Applications of Machine Learning Approaches to Predict PFAS Profiles and Fate in Wastewater

Christopher I. Olivares

chris.olivares@uci.edu

Department of Civil and Environmental Engineering University of California, Irvine



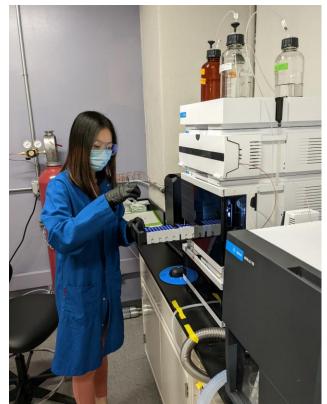
PFAS concerns

- Sampling and measurement of PFAS in WWTP
- Observed increase in quantifiable PFAS in effluents
- Biosolids heterogenous matrix

Field Sampling







May Saddle May Section May Se

Data Availability

Skilled data interpretation

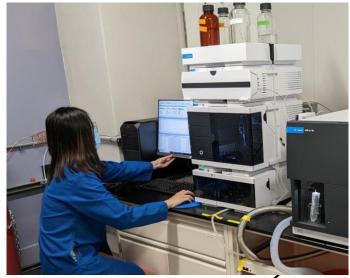
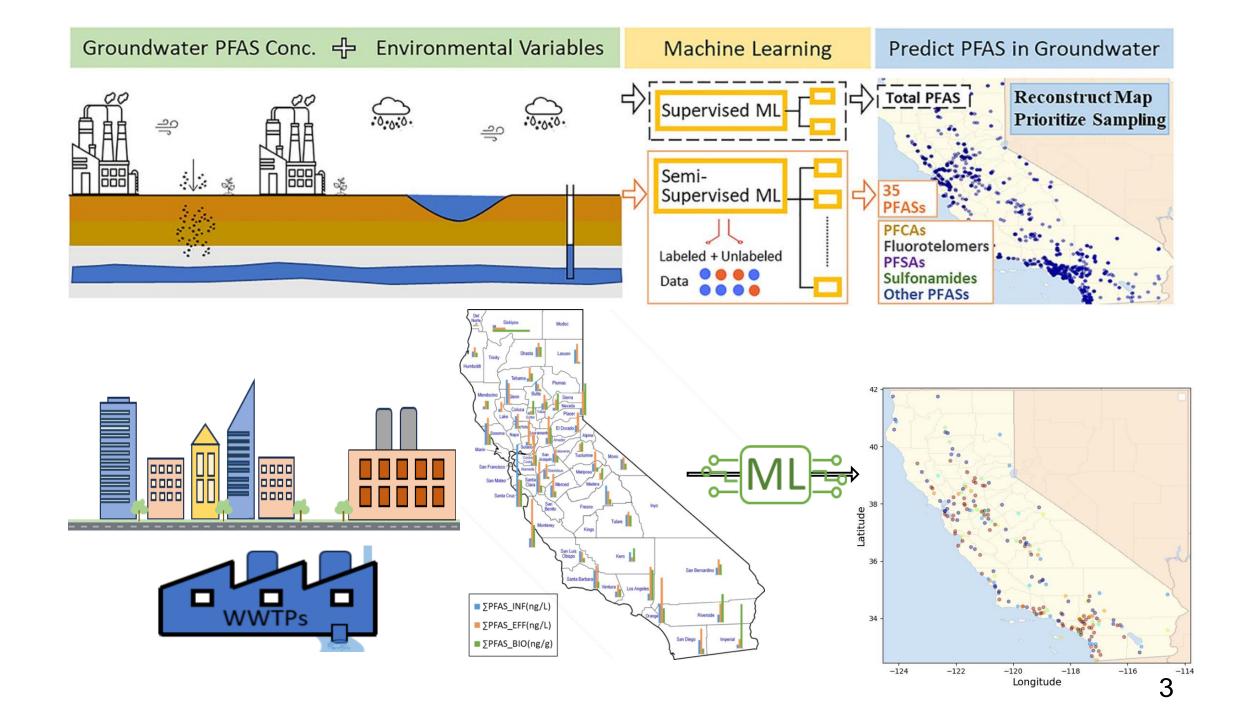
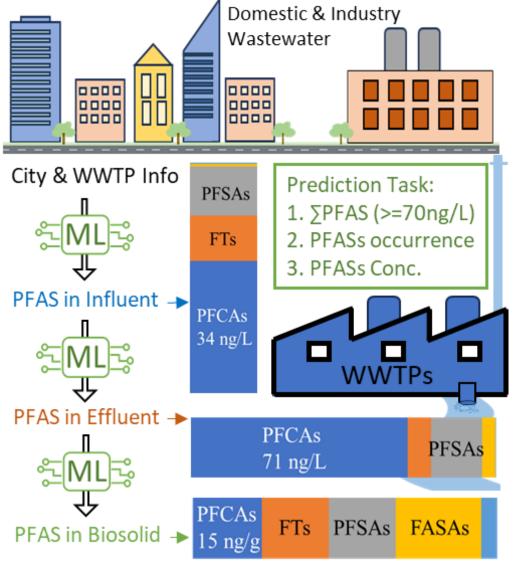


Figure adapted from AgriLife Today (TAMU)



Machine Learning Approaches to predict PFAS profiles in WWTPs

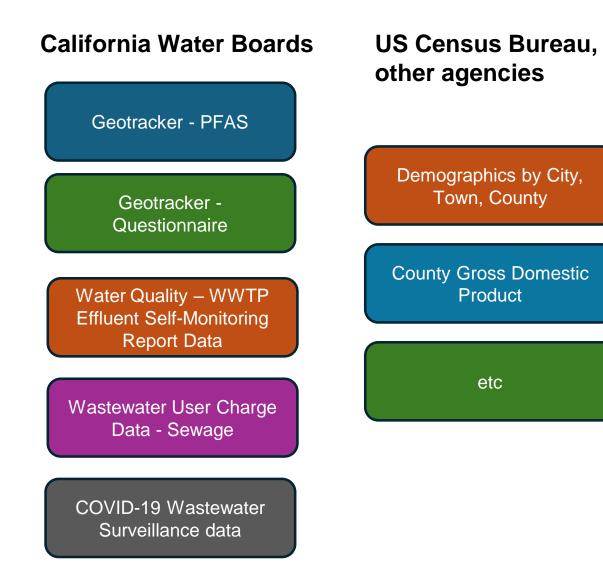
Objective



Statewide WWTP database (PFAS, wastewater characteristics, treatment)

- Machine learning models to predict PFAS in effluent and biosolids
- Identify features that influence PFAS composition in effluents, biosolids

Data sources



PFAS data validation

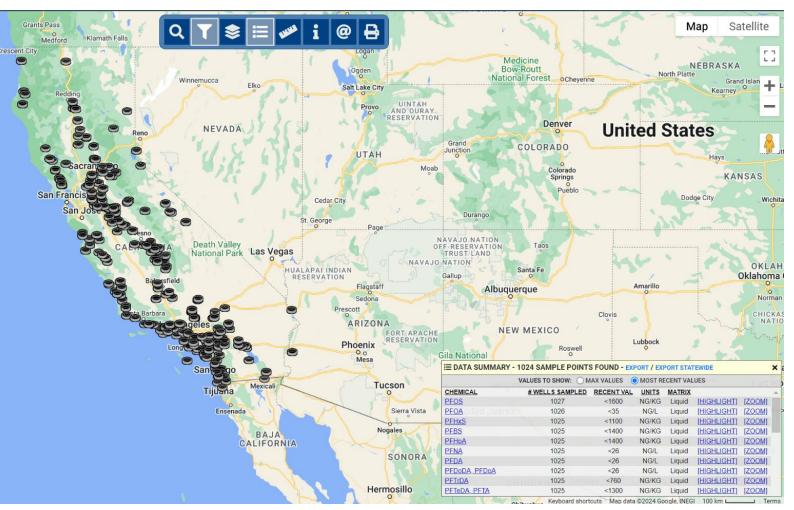
We scrutinized data in scenarios:

- PFASs were detected in both influent and effluent but not in biosolids
- No PFASs were detected in any of the liquid, solid matrices
- PFASs were detected in biosolids but not in influent or effluent
- Outliers in PFASs concentration in any of these phases.

PFAS Sampling locations & WWTP locations

The compiled dataset includes:

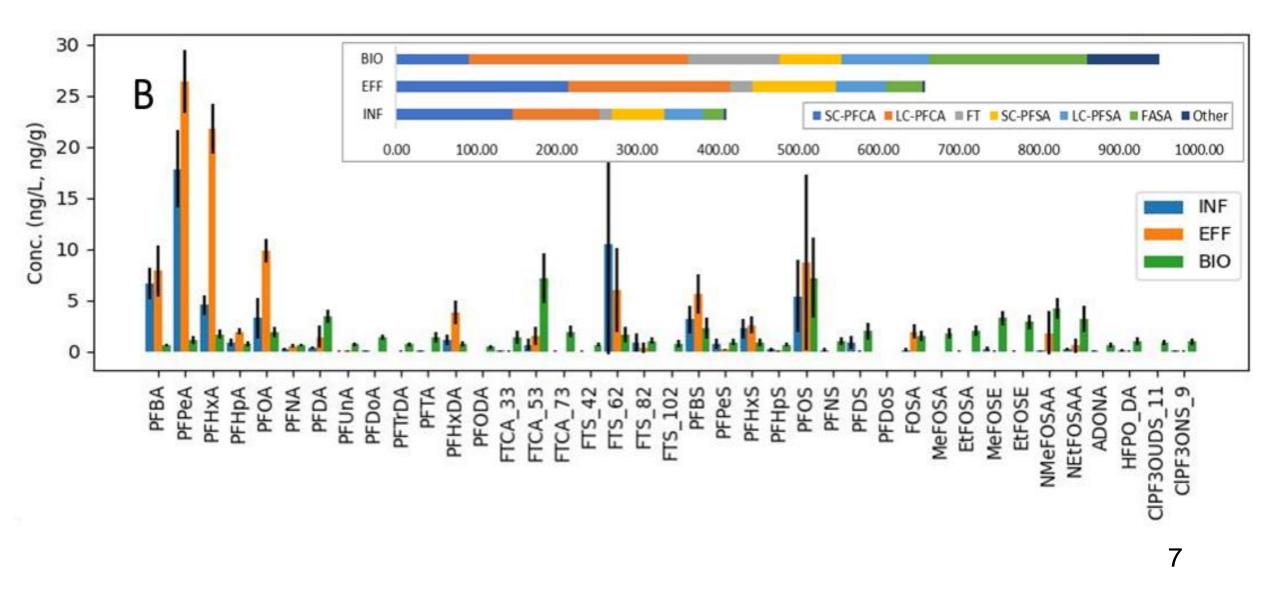
- > 213 WWTPs from California
- ≻ 2020 2023
- 39 PFASs concentrations in influent, effluent, and biosolids
- ➤ Features:
 - Wastewater source
 - WWTP information (e.g., location, sample date, WWTP size, treatment process, water quality indicators)
 - Socioeconomic factors
- ➤ 380 columns, 931 rows



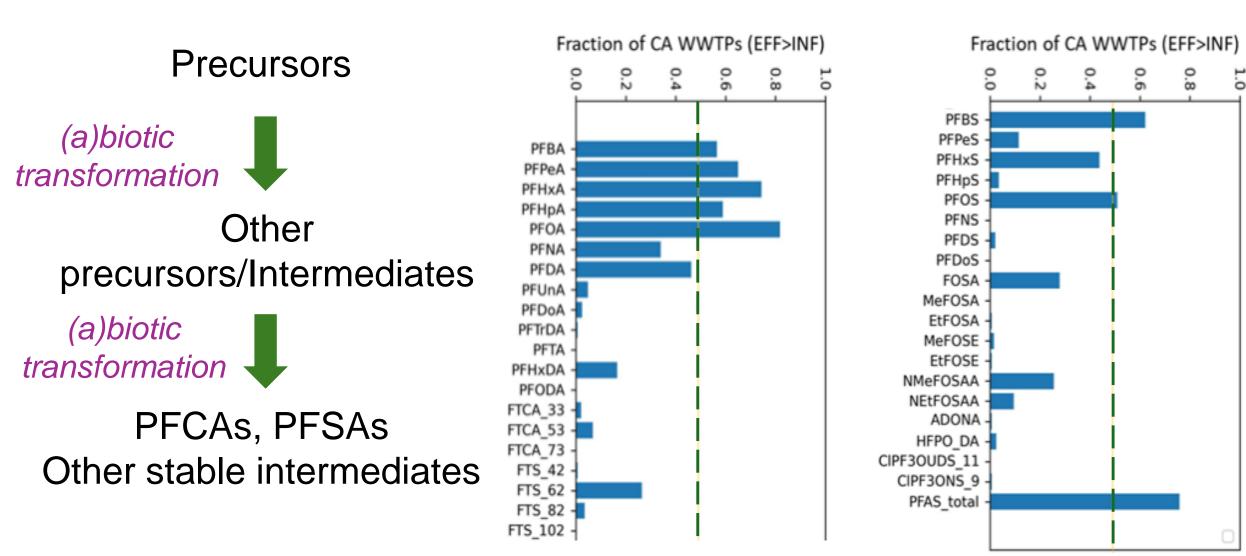
Geotracker map showing sampled WWTPs

Results – Statistical Summary

PFAS occurrence and profile in California WWTPs.



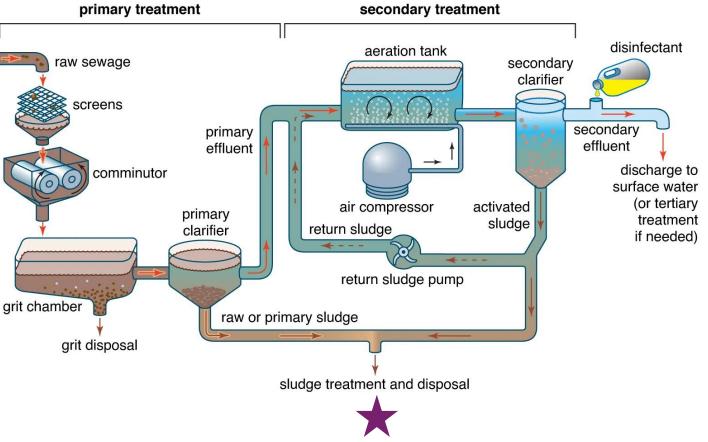
Effluent > influent concentrations for perfluorinated alkyl acids, some fluorotelomers and sulfonamides



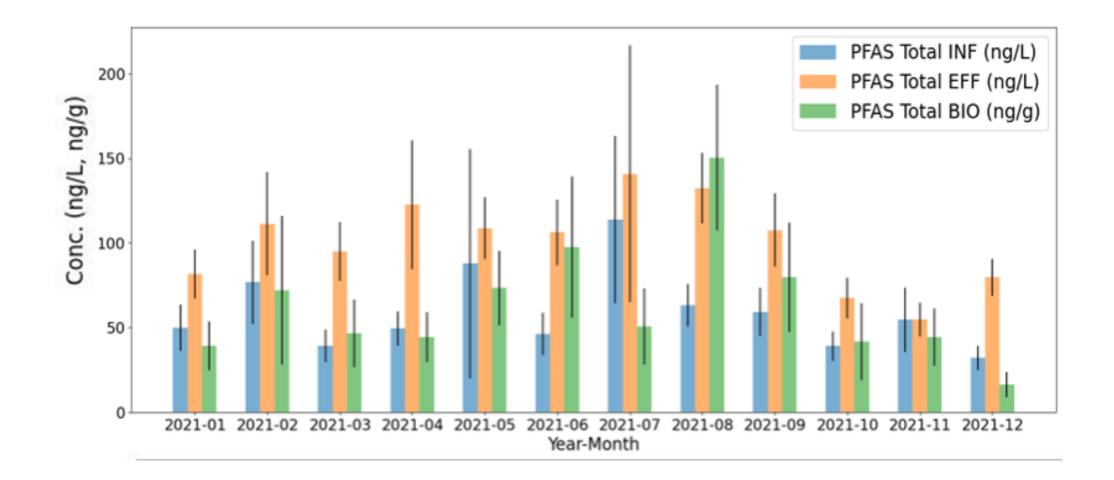
Effluent > influent concentrations for perfluorinated alkyl acids, some fluorotelomers and sulfonamides

★(a)biotic transformation Other precursors/Intermediates ★ (a)biotic transformation PFCAs, PFSAs Other stable intermediates

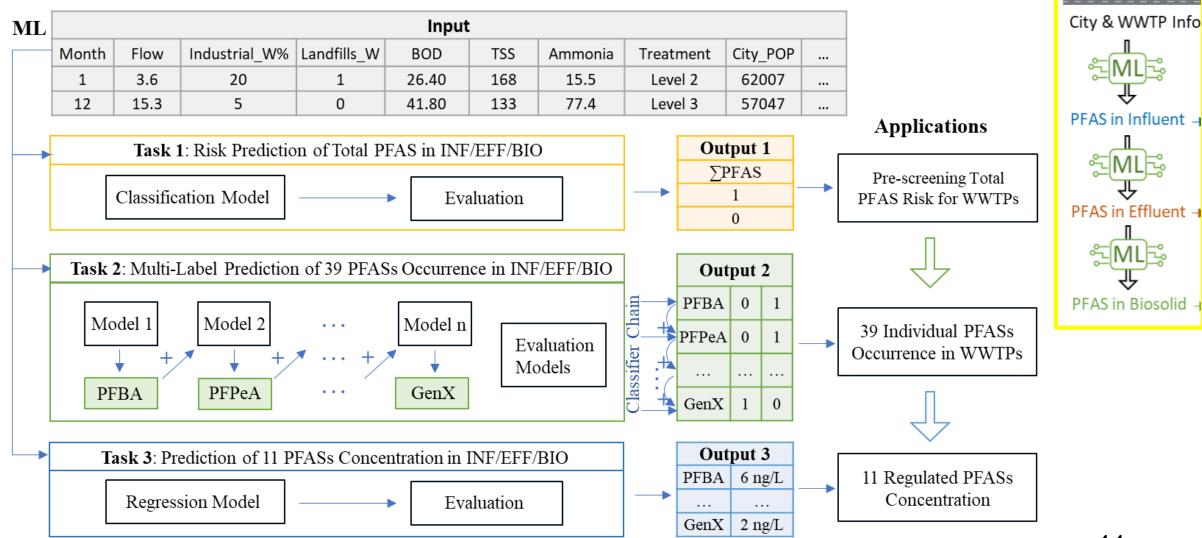
Precursors



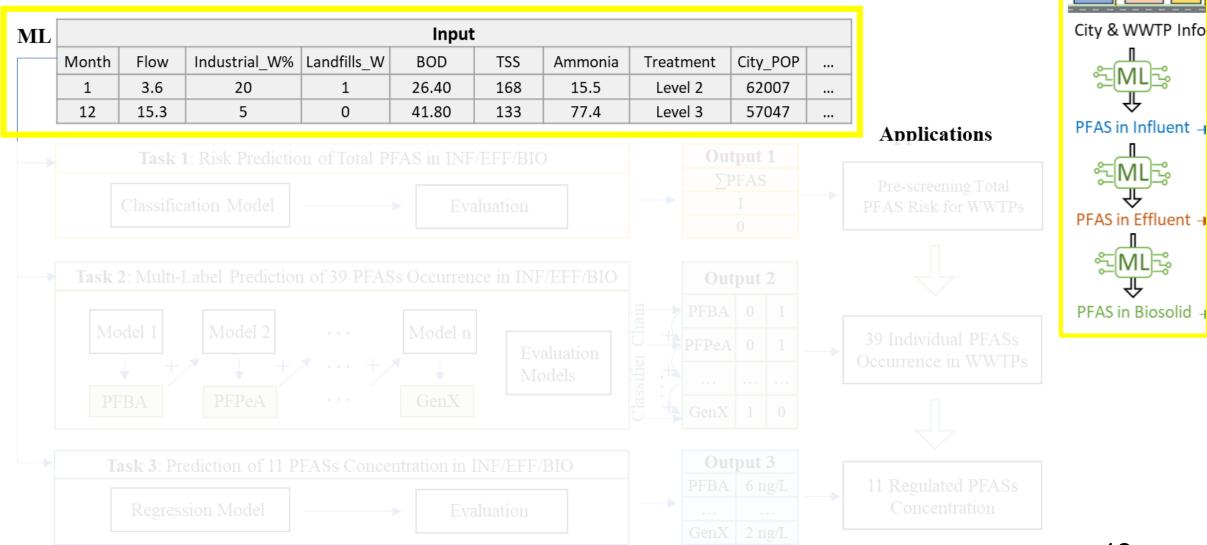
Slight seasonal pattern (year 2021)

