

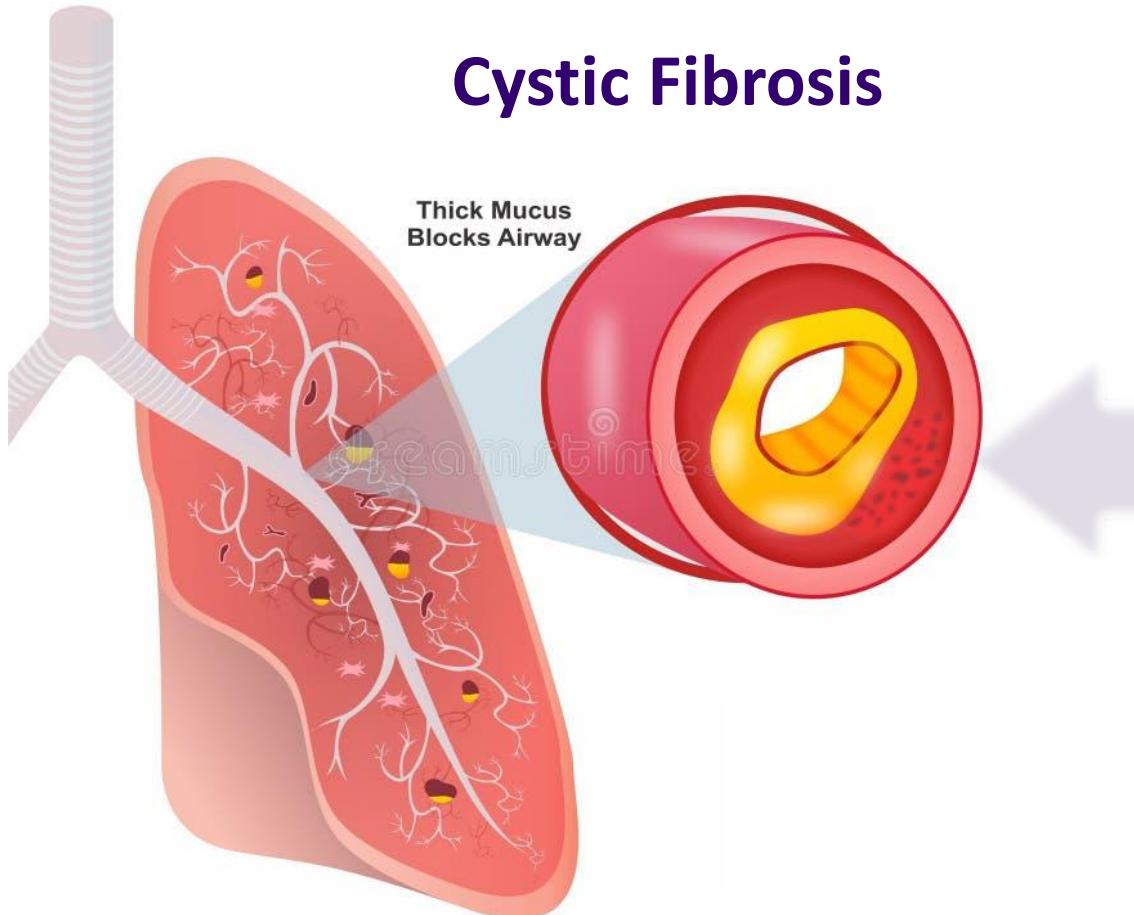
Tile-based Analysis of GC \times GC-TOFMS Data of SPME Sampled VOCs Produced from *Pseudomonas aeruginosa* and *Aspergillus fumigatus*

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17th Multidimensional Chromatography Workshop

Project Motivation



Cystic Fibrosis

- *Pseudomonas aeruginosa* and *Aspergillus fumigatus* are major pathogens found in the lungs of patients with Cystic fibrosis.
- Their coexistence worsens lung function and leads to poor clinical outcomes.
- The project aims to investigate their metabolic interactions.

