

Case Study: Using spatial data to predict PFAS in fish tissue for sampling prioritization in the Columbia River Basin

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Disclaimer

The views expressed in this presentation are those of the author and do not necessarily represent the views or policies of the U.S. Environmental Protection Agency.

Core Partner Group

- Columbia River Inter-Tribal Fish Commission
- Confederated Salish and Kootenai Tribes
- Confederated Tribes of the Grand Ronde
- Confederated Tribes of the Umatilla Indian Reservation
- Yakama Nation
- Confederated Tribes of Warm Springs
- Kalispel Tribe of Indians
- Nez Perce Tribe
- Columbia River Basin Restoration Working Group
- Oregon Health Authority

EPA Office of Research & Development

- Elaine Cohen Hubal
- Ashley Mullikin
- Ana Rappold

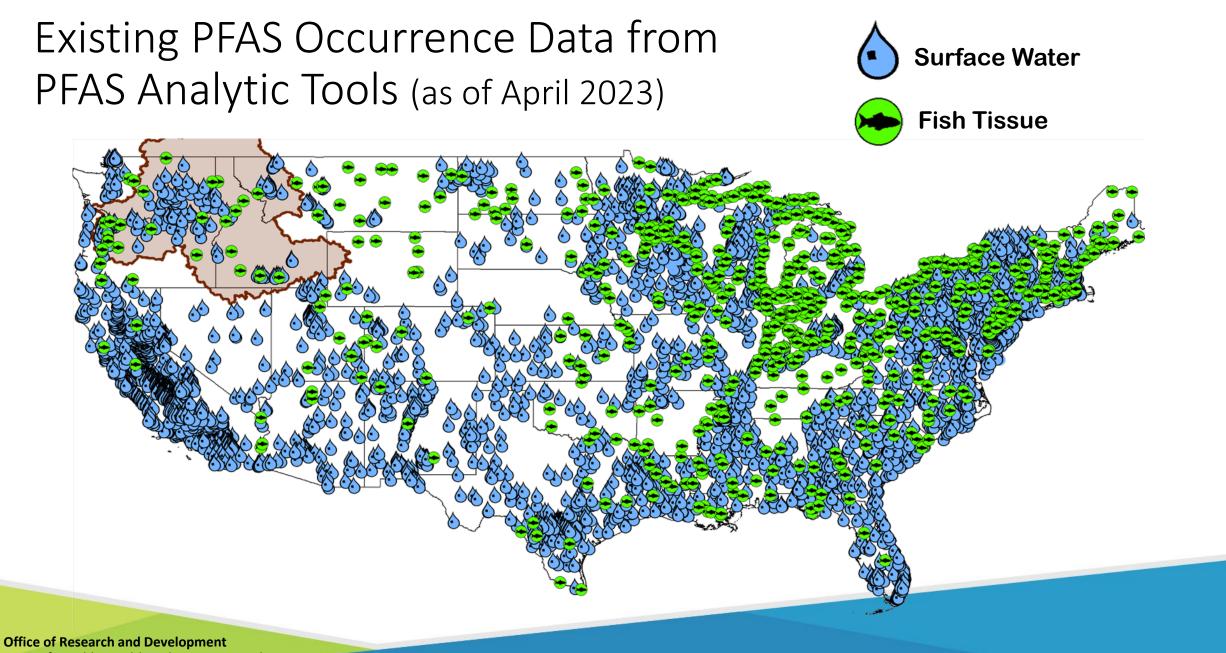
- Oregon Department of Environmental Quality
- Washington Department of Ecology
- Washington Department of Health
- Montana Department of Environmental Quality
- Missoula Valley Water Quality District
- Idaho Department of Environmental Quality
- US Geological Survey
- USEPA Regions 8, 9, 10
- USEPA Office of Research and Developement
- USEPA Office of Enforcement and Compliance Assurance