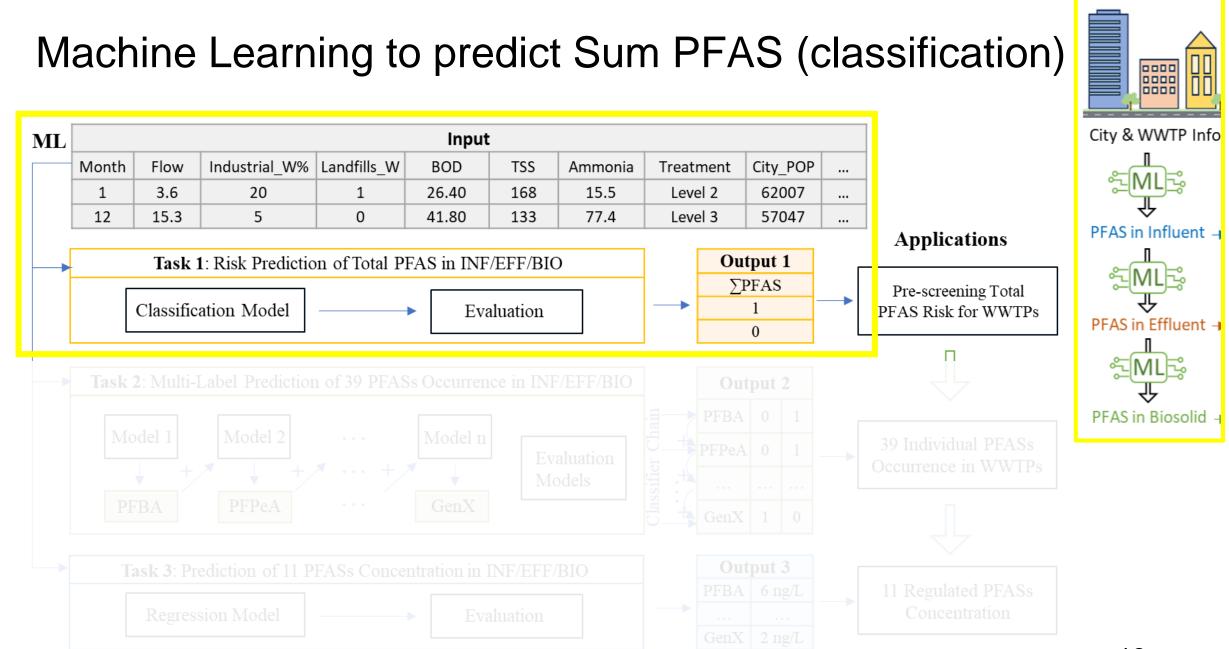


Machine Learning Approaches



Machine Learning Approaches





Result - Prediction of $\sum PFAS$: influent, effluent (70 ng/L threshold), biosolids (any detected concentration)

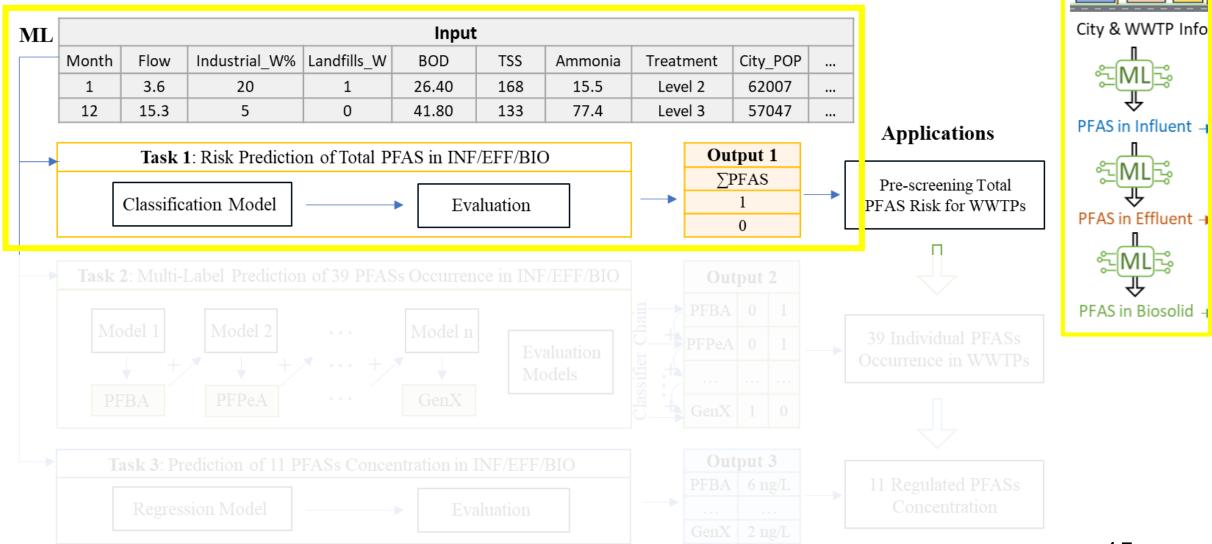
	90. AUROC	INF			hul	luul		_		Ind	Ind	Ind	I.	
y	.30	0 -												
	00. 90. ن	GaussianN	B LR	SVC	XGBoost	LGBM	CatBoost	TabNet	RF	LDA	QDA	KNN	GP	AdaBoost
	O. AUROC	0	1											
	.00	GaussianN	B LR	SVC	XGBoost	LGBM	CatBoost	TabNet	RF	LDA uracy R	QDA ecall ■Pre	KNN ecision = F	GP 1 score	AdaBoost
	AUROC 0.00	0	B LR	SVC	XGBoost	LGBM	CatBoost	TabNet	RF	LDA	QDA	KNN	GP	A daB oost

Performance evaluation of ∑PFAS in INF, EFF, BIO

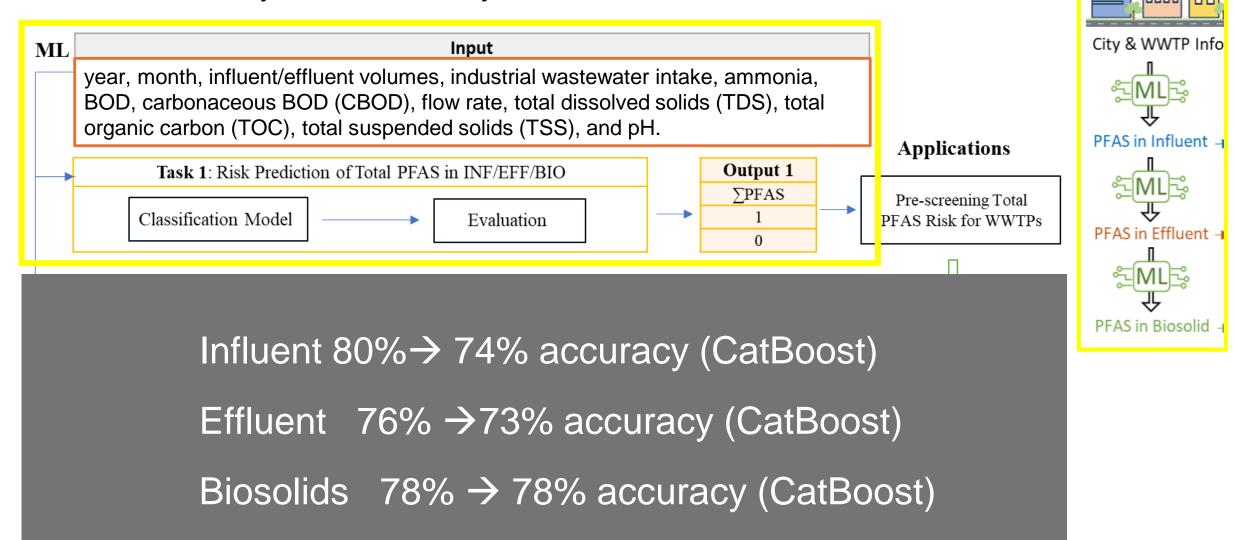
GaussianNB: Gaussian Naive Bayes; LR: Logistic Regression; SVC: Support Vector Classifier; XGBoost: Extreme Gradient Boosting; LGBM: Light Gradient Boosting Machine; CatBoost: Categorical Boosting; TabNet: Tabular Network; RF: Random Forest; LDA: Linear Discriminant Analysis; QDA: Quadratic Discriminant Analysis; KNN: K-Nearest Neighbors; GP: Gaussian Process; AdaBoost: Adaptive Boosting.

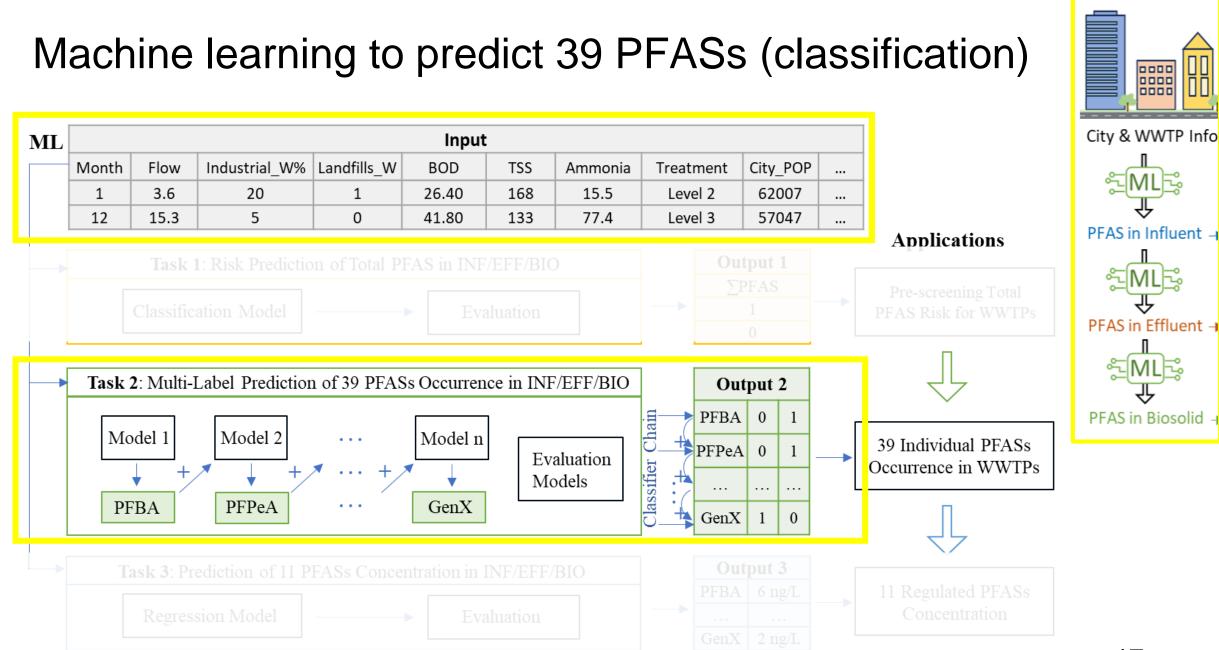
Matrix	Algorithm	Accuracy
Influent	Random Forest	80.1%
Effluent	Random Forest	76.4%
Biosolids	CatBoost	77.9%

Machine Learning to predict Sum PFAS

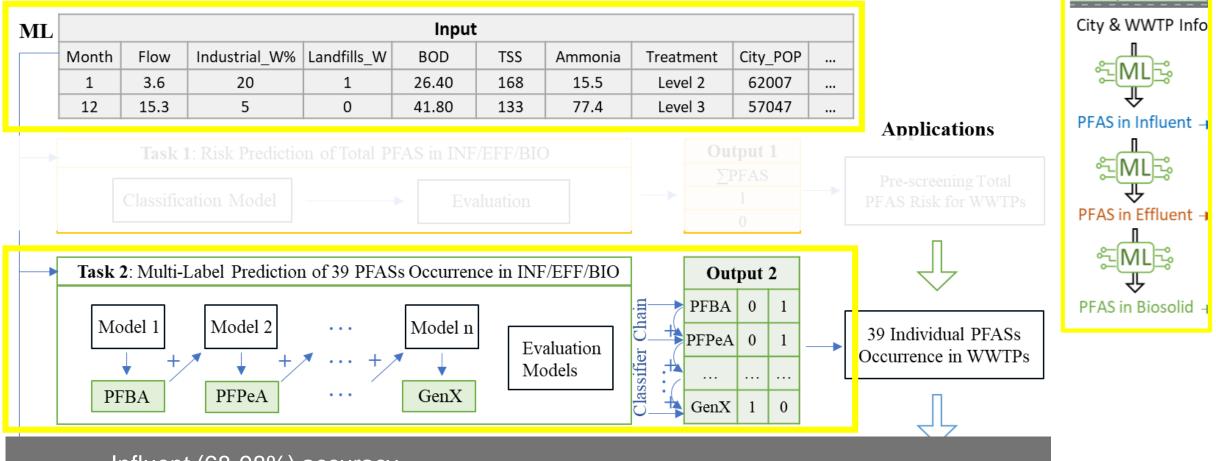


Machine Learning to predict sum PFAS (70 ng/L) But this time only with commonly measured wastewater features





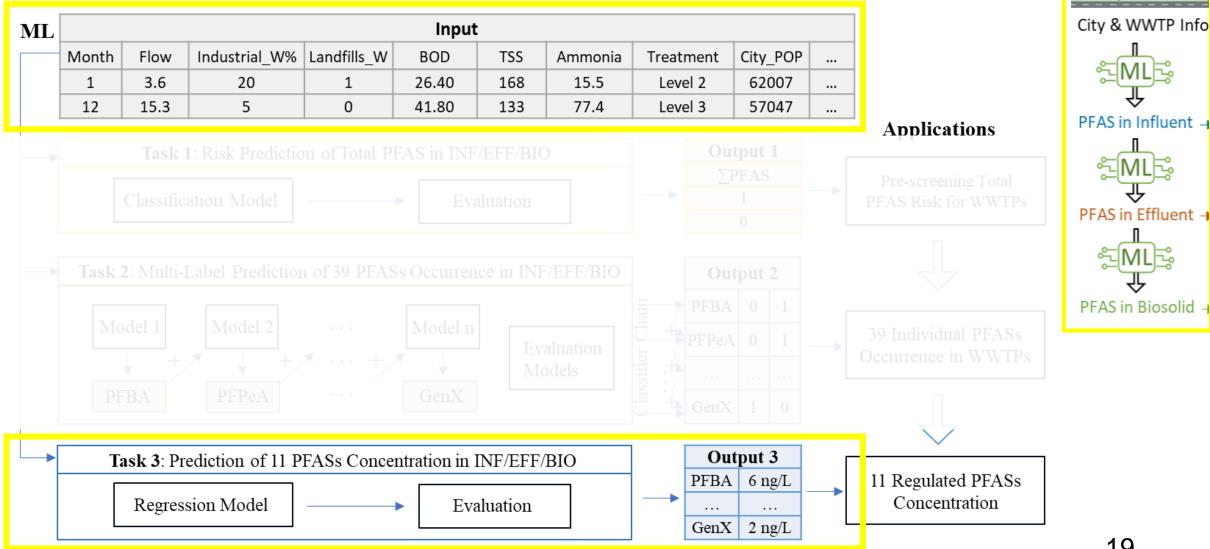
Machine learning to predict 39 PFASs (classification, 2 ng/L as threshold)



Influent (68-98%) accuracy Effluent (80-98%) accuracy Biosolids (80-98%) accuracy

Dependent on PFAS detection frequency

Machine learning to predict concentrations of 11 PFASs (regression)

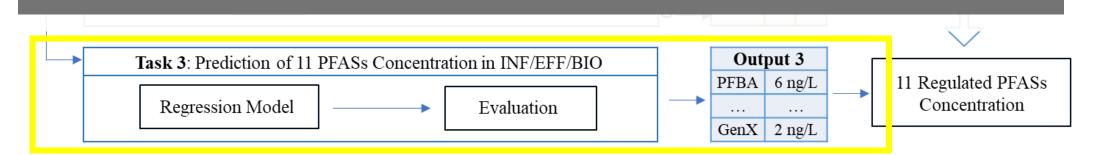


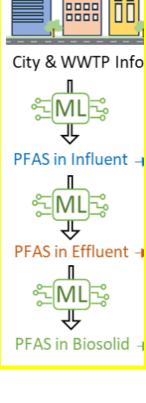
Machine learning to predict concentrations of 11 PFASs (regression)

L											
	Month	Flow	Industrial_W%	Landfills_W	BOD	TSS	Ammonia	Treatment	City_POP		
	1	3.6	20	1	26.40	168	15.5	Level 2	62007		
	12	15.3	5	0	41.80	133	77.4	Level 3	57047		
											Applications
		lask							itput 1		
		lask									

XGBoost

Influent: PFHpA, PFBA performed best 0.97 R² Effluent: PFHpA, PFBA, PFOS performed best 0.99 R² Biosolids: PFHpA, PFOA, PFNA, GenX performed best 0.98 R²

