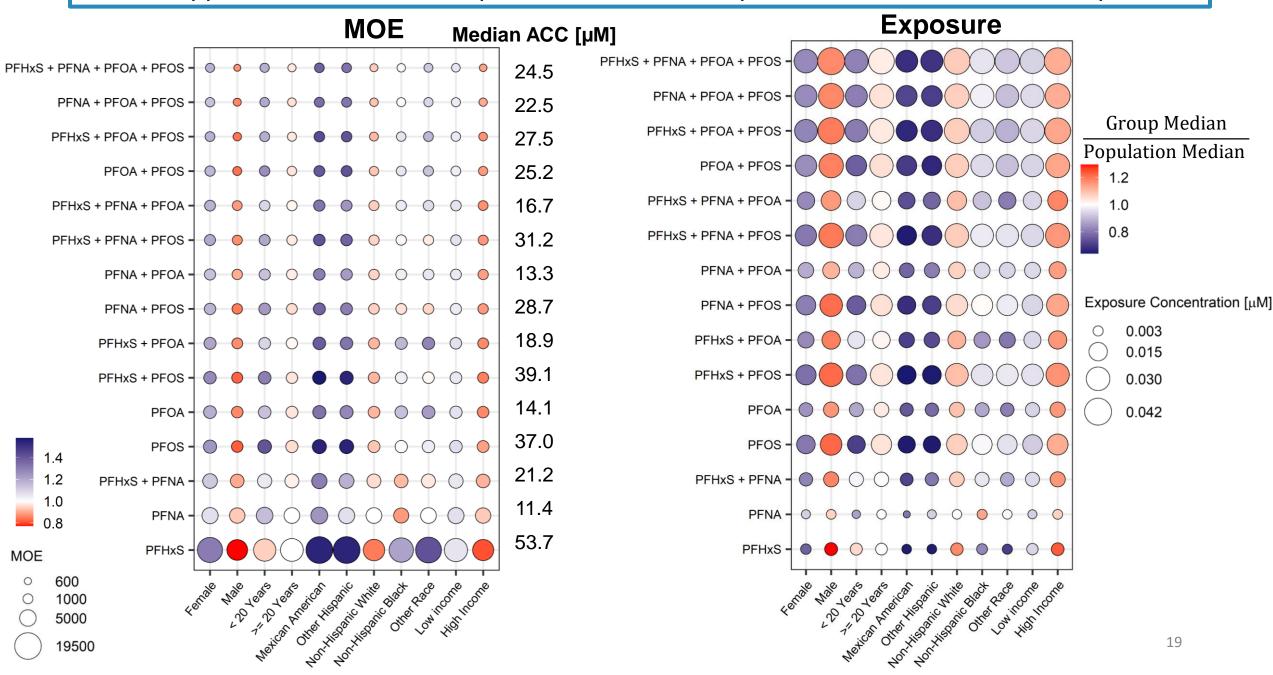
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Mixtures with both PFOA and PFOS have lowest MOE values, with the combination of all 4 chemicals being the lowest

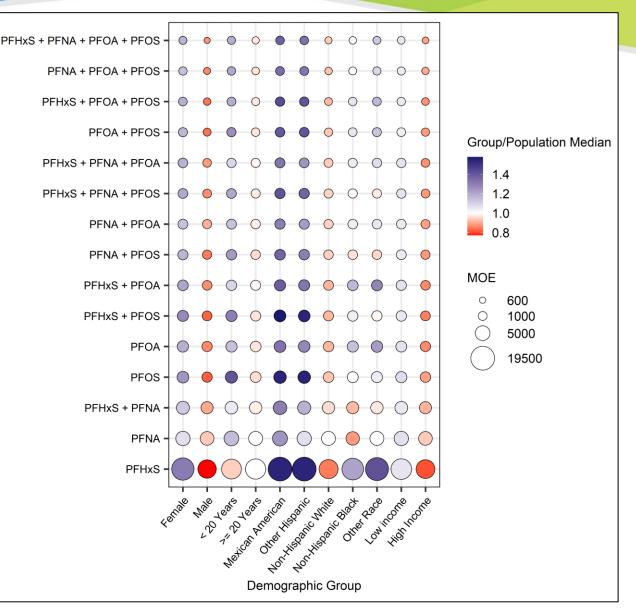


Trends appear to be related to exposure levels, not the potencies of the chemical components



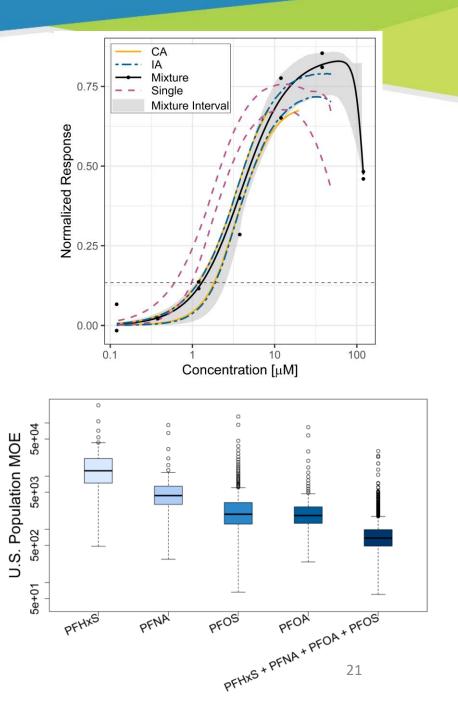
Part 2 Summary

- Examined the median margin of exposures (MOE) of 4 PFAS chemicals and mixture combinations, as detected in CDC 2009-2010 NHANES blood serum samples
 - Considered the predicted mixture's bioactivity ACC to calculate median MOE for U.S. population, comparing different demographic subsets
- Differences in MOE across demographic groups appear to be driven by differences in exposure levels
- More information needed to make conclusive observations about exposure patterns



Conclusions

- Demonstrated ability to determine mixture bioactivity from additive mathematical mixture models for an experimental set of binary chemical mixtures in ToxCast
 - Able to make conservative ACC predictions
 - Represents progress towards predictive analysis of chemical mixtures
- Similar methods can be applied to other chemical mixture combinations
 - Requires components to have been tested in ToxCast and knowledge of mixture concentration ratios
- Case study of simulated mixtures to compute margins of exposure highlights possible future applications for mixtures informed by relevant co-exposures
 - Additional investigation of exposure sources may be informative



Thank You

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- Collaborators: Madison Feshuk and Zachary Stanfield
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