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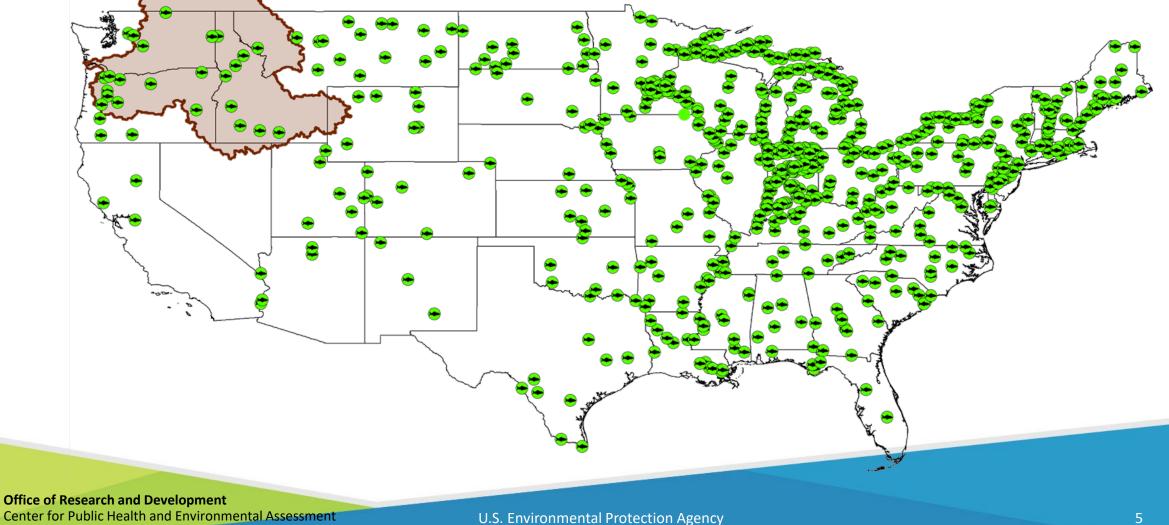
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Existing PFAS Occurrence Data from PFAS Analytic Tools (as of April 2023)





Columbia River Basin

One of the largest watersheds in the U.S. (~666,700 sq. km)

- 16 federally recognized tribal reservations
- Washington, Oregon, Idaho, Montana, British Columbia
 - Wyoming, Nevada, and Utah

Environmental Justice

- Contaminated fish pose a disproportionately high exposure risk for high fish-consuming populations
- Tribal people consume **9 to 12 times more fish** than U.S. general population (Columbia River Inter-Tribal Fish Commission and USEPA, 1994)

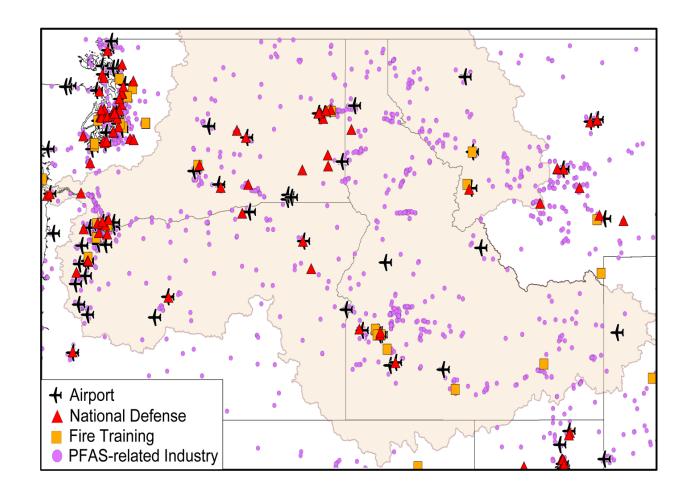




Northwest Indian Fisheries Commission

Motivation

- Jurisdictions around the world are working to efficiently identify and characterize the extent of PFAS contamination and human exposure.
- Tribes and States in EPA Regions 8 and 10 sought an efficient and cost-efficient way to prioritize site investigation and screen for PFAS contamination.
- Challenges include:
 - Many potential sources
 - Unknown facilities' PFAS use
 - Uncertain fate and transport
 - Limited PFAS measurements

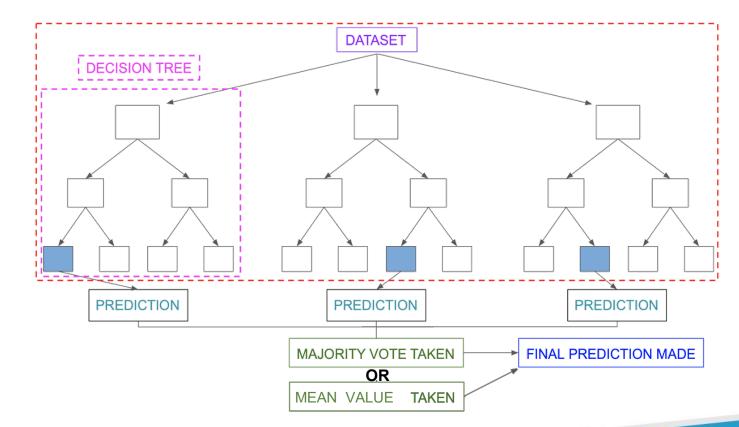


Methodology

Random Forest

Source: section.io/engineering-education/introductionto-random-forest-in-machine-learning/