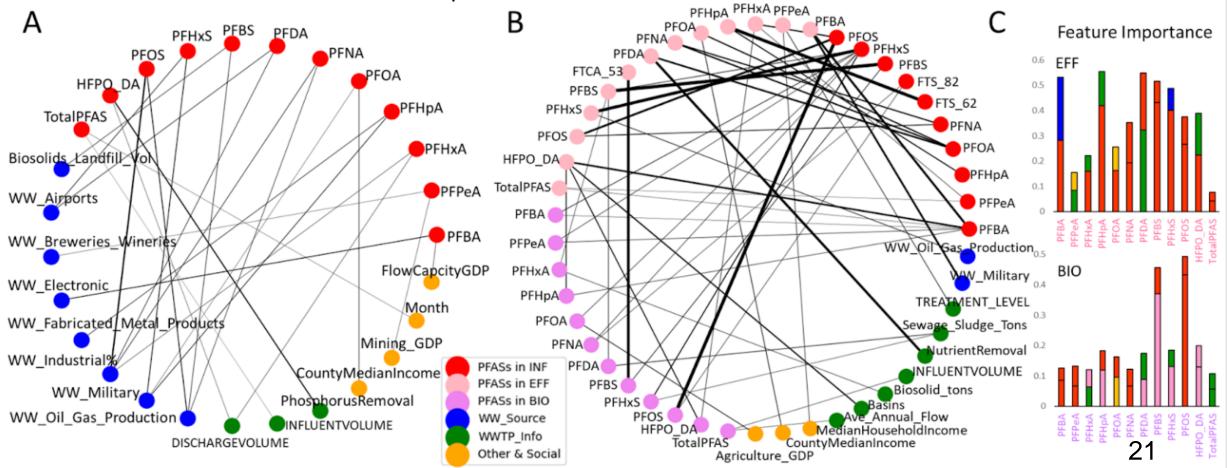




Environmental Variables Influencing PFAS Prediction

- All matrices, **size** and **operational capacity** of WWTPs significant predictors.
- Wastewater source impact on the PFAS profile in the influent, as well as the prediction. (e.g., PFBA, electronic waste)
- PFASs in influent → impact effluent PFAS
- PFASs in influent & effluent → impact biosolid PFAS

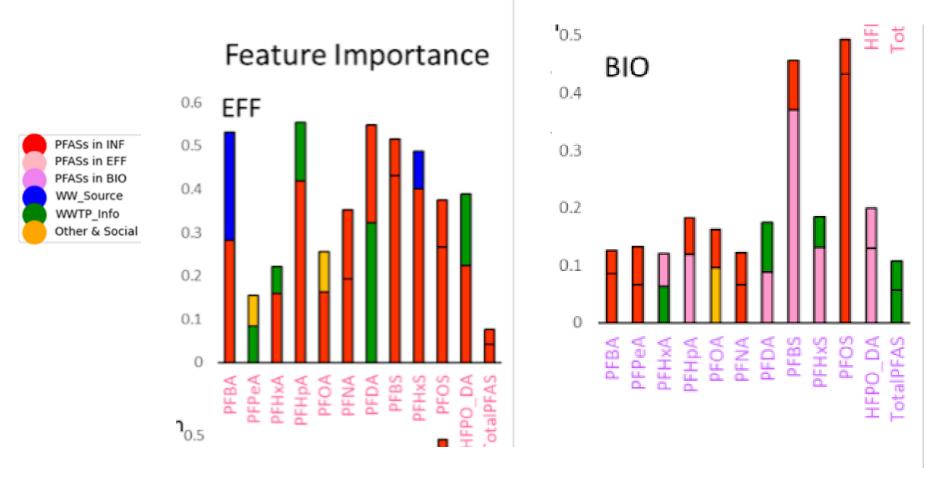
Relative contributions of the top 2 variables for the total and individual PFASs in INF, EFF, and BIO.



Result - Environmental Variables Influencing PFAS Prediction

- Socioeconomic factors (county population and GDP)
- Nutrient removal \rightarrow PFDA in the effluent.
- Influent PFBA levels \rightarrow effluent GenX concentrations and GenX in biosolids

Relative contributions of the top 2 variables for the total and individual PFASs in INF, EFF, and BIO.



We're developing a public website tool

Prediction Tool for PFAS in California WWTPs

Machine Learning for Monitoring Per- and Polyfluoroalkyl Substance (PFAS) in California's Wastewater Treatment Plants: An Assessment of Occurrence and Fate

Jialin Dong¹, Seungjun Kim², Sean D. Young^{2,3}, Chengxi Li¹, Zhichao Jin⁴, Dylan Lee ⁵, and Christopher I. Olivares¹

¹ Department of Civil and Environmental Engineering, University of California, Irvine, CA
 ² Department of Informatics, University of California, Irvine, CA
 ³ Department of Emergency Medicine, University of California, Irvine, CA
 ⁴ Department of Physics and Astronomy, University of California, Irvine, CA
 ⁵ Department of Computer Science, University of California, Irvine, CA

Correspondance: chris.olivares@uci.edu



GitHub Link

Lab Website Link

Data Link

Home Prediction PFAS for WWTP Influent Prediction PFAS for WWTP Effluent Prediction PFAS for WWTP Biosolid

WWTP_PFAS_prediction_website

Risk Prediction of Total PFAS in Influent

This ML model classifies PFAS levels in WWTP influent as high risk (1) if they exceed 70 nanograms per liter (ng/L), and low risk (0) if they fall below this threshold. The model utilizes commonly monitored standard operational parameters of WWTPs as inputs, including: year, month, influent/effluent volumes, industrial wastewater intake, total organic carbon (TOC), ammonia, biochemical oxygen demand (BOD), carbonaceous biochemical oxygen demand (CBOD), flow rate, pH, total dissolved solids (TDS), and total suspended solids (TSS).

When using the model, please ensure that all inputs are in the correct format. Input values should strictly be numerical floats; avoid using letters or non-numeric characters. The default values visible upon loading the website are set to median values derived from the dataset used during model training. These defaults serve as starting points for predictions and can be adjusted based on your specific input data.

Select a Year

169.25

2024	-	+
Select a Month		
January		~
Flow in Influent (The flow rate of the influent to the facility (MGD))		
3.1334		
Influent Volume (acre-feet/month)		
387		

Biochemical Oxygen Demand in Influent (BOD was measured in 5 days at 20 deg. C (ng/L))

255668102.2

Carbonaceous Biochemical Oxygen Demand in Influent (CBOD was measured in 5 days at 20 deg. C (ng/L))

645000000

Total Dissolved Solids in Influent (TDS (ng/L))

507170067

Total Organic Carbon in Influent (TOC (ng/L))

16043614

Total Suspended Solids in Influent (TSS (ng/L))

240900372.8

pH of Influent

7.0

Make Prediction

The PFAS risk is greater than 70 nanograms per liter (70 ng/L).

WWTP_PFAS_prediction_website 24

Discharge Volume in Influent (acre-feet/month)

Summary and Key Points

- Database of PFAS WWTP concentrations, other effluent quality parameters, socioeconomic factors
- PFAS profiles vary depending on matrix (transformation, sorption)
- Classification models (threshold) had highest accuracy for sum of PFAS predictions
- Concentration prediction worked for frequently detected PFAS
- Key predictive factors: WWTP size, wastewater source, county population, and GDP.
- Data and (data validation) is key

Acknowledgements

Olivares Lab

<u>Graduate Students</u> Jialin Dong Zixin Hu Meng-Chia Li Theodore Jagodits Seungjun Kim Sachin Thyaharajan

Undergraduate Students Takumi Takasugi **Steven Li** Xiaojun Tang Nicolas Villota **Zhichao Jin** Dylan Lee Manulya Gunasekera Funding

Startup funds for C. Olivares



Jialin Dong





Applications of Machine Learning Approaches to Predict PFAS Profiles and Fate in Wastewater

Christopher I. Olivares

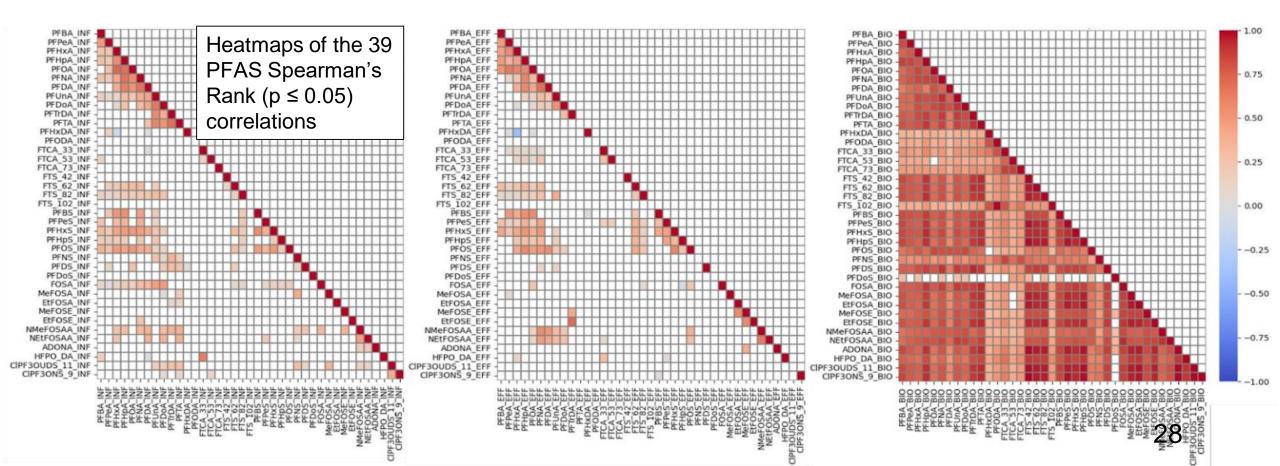
chris.olivares@uci.edu

Department of Civil and Environmental Engineering University of California, Irvine



Result - California WWTPs PFAS Analysis

- **Positive correlations** observed among various PFASs \Box sources or similar properties.
- Strong correlations (>0.6) among short-chain PFCAs, longer-chain C8–C10 PFCAs, and PFSAs with similar chain lengths;
- Effluent patterns mirror those in influent, with clearer correlations between C4–C9 PFCAs and C4–C8 PFSAs (biotransformation pathways, e.g., FTSs to PFCAs)
- Biosolids show strong correlations, C9-C13 PFCAs (correlation coefficients >0.75) (a tendency for these compounds to accumulate in the solid phase) longer chain lengths and greater affinity for solids.



Result - Prediction of total PFAS risk in INF, EFF, and BIO

- Developed a machine learning model with limited input features that consistently predicts total, with minor performance drops
- **CatBoost**, fine-tuned with SMOTE oversampling, outperformed other models in predicting PFAS concentrations
- Optimized CatBoost model to serve as a reliable tool for WWTP monitoring, even with fewer variables, suggesting broad applicability for PFAS risk prediction

		GaussianNB	LR	SVC	XGBoost	LGBM	CatBoost	TabNet	RF	LDA	QDA	KNN	GP	AdaBoost
	accuracy	0.274	0.237	0.656	0.699	0.737	0.742	0.763	0.710	0.522	0.280	0.650	0.753	0.237
Total	recall	0.493	0.500	0.555	0.591	0.631	0.627	0.500	0.606	0.475	0.520	0.583	0.681	0.500
PFAS in	precision	0.484	0.118	0.549	0.588	0.634	0.636	0.382	0.603	0.481	0.572	0.568	0.666	0.118
INF	f1 score	0.256	0.191	0.550	0.590	0.632	0.631	0.433	0.604	0.459	0.255	0.569	0.673	0.191
	AUROC	0.483	0.500	0.508	0.651	0.638	0.663	0.500	0.661	0.515	0.570	0.648	0.741	0.493
	accuracy	0.500	0.500	0.624	0.699	0.710	0.737	0.495	0.688	0.554	0.511	0.640	0.634	0.489
Total	recall	0.500	0.500	0.624	0.699	0.710	0.737	0.495	0.688	0.554	0.511	0.640	0.634	0.489
PFAS in	precision	0.500	0.240	0.630	0.701	0.716	0.745	0.249	0.694	0.445	0.515	0.643	0.641	0.373
EFF	f1 score	0.425	0.333	0.619	0.698	0.708	0.734	0.331	0.686	0.445	0.474	0.638	0.630	0.338
	AUROC	0.482	0.500	0.633	0.805	0.787	0.833	0.495	0.771	0.564	0.536	0.688	0.657	0.485
	accuracy	0.543	0.468	0.629	0.780	0.769	0.780	0.516	0.731	0.586	0.581	0.699	0.640	0.349
Total	recall	0.513	0.500	0.619	0.779	0.769	0.779	0.486	0.732	0.575	0.556	0.698	0.627	0.342
PFAS in	precision	0.603	0.234	0.632	0.779	0.768	0.779	0.362	0.731	0.585	0.639	0.698	0.653	0.334
BIO	f1 score	0.391	0.319	0.615	0.779	0.768	0.779	0.350	0.731	0.567	0.491	0.698	0.617	0.335
	AUROC	0.555	0.500	0.659	0.830	0.833	0.836	0.486	0.803	0.640	0.586	0.740	0.714	29 _{0.357}

Performance evaluation of total PFAS classification ML in INF, EFF, BIO



PFAS leachability from contaminated pavement materials

Prashant Srivastava

Australia's National Science Agency



CSIRO-Defence PFAS Research Program

Objectives

To understand:

- the presence of PFAS in relevant concrete and asphalt infrastructure and spatial and vertical distribution
- the potential water leachability of PFAS from concrete and asphalt materials

Identify and assess management strategies and/or mitigation approaches for PFAS leaching in concrete and asphalt materials for recycling, repurposing and/or reuse (on-site/off-site)



How do PFAS get into concrete and asphalt?



https://www.pdsigns.ie/contentFiles/productImages/Large/FS K2.jpg



https://content.presspage.com/uploads/1912/1920_firescienceprogrampastliveburnexercise-dmcwestcampus.jpg?10000



https://gray-wilx-

prod.cdn.arcpublishing.com/resizer/yOGIGwD3nxnkJDSMfkDzHjuXnTg=/1200x675/smart/cloud front-us-east-1.images.arcpublishing.com/gray/O5CSQV4ALZJJXPDSUCZPI4GHOE.jpg

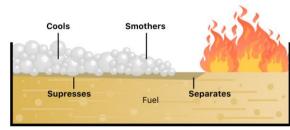


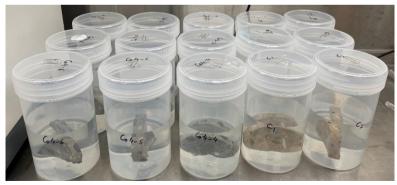
Image source: https://www.consumernotice.org/environmental/afff/

CSIRO

Leaching Approaches

- Australian Standard Leaching Protocol (ASLP)
 - Leaching period
 - Particle size
 - Temperature
- Leaching Environmental Assessment Framework (LEAF) 1313
 - pH
- Leaching Environmental Assessment Framework (LEAF) 1315
 - Intact monolith
 - Normally over 9 weeks



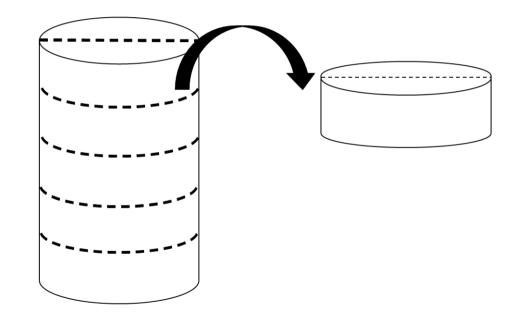


CSIRO

Sampling of PFAS-contaminated pavements









Concrete core characterisation

Core No.	Replicate	PFHxA	PFOA	PFHxS	PFOS	PFHxS + PFOS
		(µg/kg)	(µg/kg)	(µg/kg)	(µg/kg)	(µg/kg)
Core 3	R1	19	313	408	553	962
Core 3	R2	26	273	390	673	1063
Core 3	R3	39	42	155	511	666
Core 1	R1	37	28	140	458	599
Core 1	R2	39	32	155	475	629
Core 1	R3	57	412	533	824	1357
Core 4	R1	34	939	1013	1067	2081
Core 4	R2	50	834	919	1137	2056
Core 4	R3	76	309	611	778	1390
	Low Intermedia				diate	High

CSIRO

Asphalt core characterisation

Core No.	Replicate	PFHxA	PFOA	PFHxS	PFOS	PFHxS + PFOS
		(µg/kg)	(µg/kg)	(µg/kg)	(µg/kg)	(µg/kg)
Core 2	R1	13	7	41	492	533
Core 2	R2	7	6	23	219	242
Core 4	R1	12	8	43	469	512
Core 4	R2	12	9	40	471	511
Core 6	R1	5	2	6	54	60
Core 6	R2	7	3	8	58	66

Low Intermediate



High

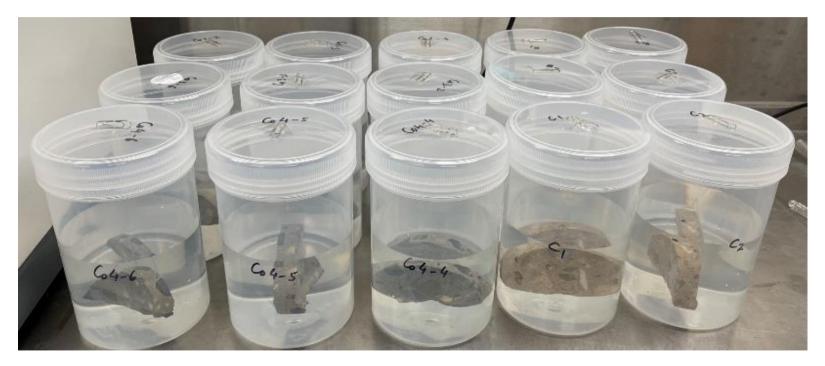


Leaching Environmental Assessment Protocol (LEAF) 1315

- Originally developed for inorganic contaminants (e.g. metals) and matrices such as soils, fly ash etc.
- Optimisation and validation required for PFAS and concrete/asphalt
- Current SERDP project (ER23-37611)
 - Texas Tech University Jenn Guelfo
 - Vanderbilt University David Kosson
 - University of Wisconsin Craig Benson
 - CSIRO Prashant Srivastava



Leaching Environmental Assessment Framework (LEAF) 1315

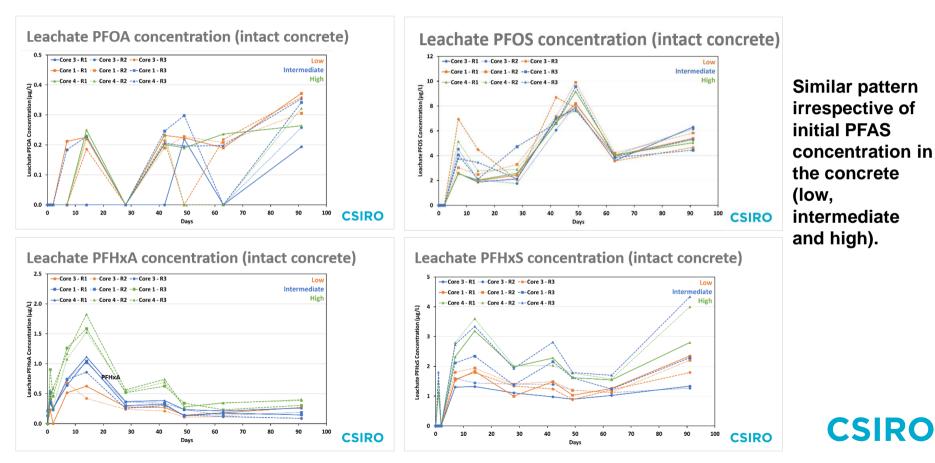




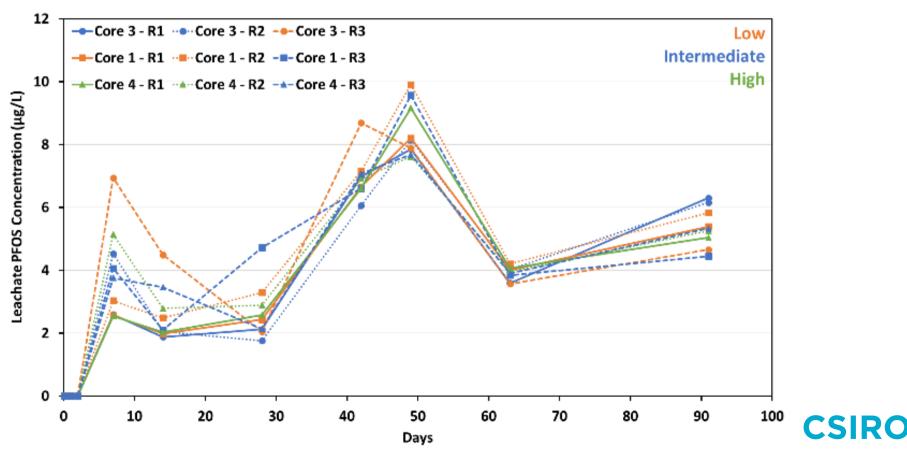
Leaching Environmental Assessment Framework (LEAF) 1315

Interval	Interval Duration (h)	Interval Duration (d)	Cumulative leaching time (d)
T01	2.0 ± 0.25	_	0.08
T02	23.0 ± 0.5	-	1
T03	23.0 ± 0.5	_	2
T04	—	5.0 ± 0.1	7
T05	—	7.0 ± 0.1	14
T06	—	14.0 ± 0.1	28
T07	—	14.0 ± 0.1	42
T08	—	7.0 ± 0.1	49
Т09	—	14.0 ± 0.1	63
Т09	-	28 ± 0.1	91

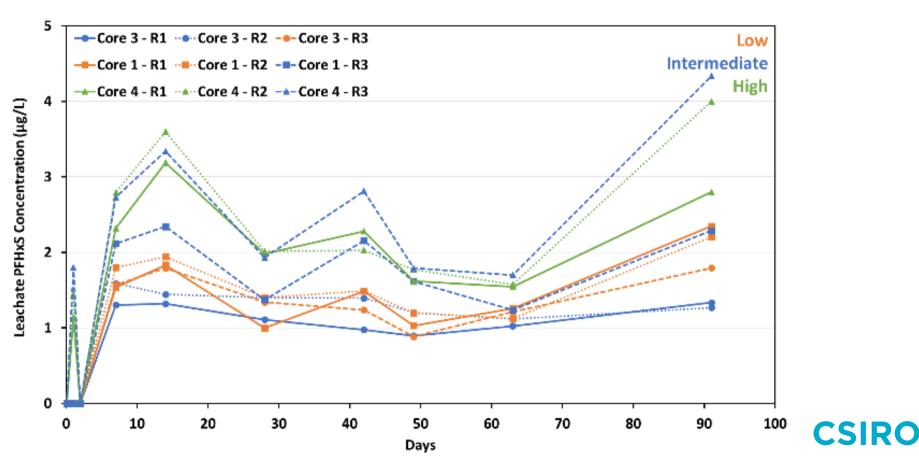
Leachate PFAS concentration (intact concrete)



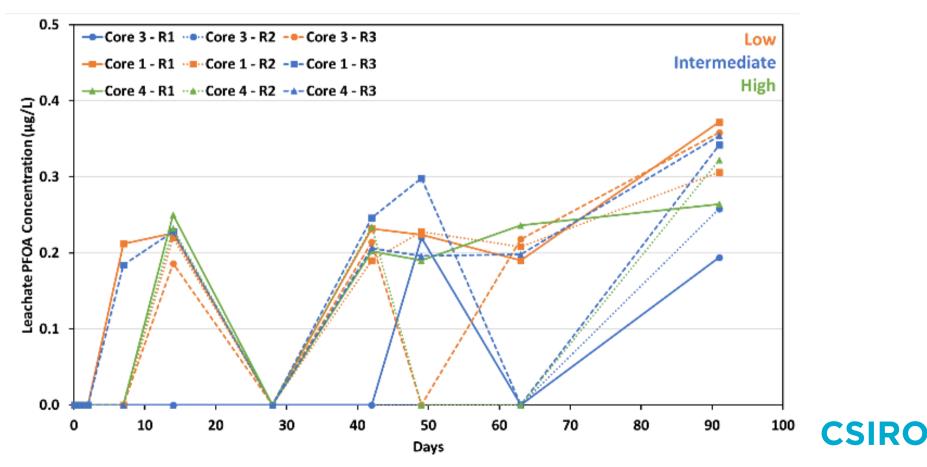
Leachate PFOS concentration (intact concrete)



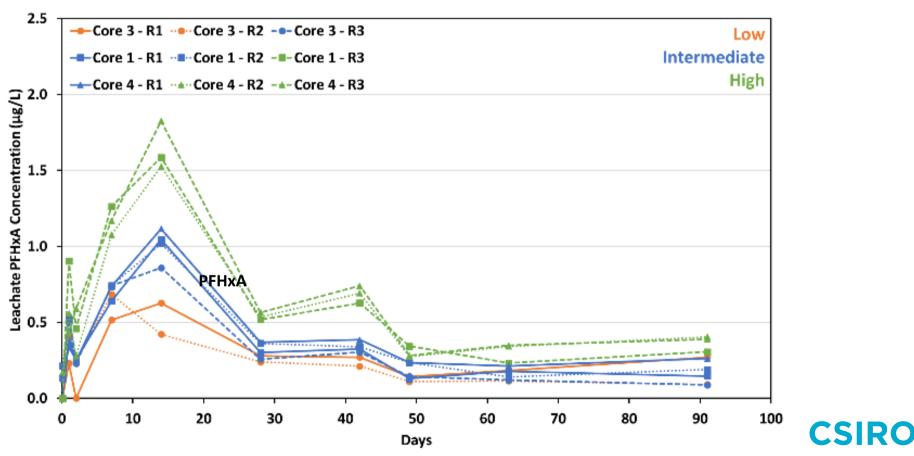
Leachate PFHxS concentration (intact concrete)



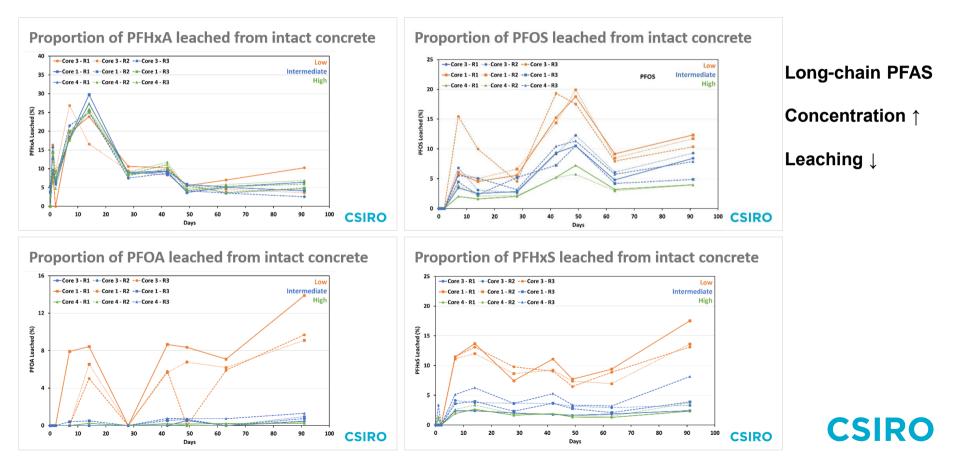
Leachate PFOA concentration (intact concrete)



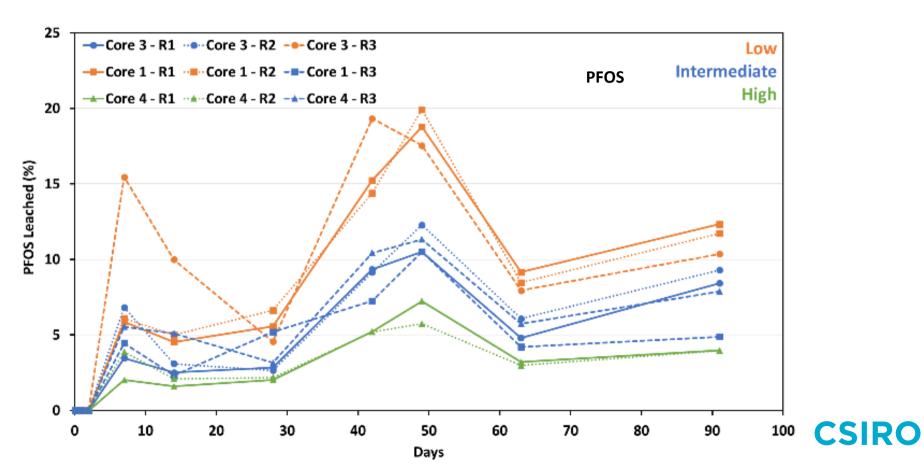
Leachate PFHxA concentration (intact concrete)



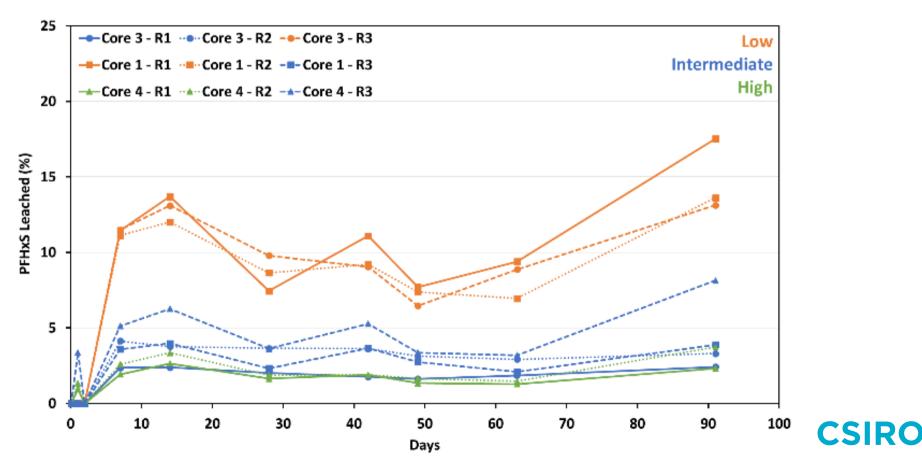
Proportion of PFAS leached from intact concrete



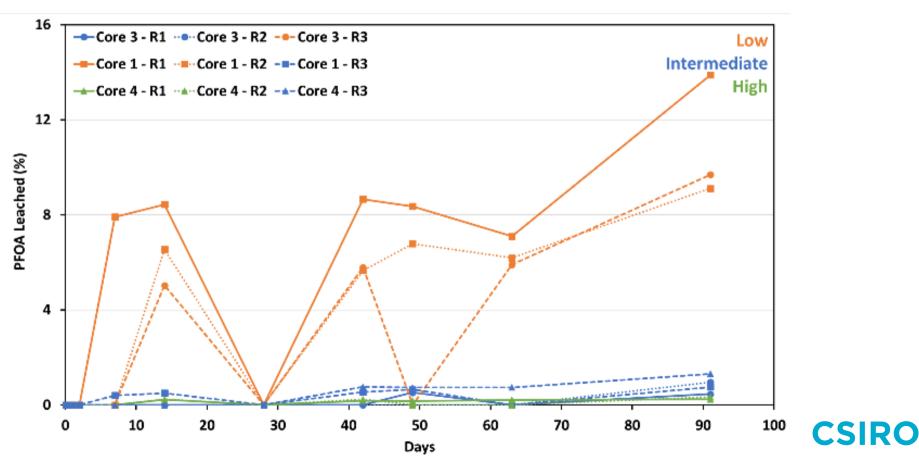
Proportion of PFOS leached from intact concrete



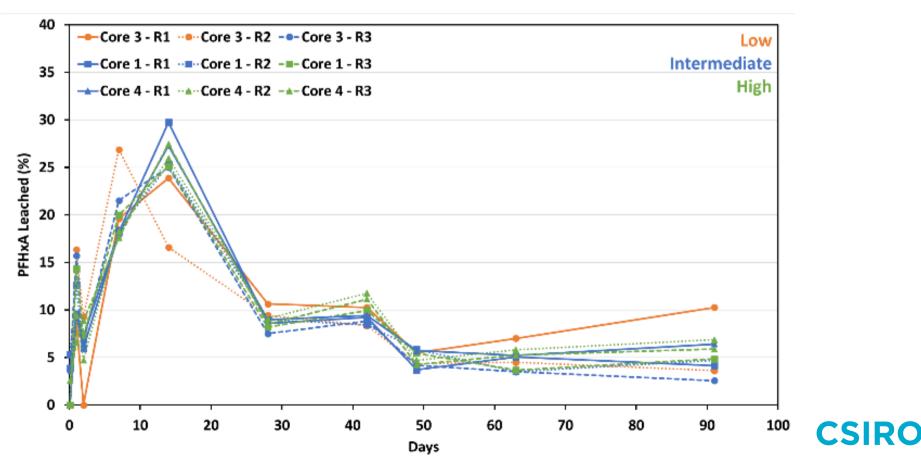
Proportion of PFHxS leached from intact concrete



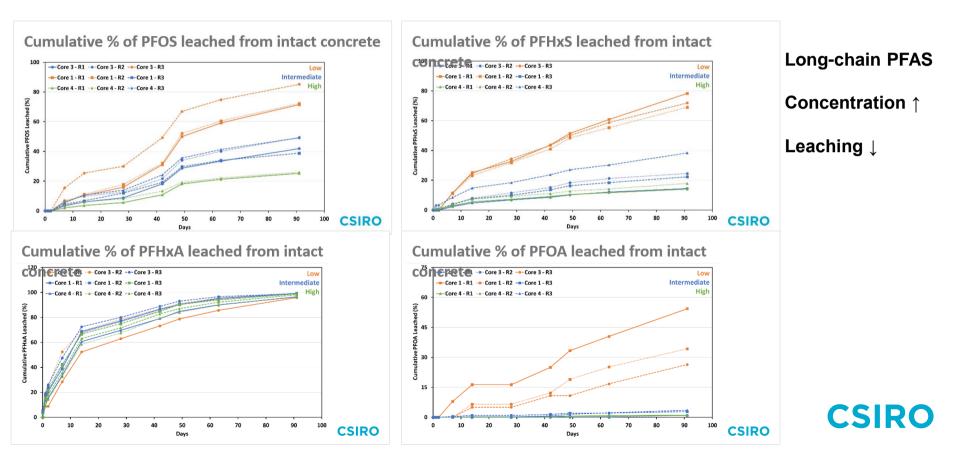
Proportion of PFOA leached from intact concrete



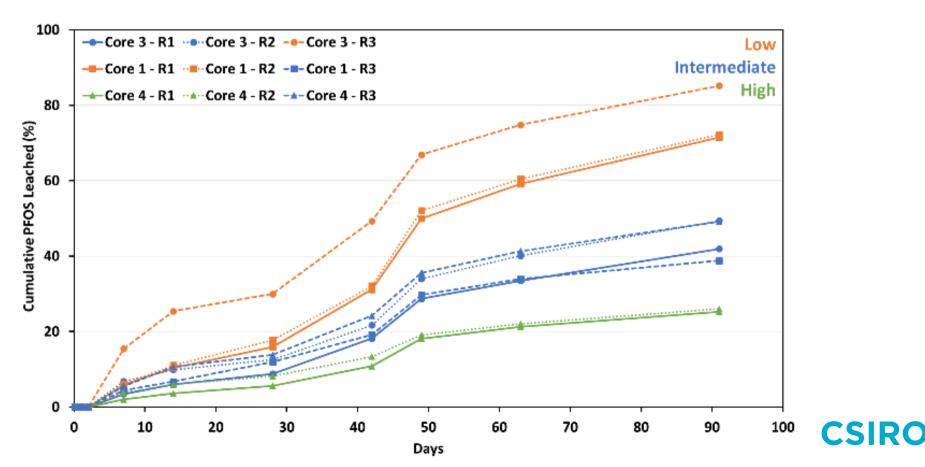
Proportion of PFHxA leached from intact concrete



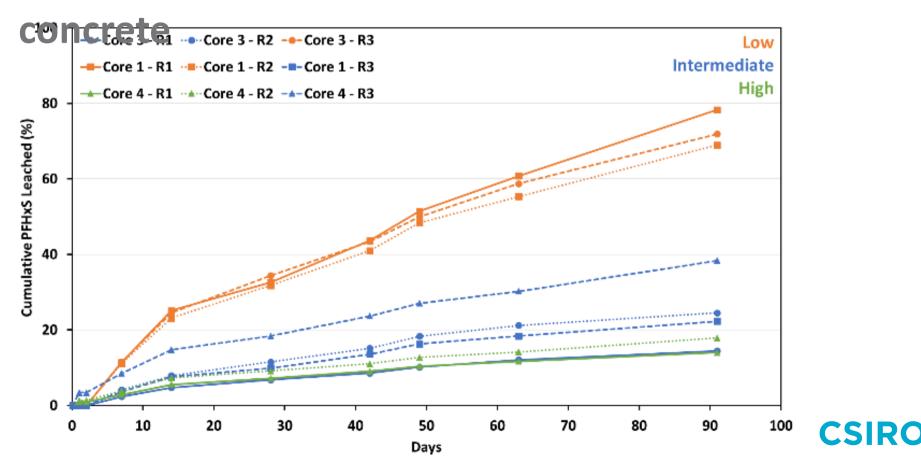
Cumulative % of PFAS leached from intact concrete



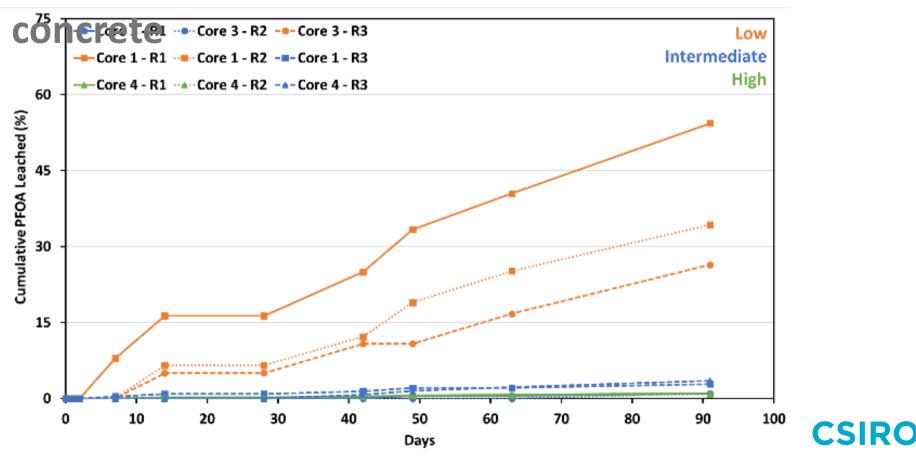
Cumulative % of PFOS leached from intact concrete