- Use widely available geospatial data and limited existing PFAS measurements in fish tissue to develop a model that can predict PFAS concentrations and identify potential hotspots in Washington and Oregon states
- Random forest modeling
 - Regression predict concentration values
 - Classification predict above/below a threshold concentration



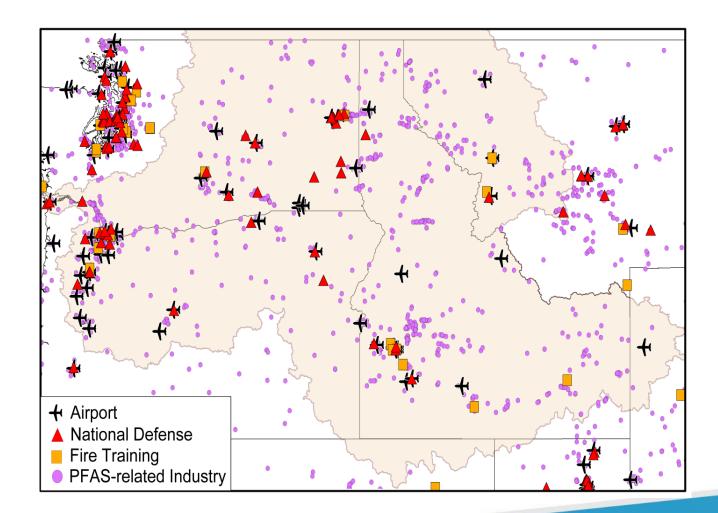
Potential Sources from PFAS Analytic Tools (as of April 2023)

PFAS-related Industries (EPA ECHO)

- Fire Training
- Airports
- National Defense
- Mining and Refining
- Landfills
- Metal Coating
- Metal Machinery Manufacturing
- Industrial Gas
- Glass Products
- Furniture and Carpeting
- Electronics
- Consumer Products
- Cleaning Product Manufacturing
- Chemical Manufacturing
- Cement Manufacturing
- Petroleum
- Industrial Gas
- Paints and Coatings
- Oil and Gas
- Plastics and Resins
- Printing
- Paper Mills
- Textiles

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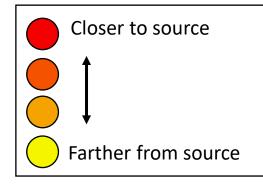
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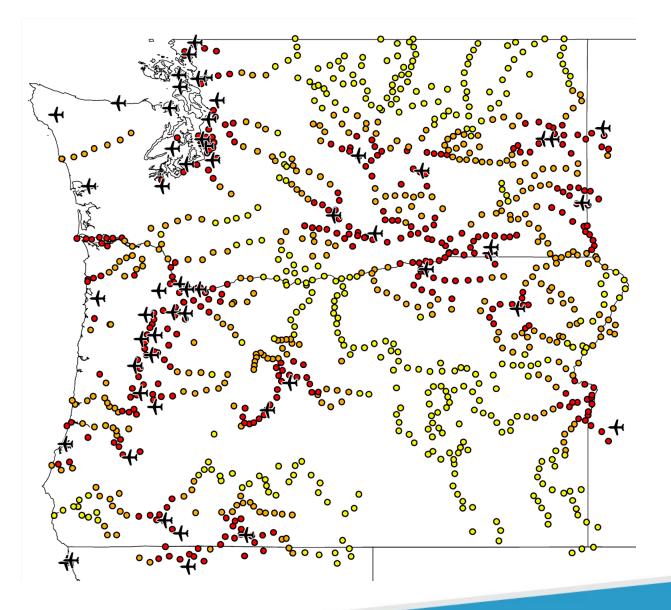


DeLuca et al., 2023

Quantifying Spatial Data

- For each industry, calculated distances from points along waterbodies to nearest potential source
- Example Distance to nearest airport

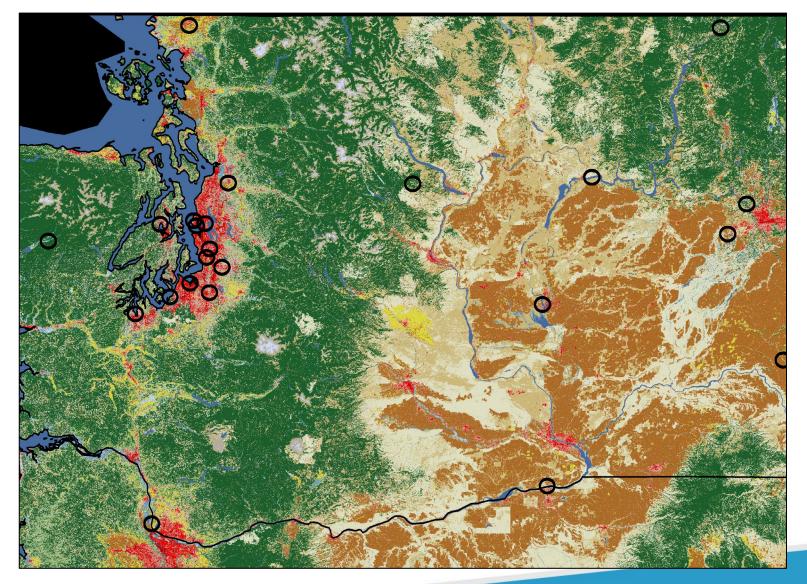




Land Cover

- USGS National Land Cover Database (NLCD)
- Calculate within 5 km buffer
 - % Developed Land
 - % Agricultural Land
 - % Natural Land

O 5 km buffer around waterbody points



Existing Fish Tissue PFAS Measurements

Data Sources

- EPA's PFAS Analytic Tools
 - Water Quality Portal 0
 - **EPA National Rivers and** 0 Streams Assessment
- Washington Department of Ecology

Data Summary

- Fillet samples (n=45)
- 2008 2019 •
- Fish Species:
 - Brook trout, sea trout
 - Brown bullhead
 - Channel catfish
 - Common carp
 - Cuthroat trout
 - Largemouth bass
 - Largescale sucker
 - Mountain whitefish
 - Northern pikeminnow
 - Peamouth
 - Pumpkinseed
 - Rainbow trout, redband trout, steelhead
 - Smallmouth bass
 - Tench
 - Tyee sucker
 - Walleye ٠
 - Yellow perch

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PFOS

PFOA

PFNA

• PFDA

• PFHxS

PFDoA

PFHxA

PFHpA

• PFUnA

PFBS

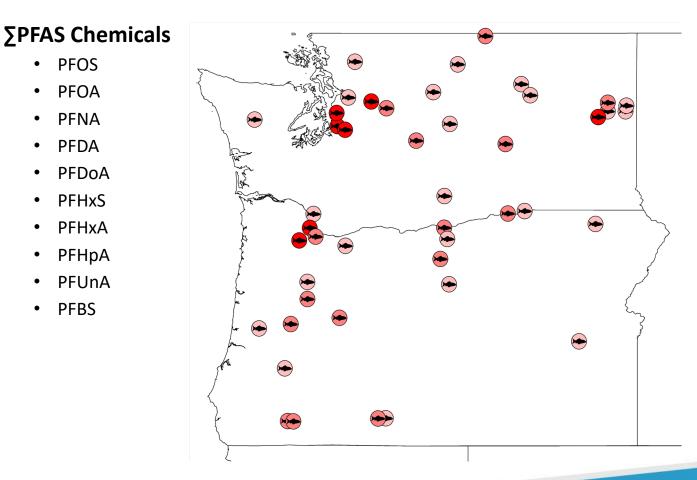
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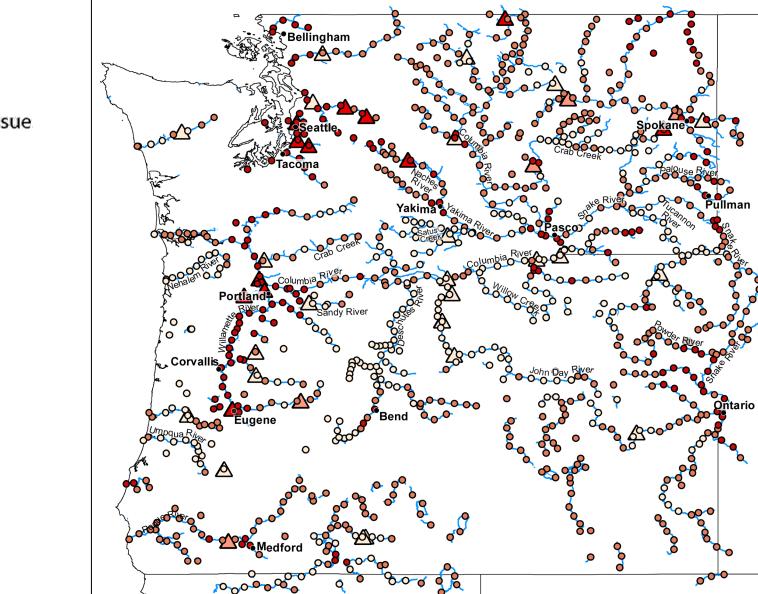
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Results - Regression