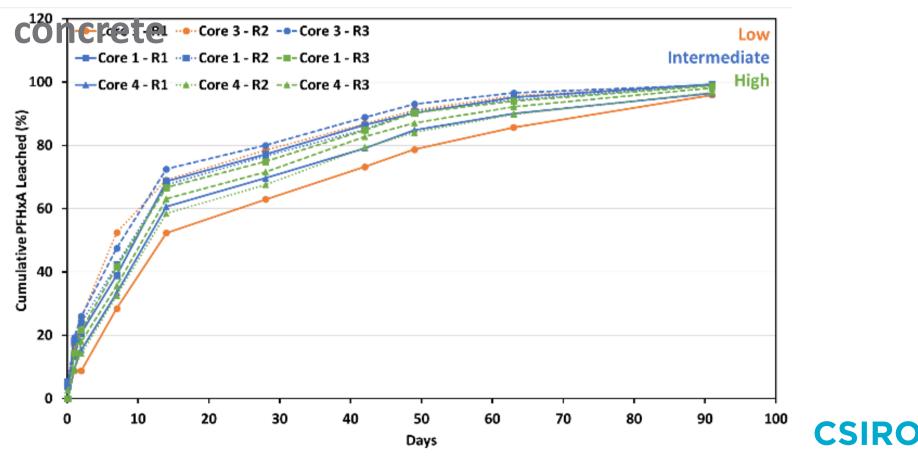
Cumulative % of PFHxS leached from intact



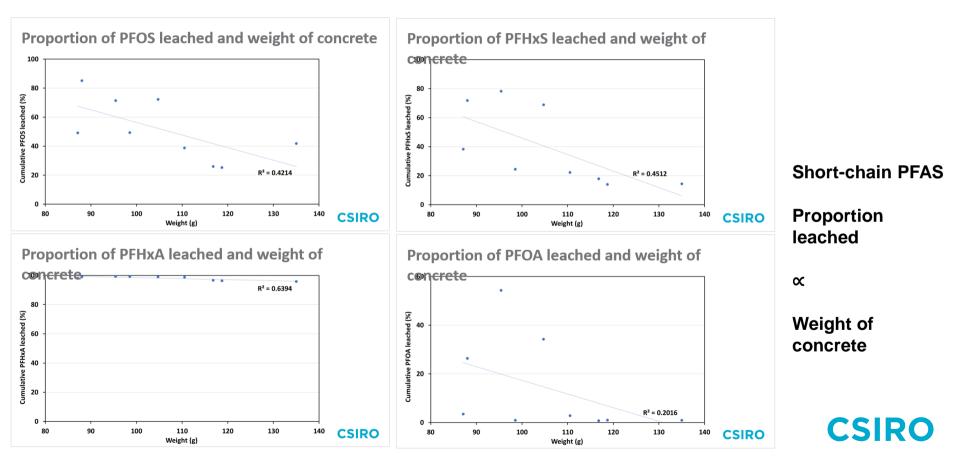
Cumulative % of PFOA leached from intact



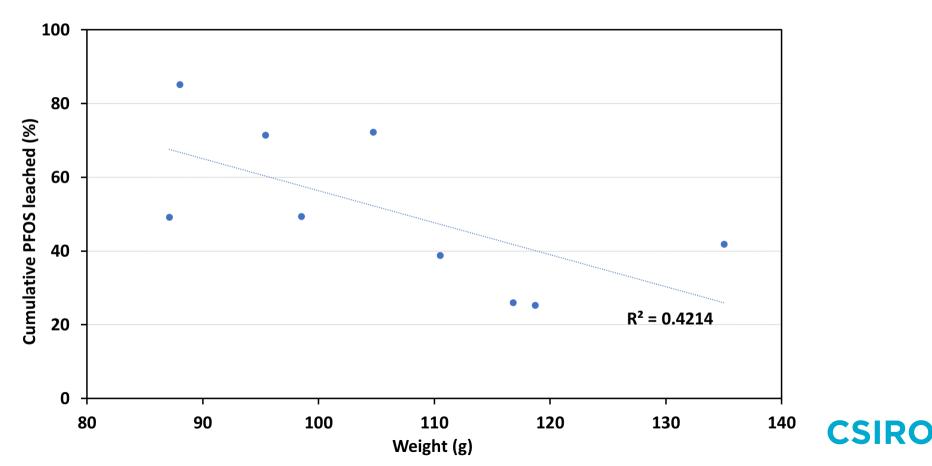
Cumulative % of PFHxA leached from intact



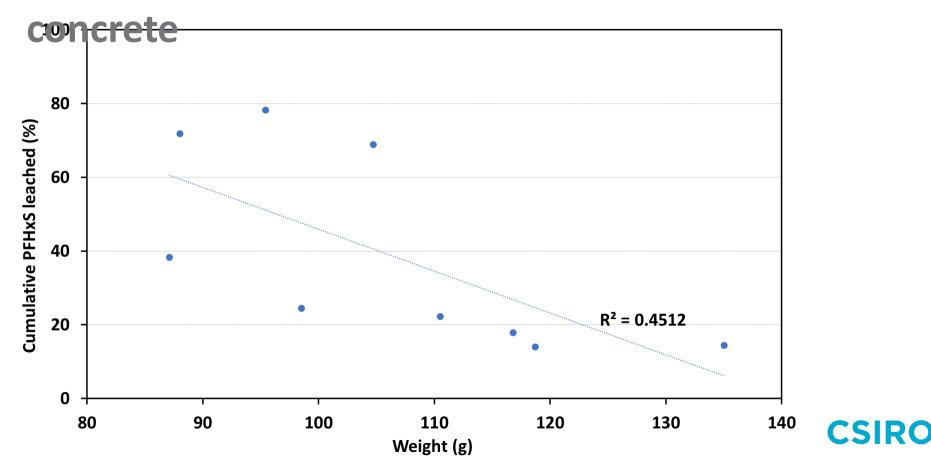
Proportion of PFAS leached and weight of concrete



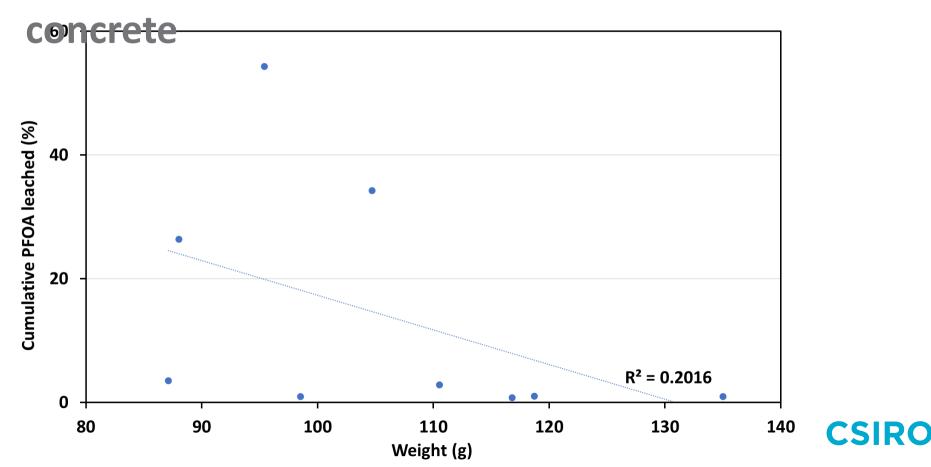
Proportion of PFOS leached and weight of concrete



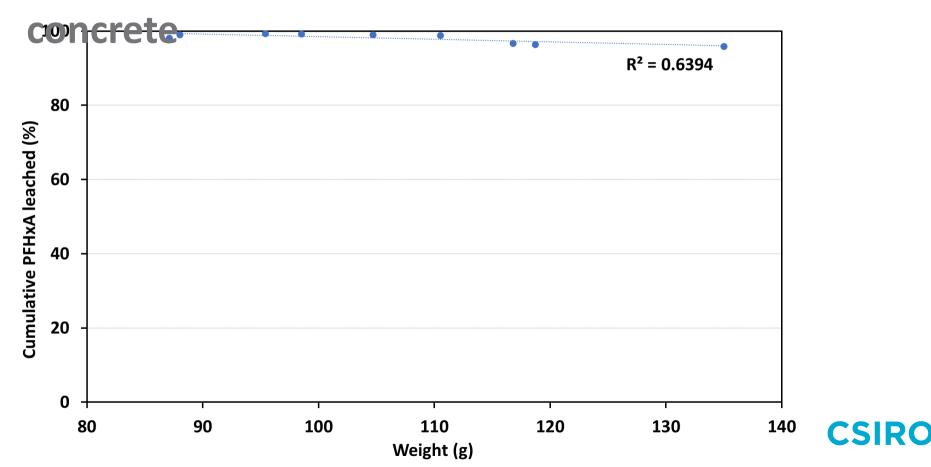
Proportion of PFHxS leached and weight of



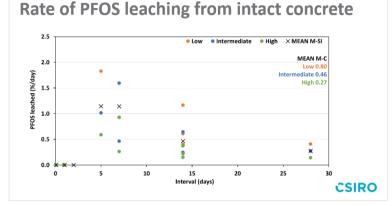
Proportion of PFOA leached and weight of



Proportion of PFHxA leached and weight of

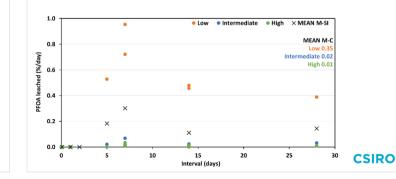


Rate of PFAS leaching from intact concrete

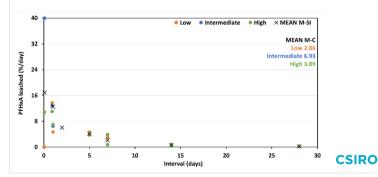


2.5 V MEAN M-S Intermediate High MEAN M-C 2.0 Low 0.76 PFHxS leached (%/day) Intermediate 0.35 High 0.28 1.5 х 1.0 : 0.5 0.0 🗴 5 10 15 20 25 **CSIRO** Interval (davs)

Rate of PFOA leaching from intact concrete



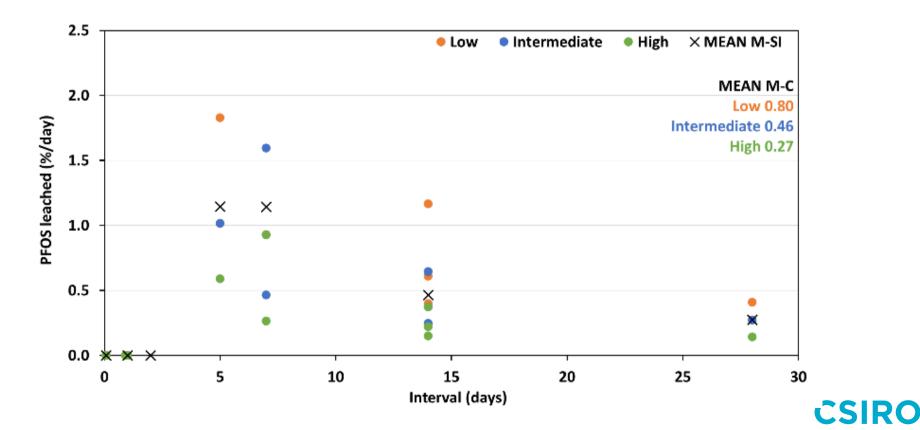
Rate of PFHxA leaching from intact concrete



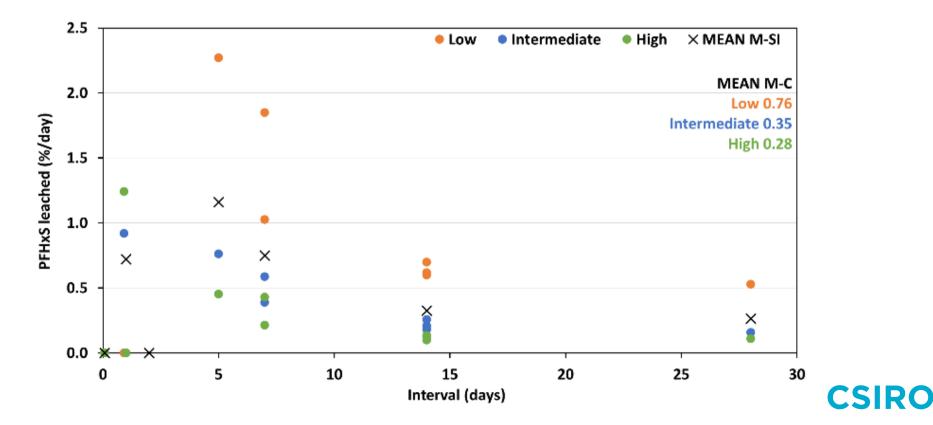
Rate of PFHxS leaching from intact concrete

CSIRO

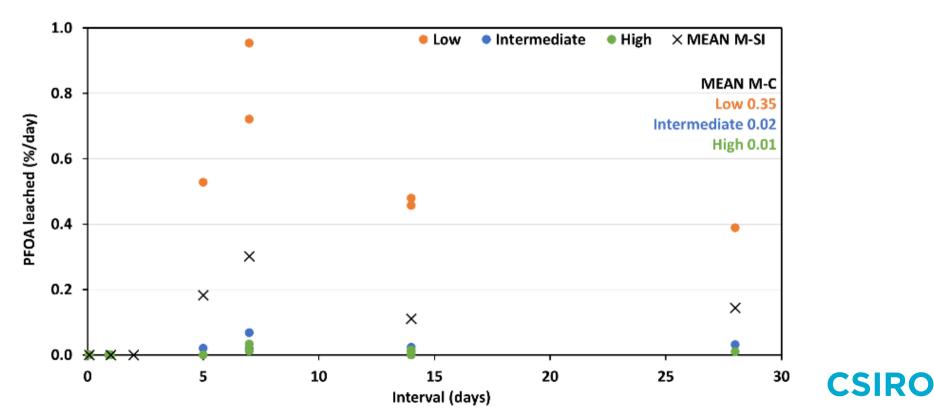
Rate of PFOS leaching from intact concrete



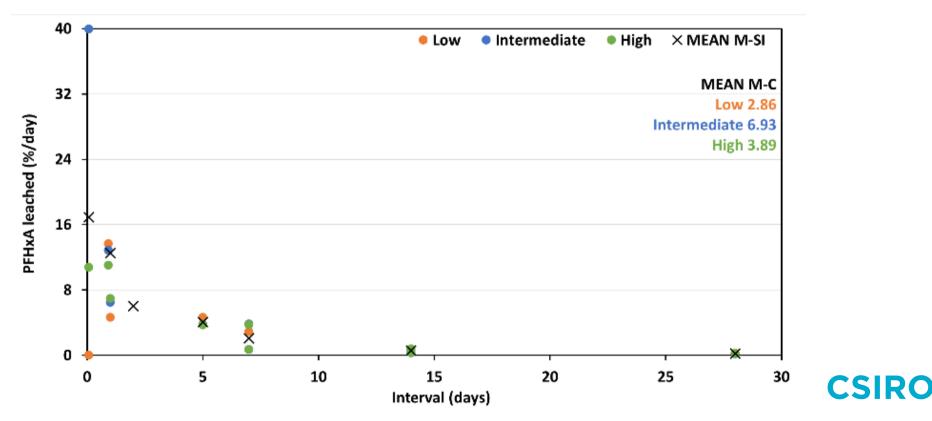
Rate of PFHxS leaching from intact concrete



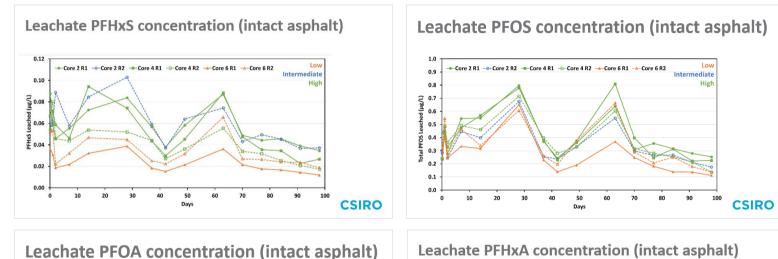
Rate of PFOA leaching from intact concrete

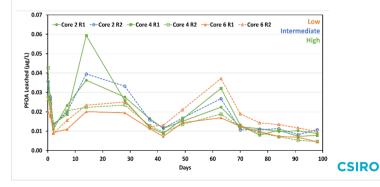


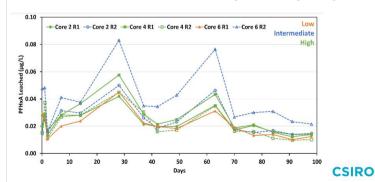
Rate of PFHxA leaching from intact concrete



Leachate PFAS concentration (intact asphalt)

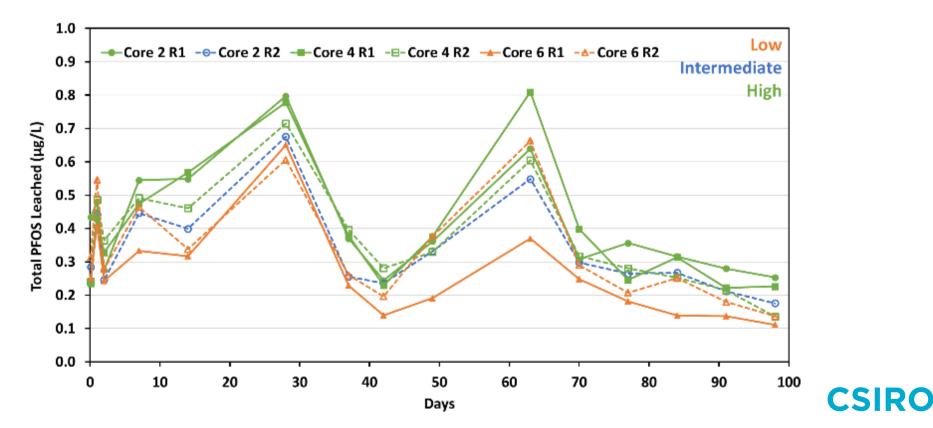




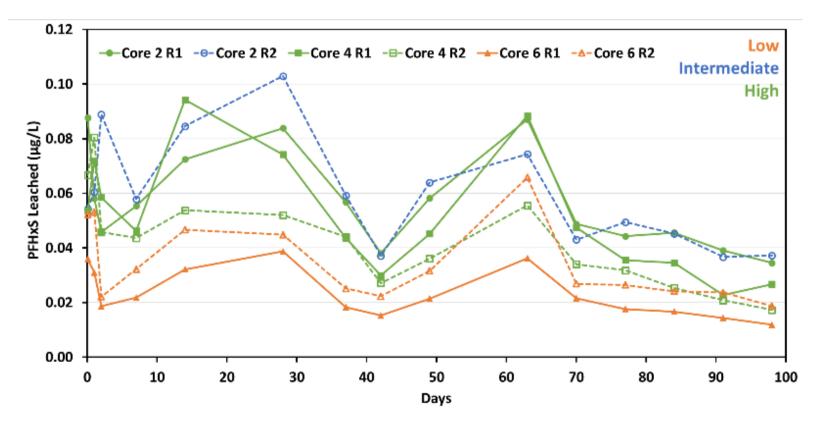


CSIRO

Leachate PFOS concentration (intact asphalt)