Existing Fish Tissue
ΣPFAS (ng/g)Predicted Fish Tissue
ΣPFAS (ng/g)▲ 0-2Ο 0-2

0 2-5

5+

Monte Carlo model evaluation

2 - 5

5+

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 100 random splits of dataset into training (80%) and test (20%) data

> Mean MAE: 7.26 ng/g Mean RMSE: 164.65 ng/g Mean Bias: 1.14 ng/g

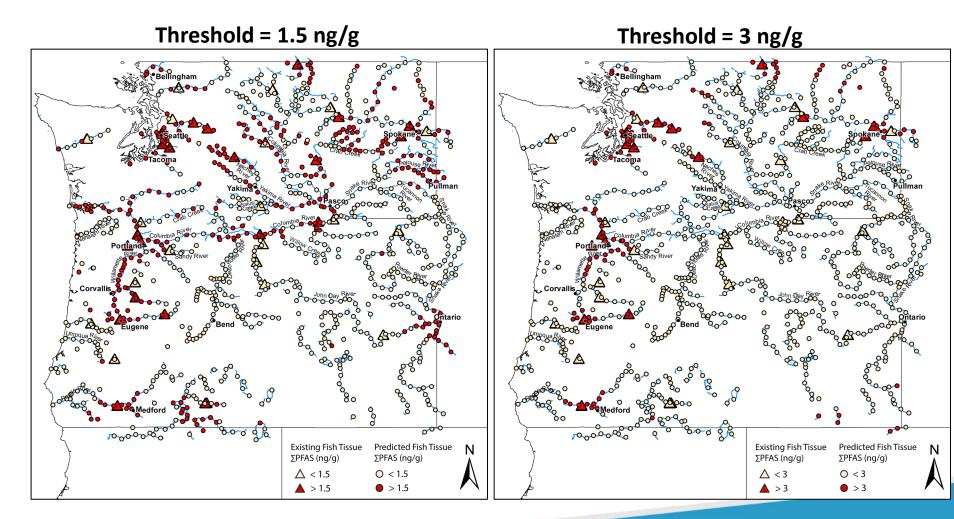
Results - Classification

- Monte Carlo model evaluation
 - 3 ng/g threshold

Mean AUC: 0.72 Mean Accuracy: 80.2% Mean Sensitivity: 74.6% Mean Specificity: 84.4%

• 1.5 ng/g threshold

Mean AUC: 0.63 Mean Accuracy: 71.0% Mean Sensitivity: 65.1% Mean Specificity: 79.4%