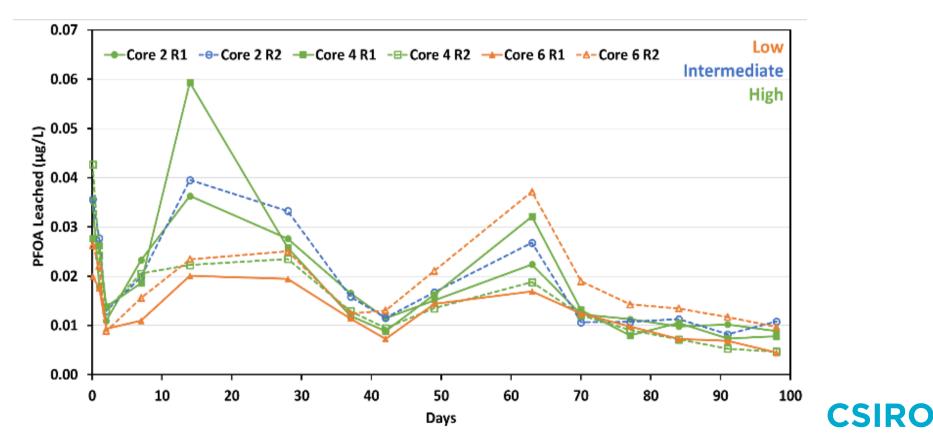


Leachate PFHxS concentration (intact asphalt)

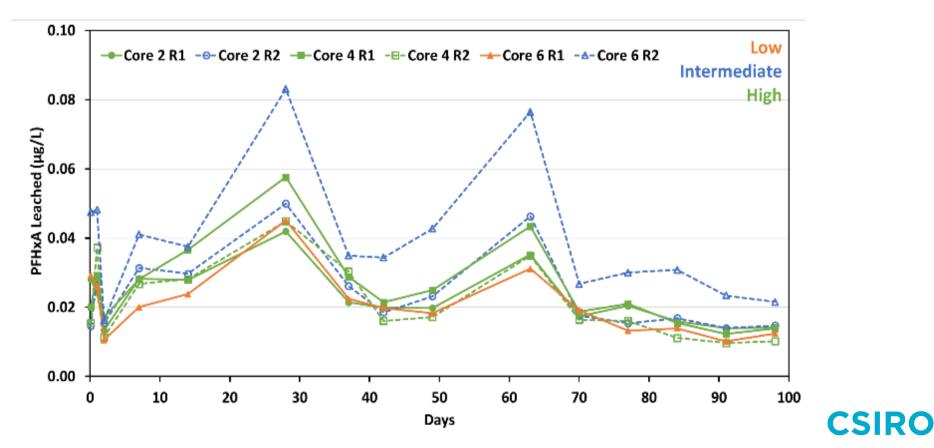


CSIRO

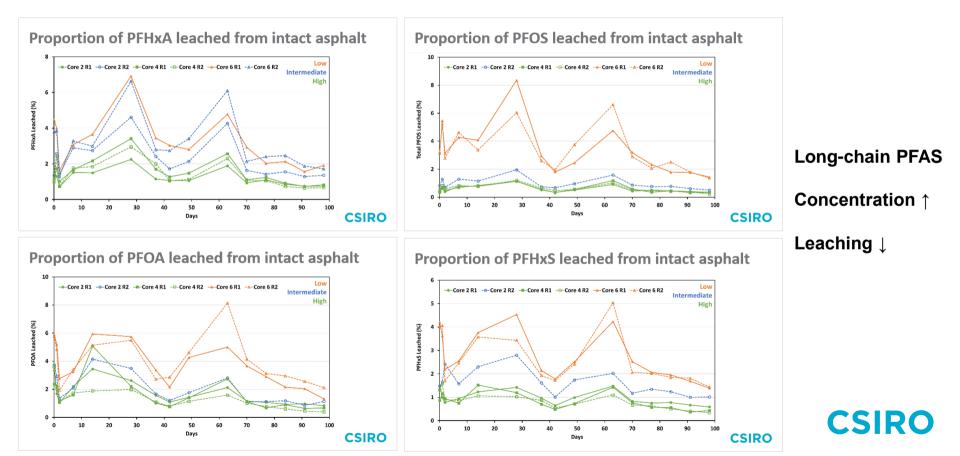
Leachate PFOA concentration (intact asphalt)



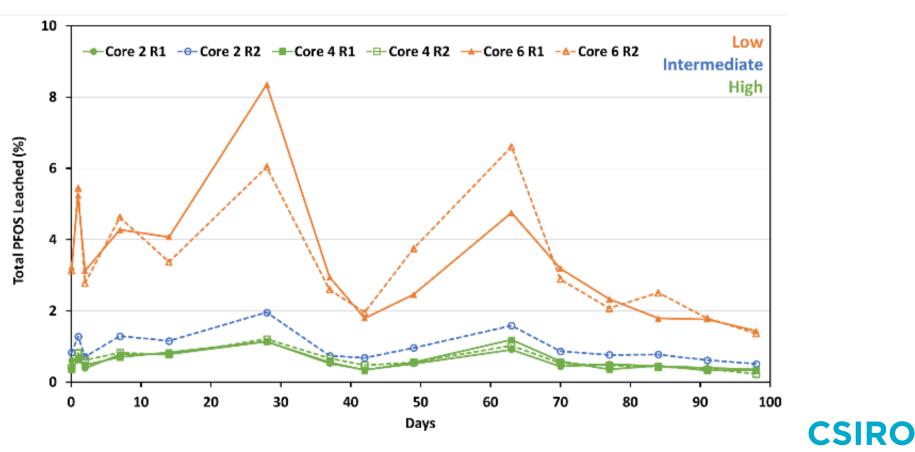
Leachate PFHxA concentration (intact asphalt)



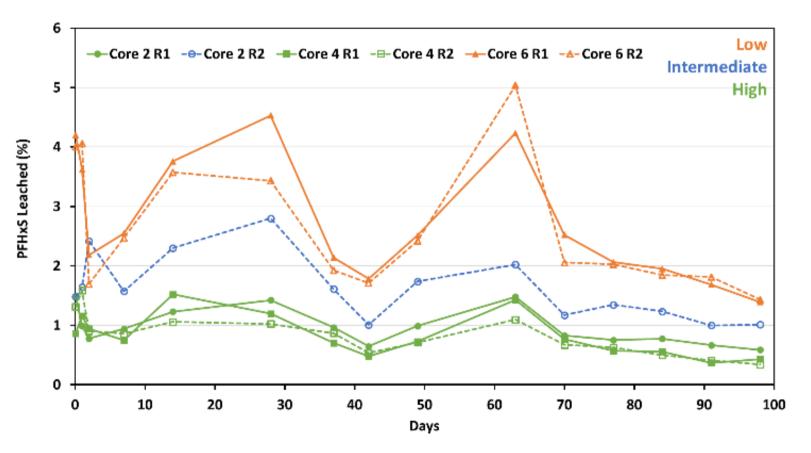
Proportion of PFAS leached from intact asphalt



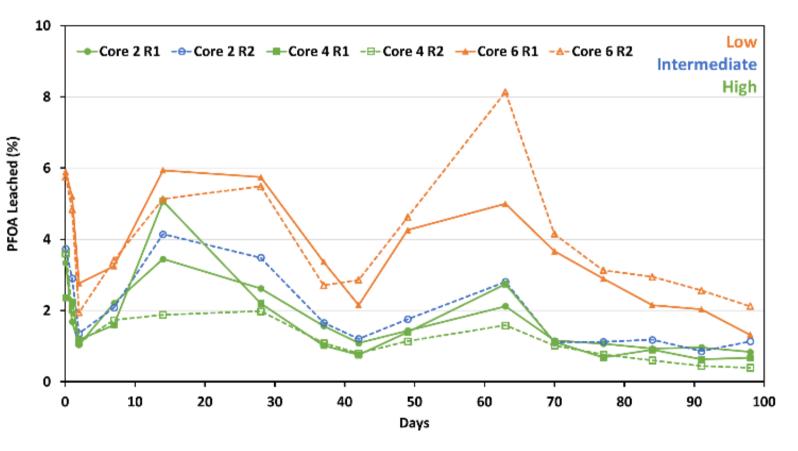
Proportion of PFOS leached from intact asphalt



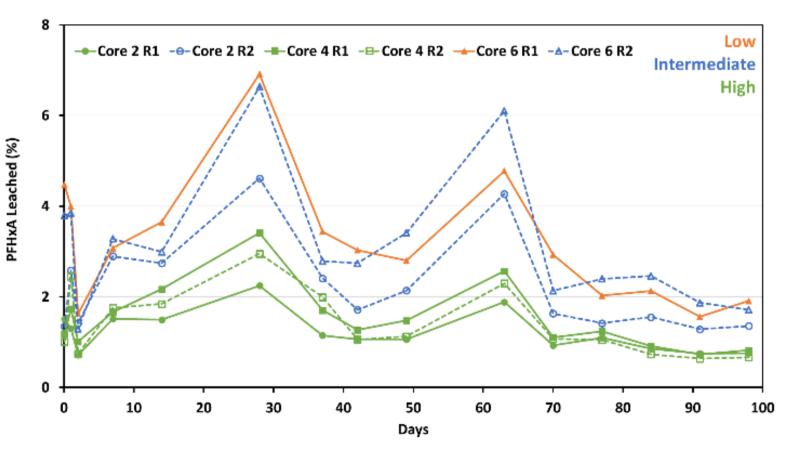
Proportion of PFHxS leached from intact asphalt



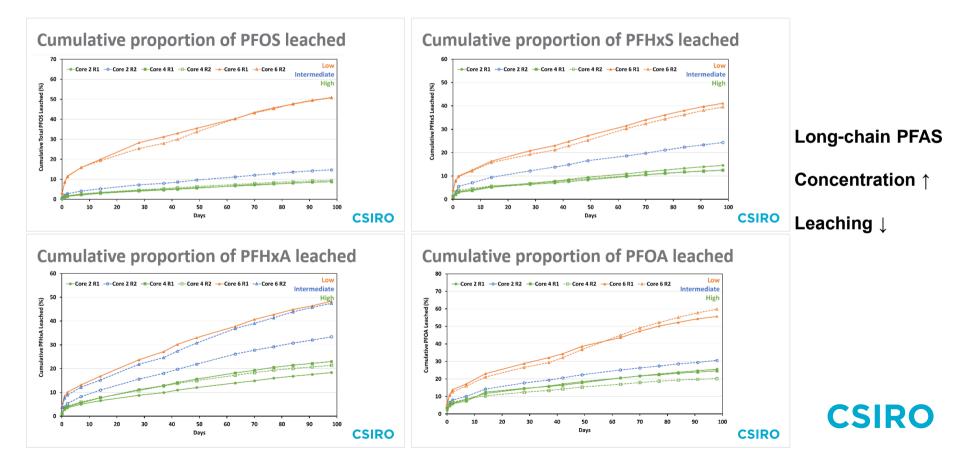
Proportion of PFOA leached from intact asphalt



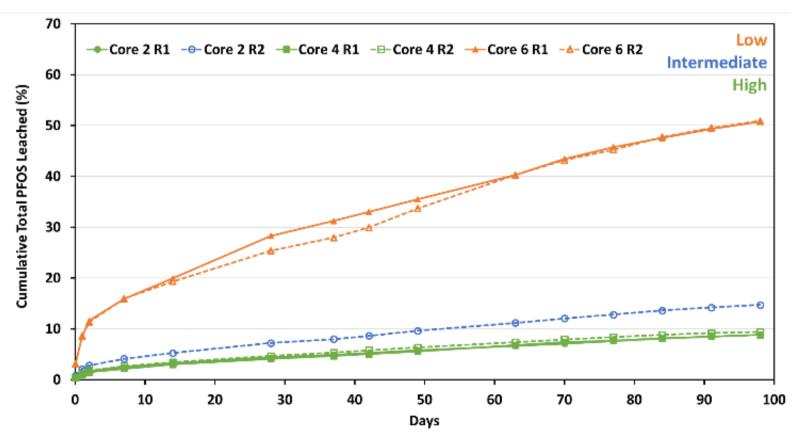
Proportion of PFHxA leached from intact asphalt



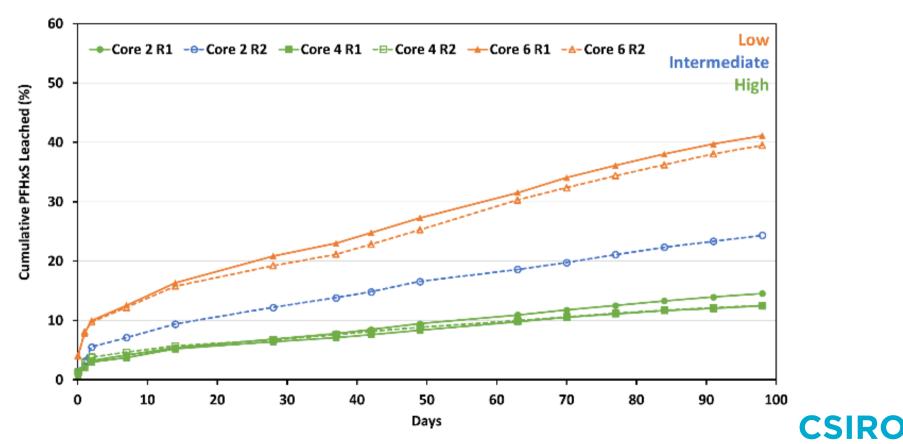
Cumulative % of PFAS leached from intact asphalt



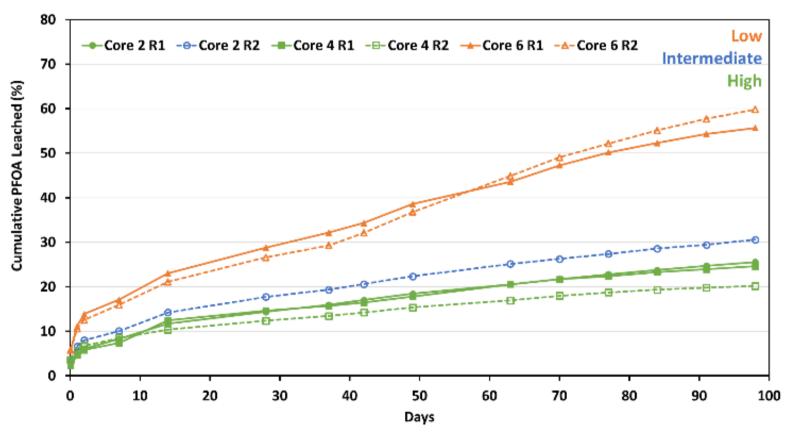
Cumulative proportion of PFOS leached



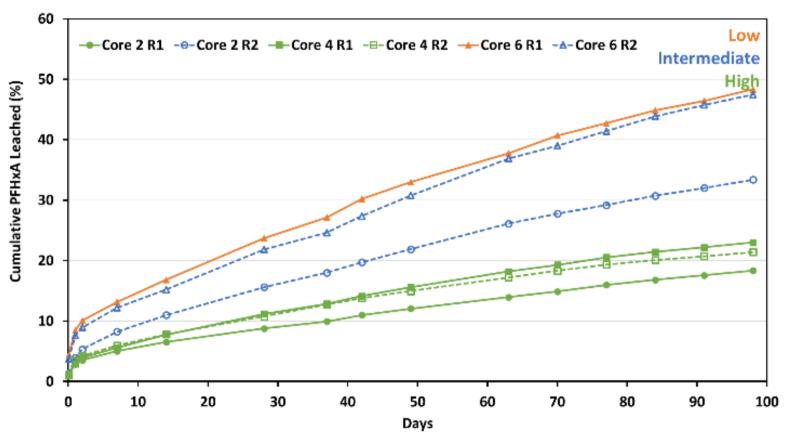
Cumulative proportion of PFHxS leached



Cumulative proportion of PFOA leached



Cumulative proportion of PFHxA leached

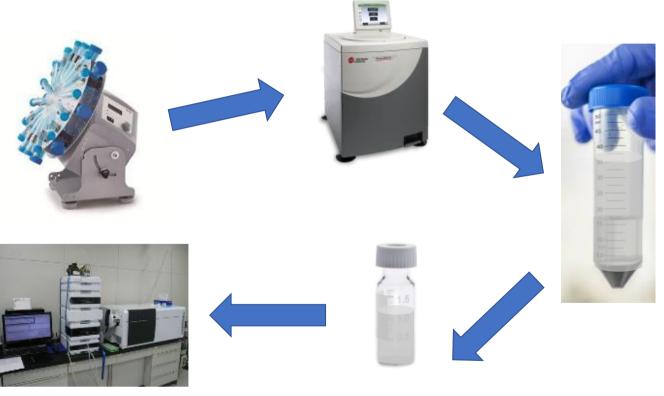


PFAS leaching from powdered concrete and asphalt



PFAS leaching from powdered concrete and asphalt

ASLP LEAF 1313



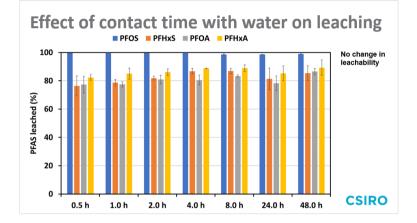


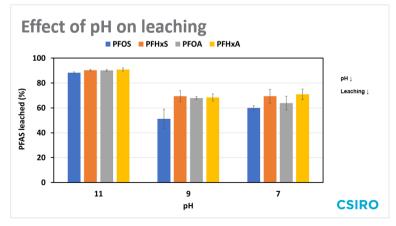
Experimental conditions for ASLP leaching

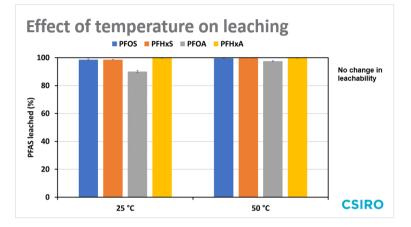
Variable	Shaking time (h)	Temperature	Particle size	Liquid/Solid (L/S) ratio
Contact time	0.5, 1, 2, 4, 8, 24, 48	25 °C	< 2 mm	20
Particle size	24	25 °C	< 2 mm, 2-20 mm, > 20 mm	20
Temperature	24	25 °C / 50 °C	< 2 mm	20