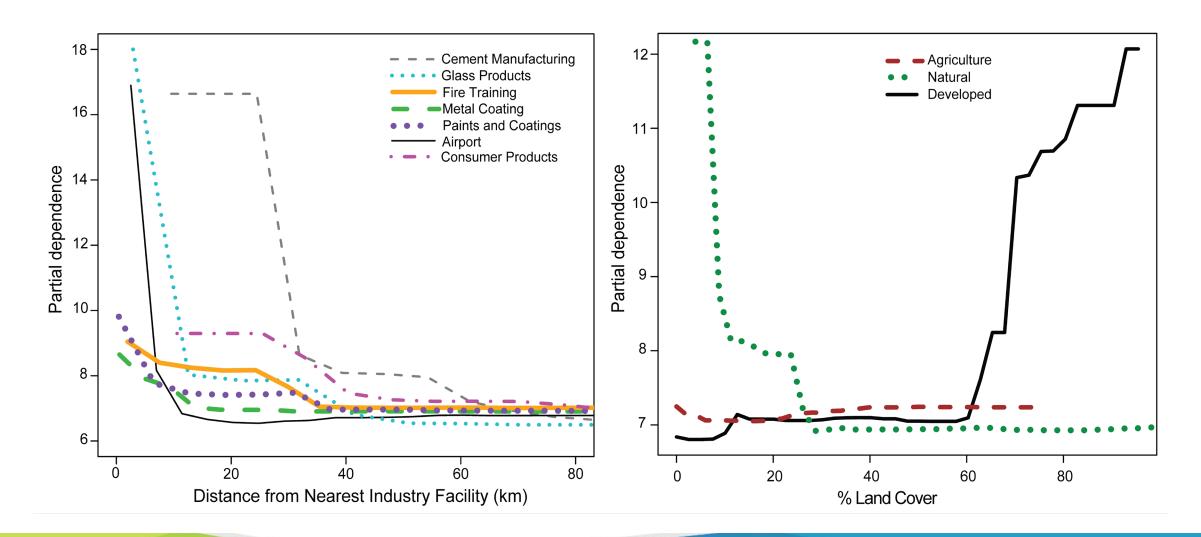
Results – Variable Importance

	g			Oldsonication (ollig/g)
Cement Manufacturing	• • • •	More	Paints and Coatings	••••
Glass Products	• • • • • • • • • • • • • • • • • • • •	Important	Metal Machinery Manufacturing	• • • • • • • • • • • • • • • • • • • •
% Developed Land	• • • • • • • • • • • • • • • • • • • •		Landfill	
Fire Training		Variables	Metal Coating	
Metal Coating			Mining and Refining	• • • • • • • • • • • • • • • • • • • •
Paints and Coatings	••••••		Fire Training	
Airport	••••••	Т	% Developed Land	
Consumer Products	••••••		Furniture	
% Natural Land	•••••		Petroleum	••••••
Industrial Gas	••••••		Textiles	••••••
Furniture			Wastewater Treatment	••••••
Metal Machinery Manufacturing	••••••		Industrial Gas	••••••
Textiles	••••••		Cleaning Product Manufacturing	•••••••
Mining and Refining	•••••••••••••••••••••••••••••••••••••••		Glass Products	•••••••
National Defense			Cement Manufacturing	• • • • • • • • • • • • • • • • • • • •
Cleaning Product Manufacturing	••••••		Consumer Products	••••••
Printing	••••••		Oil and Gas	•••••
Petroleum	••••••		Electronics	
Chemical Manufacturing	•••••		National Defense	• • • • • • • • • • • • • • • • • • • •
Landfill	•••••	L	Plastics and Resins	•••••
Wastewater Treatment	•••••		Airport	•••••
Electronics	••••		% Agricultural Land	•••••
% Agricultural Land	•••••	Less	% Natural Land	•••••••
Oil and Gas	••••	Important	Printing	
Paper Mill	••••		Paper Mill	-
Plastics and Resins	••••	Variables	Chemical Manufacturing	· •
	-2 0 2 4 6 8 10			-2 0 2 4 6 8 10
	% Increase MSE			Mean Decrease Accuracy

Regression

Classification (3 ng/g)

Results – Regression, Partial Dependence Plot



Summary

- Piloted a modeling approach for PFAS contamination in fish tissue that can be used for screening and site prioritization in the Columbia River Basin
- Spatial data for potential sources (EPA's PFAS Analytic Tools) and land cover used as predictor variables in Random Forest models
- Found existing PFAS measurements in environmental media (EPA's PFAS Analytic Tools) to develop and evaluate models
- Regression models could be used to identify hotspots of PFAS in fish tissue
- Regulators could utilize classification models by choosing a threshold concentration value meaningful to their jurisdiction's health advisory, chemical, or vulnerable populations
- More details can be found in publication:

DeLuca, N. M., Mullikin, A., Brumm, P., Rappold, A. G., & Cohen Hubal, E. (2023). Using Geospatial Data and Random Forest To Predict PFAS Contamination in Fish Tissue in the Columbia River Basin, United States. *Environmental Science & Technology*, *57*(37), 14024-14035.

https://pubs.acs.org/doi/full/10.1021/acs.est.3c03670