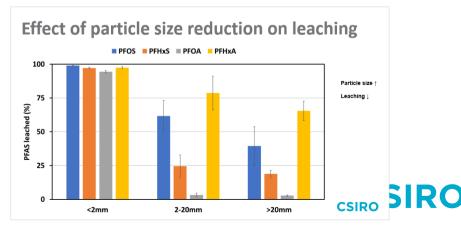


PFAS leaching from powdered concrete



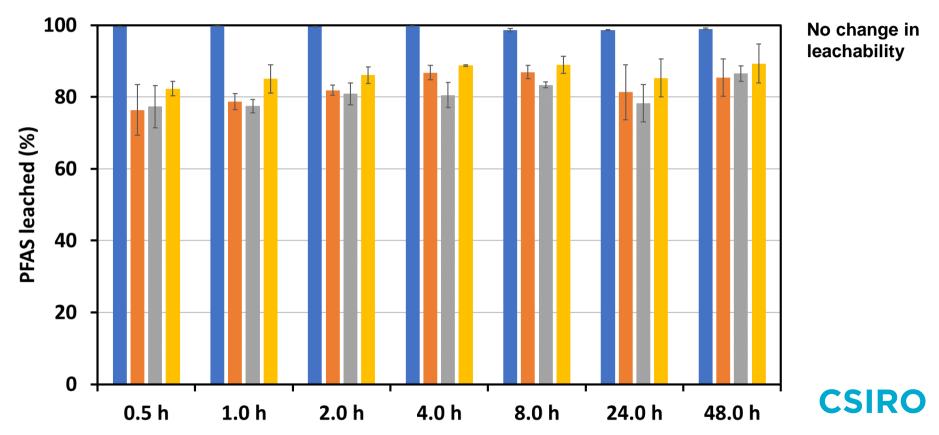






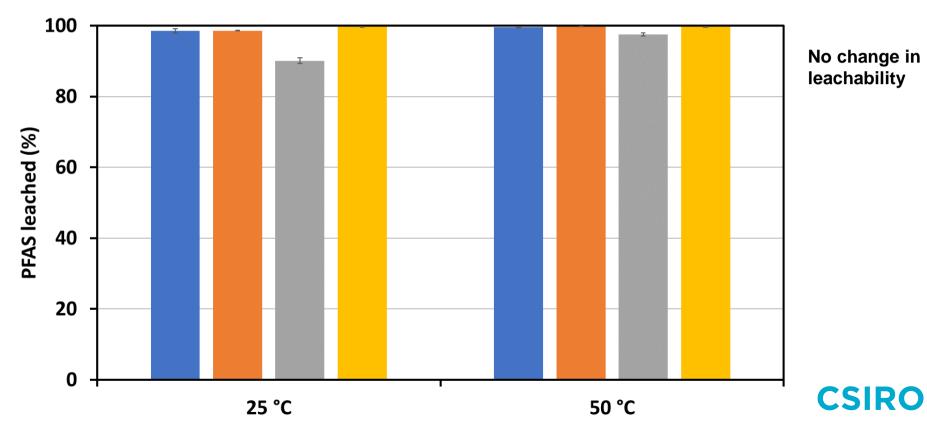
Effect of contact time with water on leaching

PFOS PFHxS PFOA PFHxA



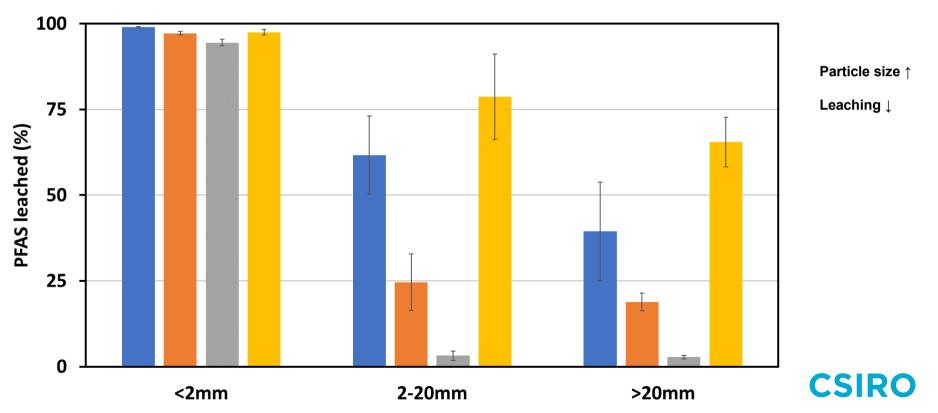
Effect of temperature on leaching

■ PFOS ■ PFHxS ■ PFOA ■ PFHxA



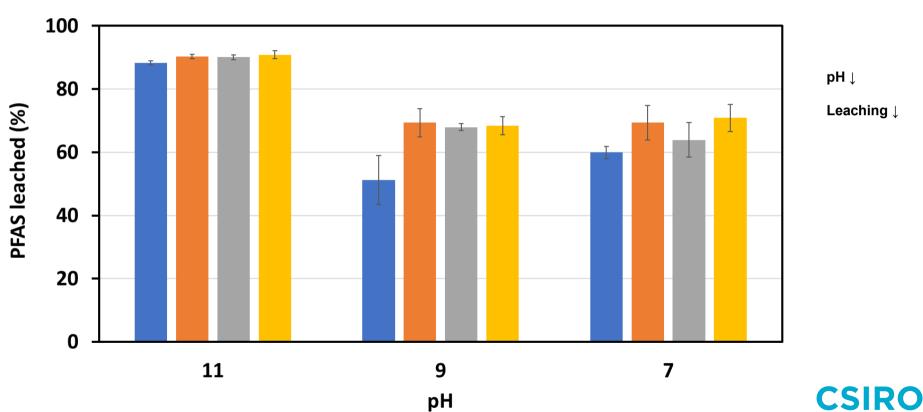
Effect of particle size reduction on leaching

PFOS PFHxS PFOA PFHxA



Effect of pH on leaching

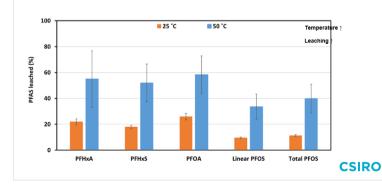
PFOS PFHxS PFOA PFHxA



PFAS leaching from powdered asphalt

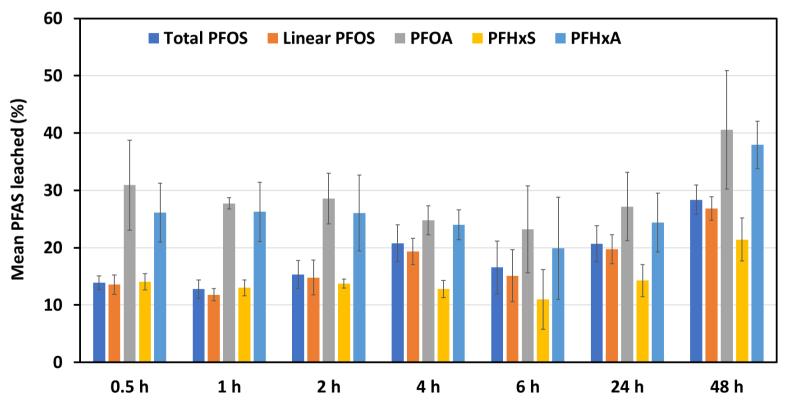


Effect of temperature on leaching

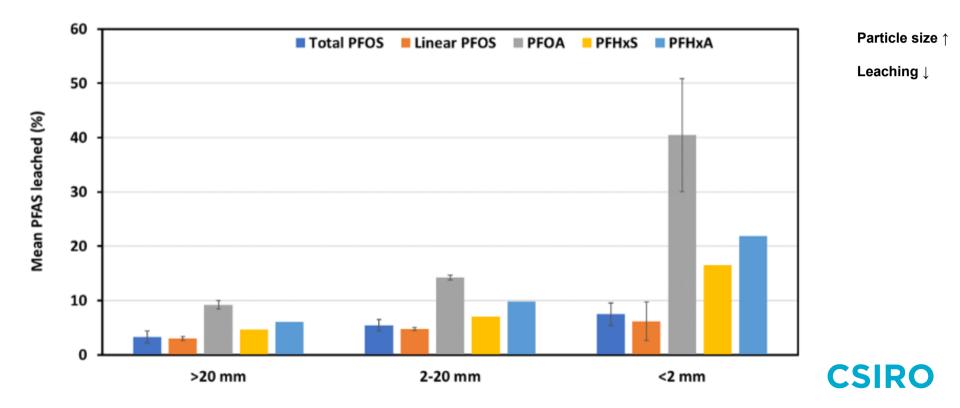




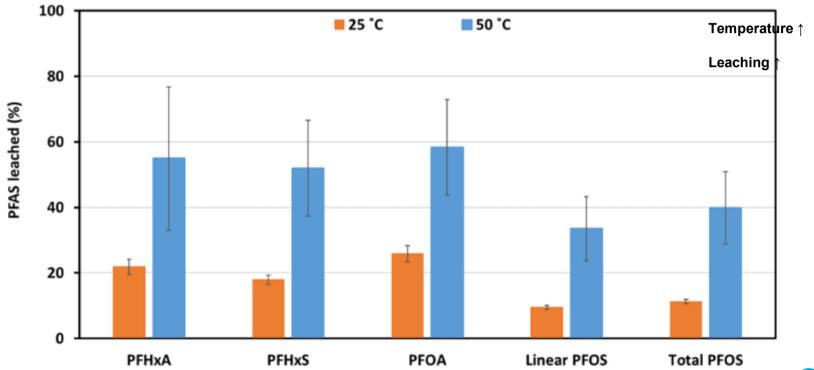
Effect of contact time with water on leaching



Effect of particle size reduction on leaching



Effect of temperature on leaching



Key takeaways

General findings

- · Significant variability in PFAS concentration in concrete/asphalt;
- PFAS readily leachable from contaminated concrete/asphalt;
- Concrete more leachable than asphalt;

CSIRO

• PFAS chemistry dominant factor in determining their leaching from concrete/ asphalt.

CSIRO

PFAS leaching from Intact concrete/asphalt

- PFAS leaching time-dependent; and
- Low-concentration monoliths leached a relatively greater proportion of PFAS.

PFAS leaching from powdered concrete/asphalt

- PFAS leaching time-independent;
- · Particle size reduction increased PFAS leaching;
- · Increased temperature increased PFAS leaching from asphalt; and
- Lowering the pH of concrete reduced PFAS leaching.



General findings

- Significant variability in PFAS concentration in concrete/asphalt;
- PFAS readily leachable from contaminated concrete/asphalt;
- Concrete more leachable than asphalt;
- PFAS chemistry dominant factor in determining their leaching from concrete/ asphalt.

PFAS leaching from Intact concrete/asphalt

- PFAS leaching time-dependent; and
- Low-concentration monoliths leached a relatively greater proportion of PFAS.



PFAS leaching from powdered concrete/asphalt

- PFAS leaching time-independent;
- Particle size reduction increased PFAS leaching;
- Increased temperature increased PFAS leaching from asphalt; and
- Lowering the pH of concrete reduced PFAS leaching.



Acknowledgements

- Funding support: Department of Defence
- Operational and logistical support: Dr Karl Bowles, Mr Garbis Avakian, Ms Nelma Akhund, Mr Stephen Corish, Mr Darren Skuse and Mr Mark Bauer.
- CSIRO researchers: Dr Grant Douglas, Dr Rai Kookana, Dr Greg Davis, Dr Mike Williams, Dr Jason Kirby
- CSIRO Laboratory support: Dr Trinh Nguyen, Ms Thanh Huong Nguyen, Ms Manvi Gandhi, Ms Claire Wright, Ms Catherine Fiebiger, Ms Caroline Johnston.





Thank you

Prashant.Srivastava@csiro.au

Australia's National Science Agency



REMISSION CONTAMINANTS

OCTOBER 15-17, 2024

Practical Considerations of Linear and Branched PFAS Isomers for Evaluating PFAS Fate, Transport and Attenuation Track 7 PFAS Fate and Transport Considerations

Dora Chiang, Ph.D., P.E. Jacobs





Distributions of PFAS Isomers in the Environment

Significance of PFAS Isomers on Site Investigation

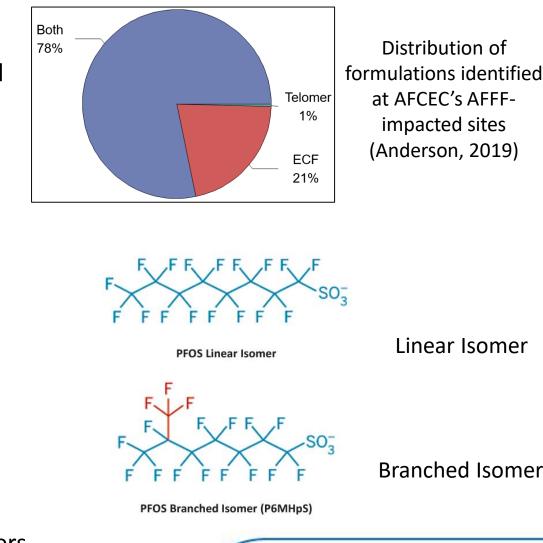
PFAS Isomer Distribution Impacted by Remediation

Case Study

Summary

Introduction

- ECF processes produced linear and branched perfluorinated isomers from the 1950s through 2002
- Research conducted on 3M Co. ECF process (Benski et al, 2010; Buck el.al, 2011; Londhe et. al, 2022):
 - PFOA 78% linear and 22% branched isomers
 - PFOS 70% linear and 30% branched isomers
- Following the 2002 phase-out, linear PFOA (L-PFOA) and other L-PFCAs have predominantly been produced by the fluorotelomerization process
- Different branched isomers exhibit varying chemical and physical properties, consequently leading to distinct toxicological, persistence and bioaccumulation potentials (Fang et al. 2016; Houde et al. 2008).
- Most sites have been impacted by both ECF and FT formulations, the significances of linear and branched isomers on conceptual site model are rarely investigated.



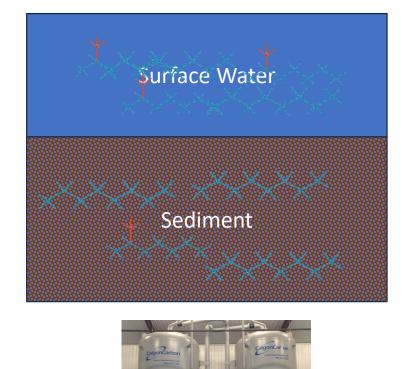
#RemTE



Environmental Factors Contributing to the Variations of Linear vs Branched Isomer Ratios

Once PFAS isomers were released into the environment, both natural and anthropogenic processes can modify the ratio between linear and branched isomers.

- Interactions with hydrogeological and geochemical processes in situ. For example, L-PFAS sorption on aquifer materials(e.g., sediment) would lead to higher Br-PFAS ratios (enriched Br-PFAS) in water matrices
- Isomer-specific precursor transformation. As terminal products (e.g. PFOS) are likely generated through more than one pathways or precursors, the final isomer ratio is determined by several upstream reactions
- Interactions with remediation activities at the site (preference of removing/degrading linear or branched PFAS)
- However, the isomer research is also limited by the availability of branched isomer standards

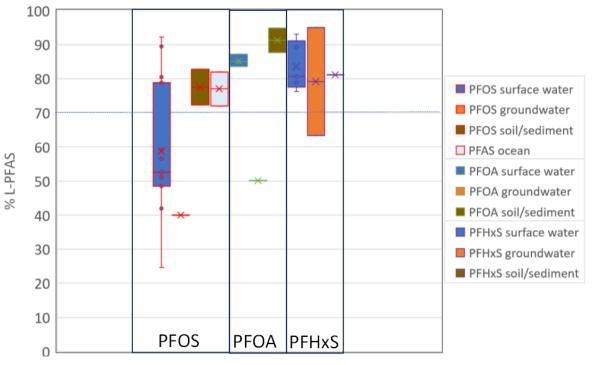


#RemTEC

Br-PFOS/L-PF

Br-PFOS/L-PFOS=0.63

PFAS Isomers Distribution in the Environment



(Original data: Schultz et al. 2020)

- L-PFOS, L-PFOA and L-PFHxS were found enriched in sediment and soil due to higher hydrophobic sorption thank branched PFAS
- L-PFOS% in surface water exhibit wide variability with averages lower than 70%, indicating the potential enrichment of branched PFOS (Br-PFOS) in surface water because of the L-PFOS sorption onto soil/sediment
- Br-PFOS "precursors" are more susceptible to transformation into Br-PFOS, in soil microcosm (Liu et al 2019)
 - Preferential transformation of branched precursors can therefore lead to an enrichment of Branched terminal PFAA products.

#RemTEC

REMIE

 Overall, the distribution of PFAS data reveal *significant variations* in the concentrations of branched and linear PFAS in the environment

Significance of Remediation to L/Br-PFAS Ratios

Treatment	Mechanism	Enriched isomers in effuent	Significance	Reference
GAC	Sorption	Branched	Preferential removal of linear isomer, rapid breakthrough of branched isomers	Belkouteb et al 2020; Eschauzier et al., 2012; McCleaf et al., 2017 ; Rodowa et al, 2020
Anion exchange resin	Electrostatic interactions	None	Br- and L- have similar electrostatic interactions	McCleaf el. Al., 2017; Park et. al., 2020
Reductive defluorination	Reductive defluorination	Linear	Greater Br- isomer degradation than L- isomers	Ochoa-Herrera et al. 2008
Electron beam	Electron affinity	Linear	Greater Br- isomer degradation than L- isomers	Trojanowicz et al. 2020
UV-Sulfite	Degradation	Linear	Tertiary -CF3 in Br-isomer is more susceptible to degradation	Yamamoto et al. 2007; Gu et al. 2016; Liu et al. 2020
Electrochemical oxidation	PFAS sorption followed by direct electron transfer and mineralization by hydroxyl radicals	Branched	Br-PFAS enriched in treated water because sorbed L-PFAS are more susceptible to electrochemical oxidation	Chaplin 2014; Radjenovic and Sedlak 2015; Uwayezu et al. 2021; Wang et al. 2020

Branched Isomer Analysis

linear PFOS

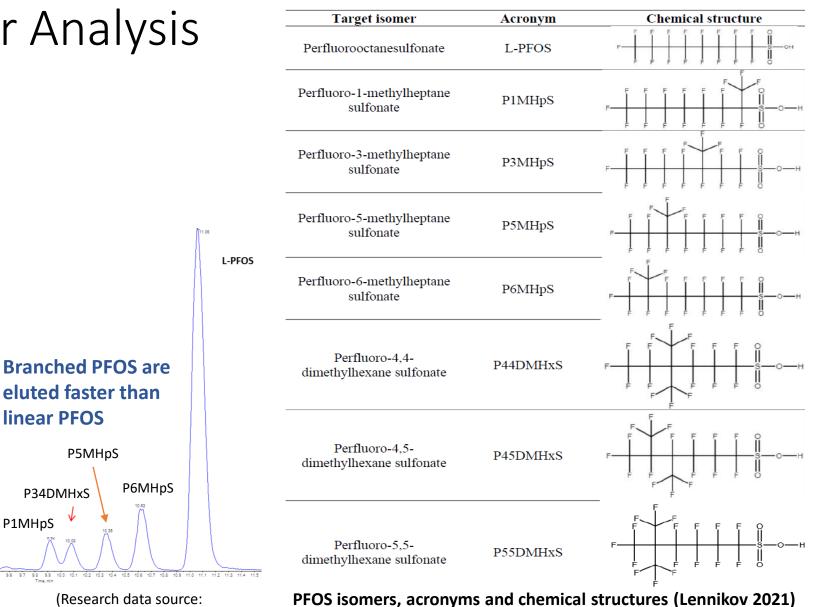
P34DMHxS

Jenny Zenobio, Jacobs)

P1MHpS

9.6 9.7 9.8 9.9 10.0

- PFAS isomers are difficult to be ۲ separated and identified individually by established analytical methods
- USEPA methods currently • mandate the users to integrate branched and linear peaks together and report total concentrations rather than isomer-specific concentrations.
- The final revision of USEPA • Method 1633 still necessitates PFAS to be reported as a single result

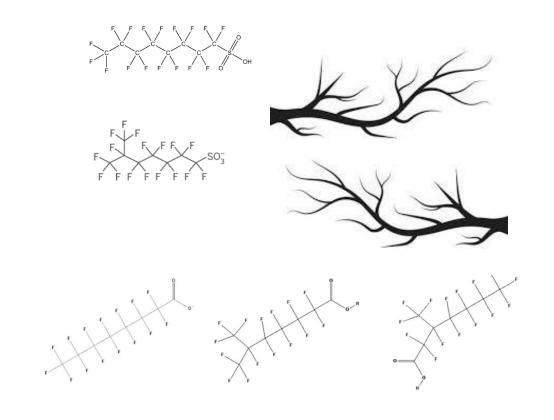


General Understanding of Linear vs Branched Isomers in Groundwater at a Site

- Linear isomers
 - PFAS in ECF-based AFFF are expected to have Br/L ratios close to 30/70,
 - Fluorotelomers and their transformation products (PFCAs) would be theoretically all linear (Br/L 0/100)
 - When a site used both ECF and FT-based AFFF, Br/L PFCA ratios can be much less than 30/70 due to linear precursor transformation into linear PFCAs

Branched isomers

- Branched isomers exhibit lower partition coefficients between soil and water (Kd) or soil organic matter and water (Koc), branched isomers are less retarded than linear isomers
- Br/L can be greater than 30/70 downgradient from the source when more L- isomers are retarded in the source area





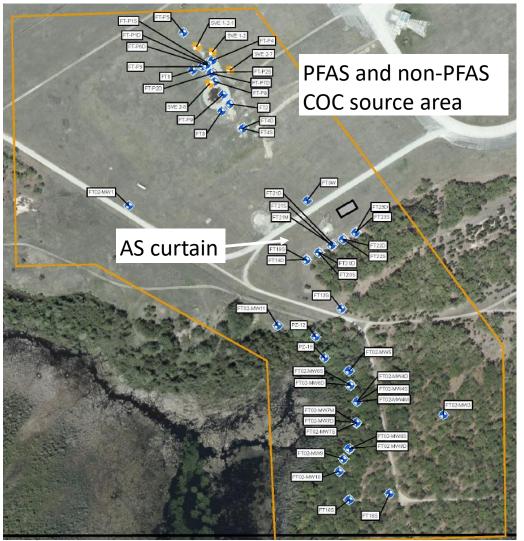


Case Study





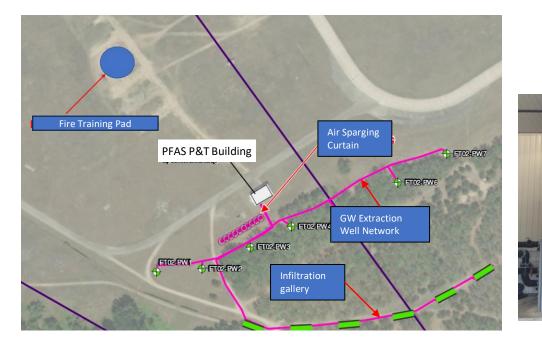
Site Background



- Non-PFAS remedial actions completed at the Site
 - SVE for contaminated soils
 - Air sparging to treat smear-zone soils
 - MNA for VOC in groundwater
 - In-situ anaerobic biodegradation pilot study for nitrate reduction proximal to the source
 - AS curtain installation downgradient of the source area to address dissolved-phase
- The PFAS groundwater treatment system was installed and has been operated for ~10 years



GAC Treatment for PFAS Removal



Groundwater Extraction and Treatment

Full-scale GAC treatment of extracted groundwater

CalgonCarbon



Estimated quantities of PFAS and spent media requiring waste management after 20 years of system operation (Data source: WSP)



Detected PFAS in Groundwater at the Site

Groups	Chain length	ECF- Linear	ECF- Branched	FT Linear	Relevance to CSM
PFCAs	C6-C14		\blacksquare		
PFSAs	C4, C6 and C8	\checkmark	\checkmark		
Fluorotelomer sulfonates	4:2, 6:2, 8:2				Intermediates of FT-based AFFF ingredient transformation
N-SP-FASA	C4-C6	\checkmark	\checkmark		
FASA	C3-C6	\checkmark	\blacksquare		
PFASi	C4-C6, C8	\checkmark	\checkmark		Possible indicator of anaerobic conditions
n:2 FASO2PA-MePS	C6, C8				Possible oxidized degradation product of FT-based AFFF active ingredient
N-CMAmP- <i>n</i> :2 FASA	C4-C6				Possible active ingredient in FT-based AFFF
N-AP-n:2 FASA	C4-C6				Possible active ingredient in FT-based AFFF

13

Focused Pilot Study to Investigate PFAS Removal Using GAC

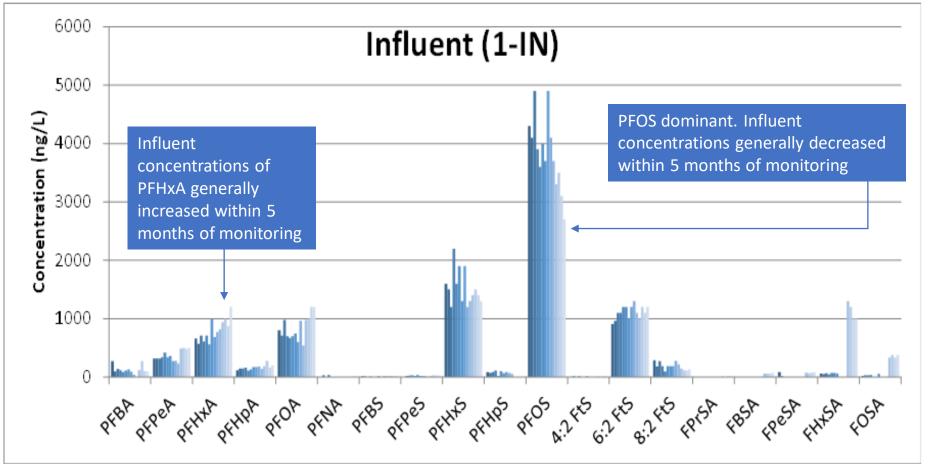
- Monitor for sorption of PFAS precursors
- Investigate linear and branched PFAS in a GAC treatment system
- Weekly sampling for 5 months



Analytical Parameter	Frequency	Laboratory	Method
PFOS and PFOA (screening purpose)	Weekly	Dr. Huang/UGA	UPLC/MS/MS
Volatile Organic Compounds	Monthly	Commercial Labs	USEPA 8260B
Total Organic Carbon	Monthly	Commercial Labs	TOC 5000A
Fluoride	Monthly	Dr. Huang/UGA	IC-MS
Optimized WAFB Site Specific PFAS List	Weekly	Dr. Field/OSU	Orthogonal HPLC MS/MS Quadrupole Time-of-Flight Mass Spectrometry
Total Oxidizable Precursor Assay (TOPA)	Weekly	Dr. Field/OSU	PFAS analyses, same as above
Particle-Induced Gamma-ray Emission (PIGE)	Weekly	Dr. Graham Peaslee/U of Notre Dame	PIGE



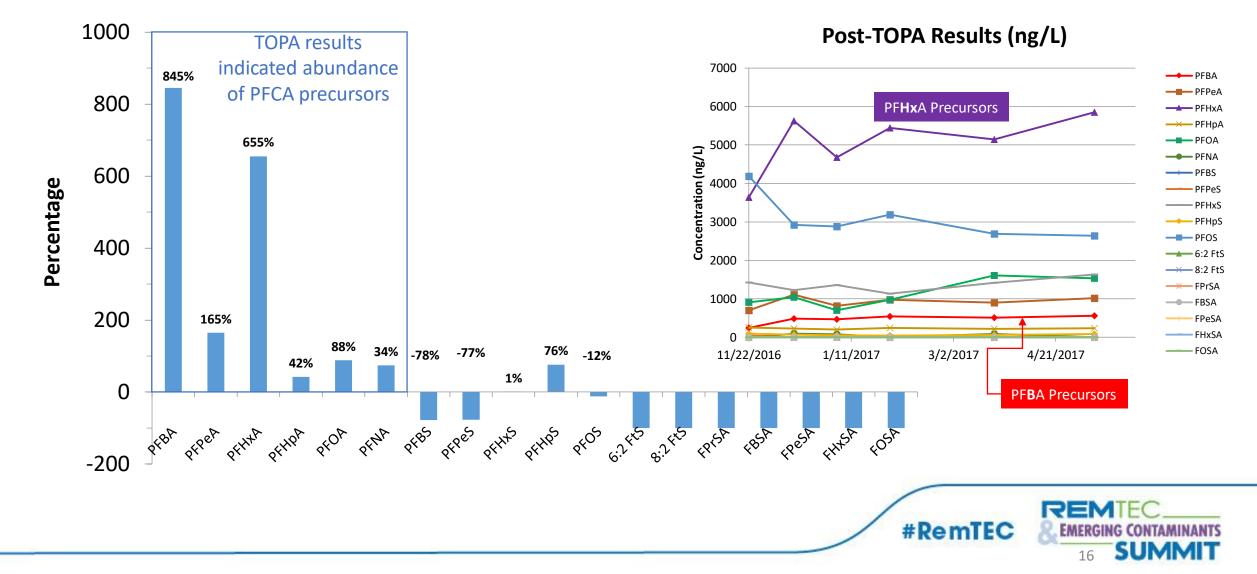
PFAS in the Influent of Pilot-Scale GAC System



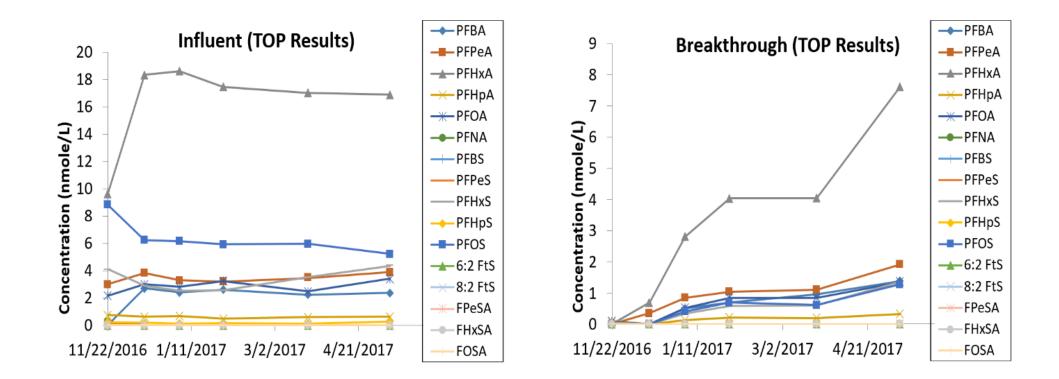
Many PFAS detected in groundwater were not detected in the influent



PFAS Precursors in the Influent of Pilot-Scale GAC System (TOP Assay Results)



Breakthrough of PFAS Precursors in the Effluent of Pilot-Scale GAC System (TOP Assay Results)





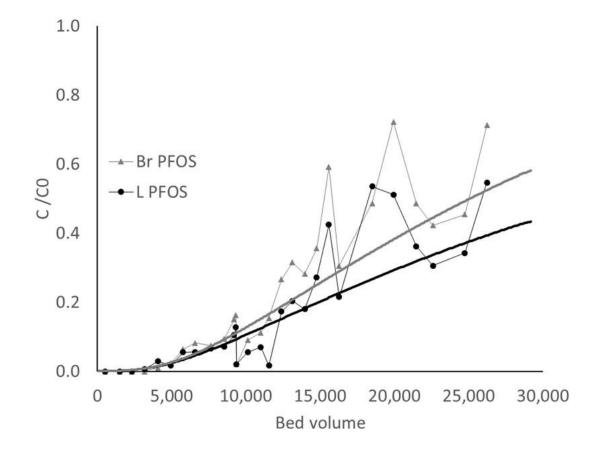
Linear vs Branched Isomers in in the Influent of Pilot-Scale GAC System

		Average Influent Concentration (ng/L)		Branched/Linear (30/70=0.43)	<0.43: Linear isomers are dominant and enriched	
	PFAS	Branched	Linear			
	PFBA	<loq< td=""><td>100 ± 22</td><td><0.43</td><td></td></loq<>	100 ± 22	<0.43		
	PFPeA	19 ± 2.5	350 ± 47	0.054	L- PFCAs were enriched (Br/L<0.43) in	
	PFHxA	<loq< td=""><td>740 ± 70</td><td><0.43</td><td></td></loq<>	740 ± 70	<0.43		
	PFHpA	15 ± 2.0	150 ± 20	0.1	the influent possibly due to FT-precursor	
	PFOA	81 ± 7.2	820 ± 73	0.1	transformation at the site	
	PFNA	<lod< td=""><td>18 ± 5.7</td><td><0.43</td><td></td></lod<>	18 ± 5.7	<0.43		
	PFBS	<lod< td=""><td>17 ± 3.0</td><td><0.43</td><td>Low detections, no branched isomers</td></lod<>	17 ± 3.0	<0.43	Low detections, no branched isomers	
	PFPeS	<lod< td=""><td>30 ± 9.0</td><td><0.43</td><td>detected</td></lod<>	30 ± 9.0	<0.43	detected	
-	PFHxS	190 ± 29	1400 ± 210	0.14		
	PFHpS	29 ± 5.5	62 ± 11.9	0.47		
Linear only,	PFOS	1200 ± 110	1900 ± 170	0.63		
indicators of FT	4:2 FtS	NA	13 ± 3.0		Br-PFSAs and Br-FASAs were enriched	
based foams	6:2 FtS	NA	1000 ± 76	No branched	(Br/L>0.43) possibly due to L-PFSAs	
	8:2 FtS	NA	170 ± 35	isomers	retained in the aquifer materials	
Transformation products of ECF – foams	FPeSA	<loq< td=""><td>125 ± 5.8</td><td><0.43</td><td></td></loq<>	125 ± 5.8	<0.43		
	FHxSA	700 ± 48	2100 ± 140	0.33		
	FOSA	100 ± 9.4	200 ± 19	0.5		
	N-TAmP FHxSA	51 ± 21	97 ± 40	0.52	REMTEC	

#RemTEC



Breakthroughs of Linear and Branched PFOS in the Effluent of Pilot-Scale GAC System

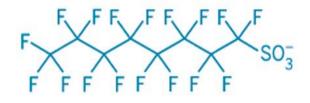


- Average Br-PFOS/L-PFOS = 0.63 in the influent
- The breakthrough curve shows Br/L>1 indicating greater retention of L-PFOS in GAC vessels

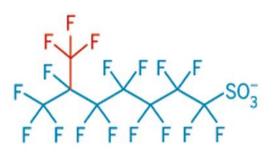


Takeaways

- Data collected from influent monitoring
 - Characterization of influent provides preliminary assessment of PFAS fate and transport
 - Mixed uses of FT and ECF AFFF could be verified using multiple lines of data collection (target and isomer analyses)
 - L- PFCAs were enriched (Br/L<0.43) in the influent suggesting abundant uses and releases of fluorotelomers (L isomers only) that have been converted into linear PFCAs over time
 - L-PFCA precursor transformation may be impacted by the site remediation actions (SVE and air sparging)
 - GW data showed branched PFSA and FASA isomer enrichment (Br/L>0.43) in the influent
 - Linear PFSA and FASA adsorbed in the aquifer materials



PFOS Linear Isomer

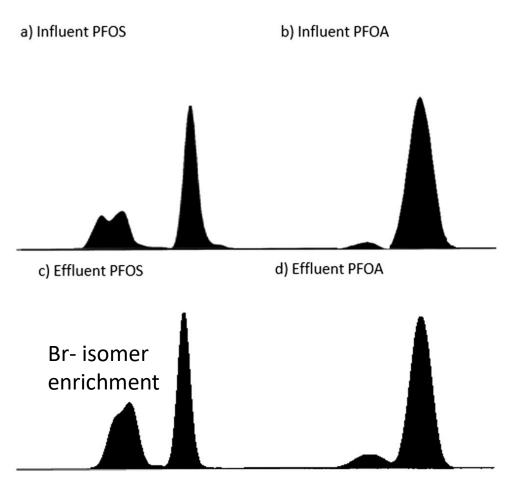


PFOS Branched Isomer (P6MHpS)

#RemTEC

Takeaways

- Precursors do migrate and travel in this case study. The branched and linear isomers do have impact to the understanding of PFAS CSM and treatment effectiveness
- Branched PFSAs would also be more leachable in the aquifer than linear isomers. This is evident by the enriched Br isomers in PFSAs and FASAs in the influent
- Branched isomers with lower Kd and Koc than linear isomers
 - Break through GAC faster than L isomers
 - Br isomer enrichment in the effluent



Br and L isomers for a) PFOS and b) PFOA in influent and c) PFOS and d) PFOA in lead vessel effluent at 11 000 bed volumes.





Summary







Standardized analytical approaches are needed to differentiate and quantify the PFAS isomers for site investigation and treatment

_	

PFAS isomer characterization will allow the system design and operation to cope with Br-PFAS that are more difficult to be removed by GAC



When the stringing federal and state regulatory limits are near zero, enrichment of branched isomers in the influent can increase OM&M costs



More research is needed to assess degradation rates, reaction mechanisms, and competitive sorption of specific isomers not only in the environmentally realistic mixtures, but also in the phase of bioaccumulations and evaluating PFAS treatment options.



Higher resolution of PFAS investigation and establishment of monitoring program to monitor for the changes of PFAS composition overtime will help establish a better CSM and ultimately optimize the selection of PFAS treatment technologies for site cleanup.



Acknowledgement





University of Georgia Dr. Jack Huang and his research team Oregon State University Dr. Jennifer Field and Dr. Alix E. Rodowa







Thank You

Dora Chiang, Ph.D. Jacobs Atlanta, GA <u>Dora.Chiang@jacobs.com</u> (404)4051214