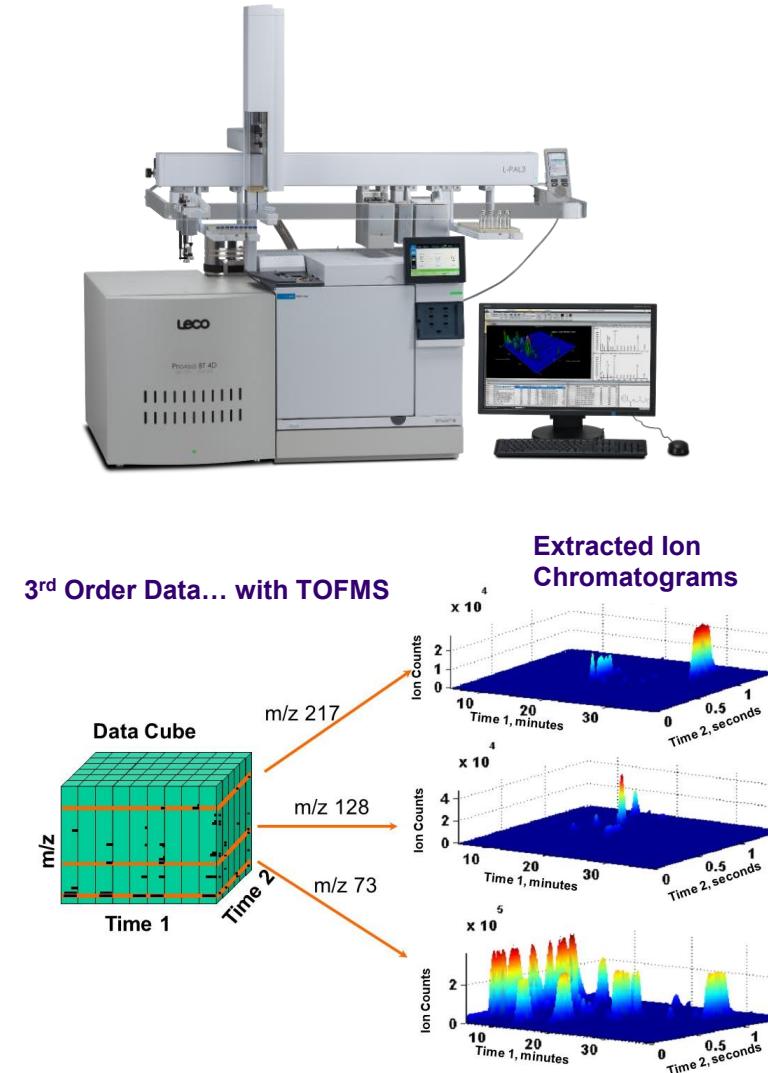
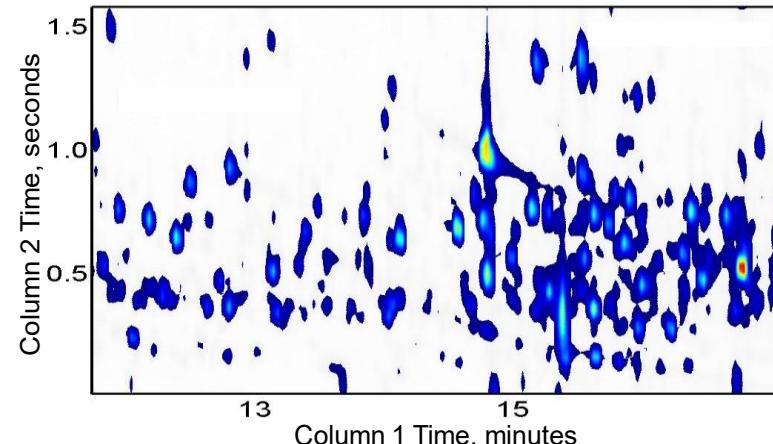
Aim and Methodology

- To discover analytes that are **statistically different** in the headspace across all sample classes (Media, PA monoculture, AF monoculture, and their coculture) using comprehensive two-dimensional gas chromatography with time-of-flight mass spectrometry (GC \times GC-TOFMS) and chemometrics.
- To explore appropriate experimental design approaches for generating high quality data and enabling high throughput experimentation.

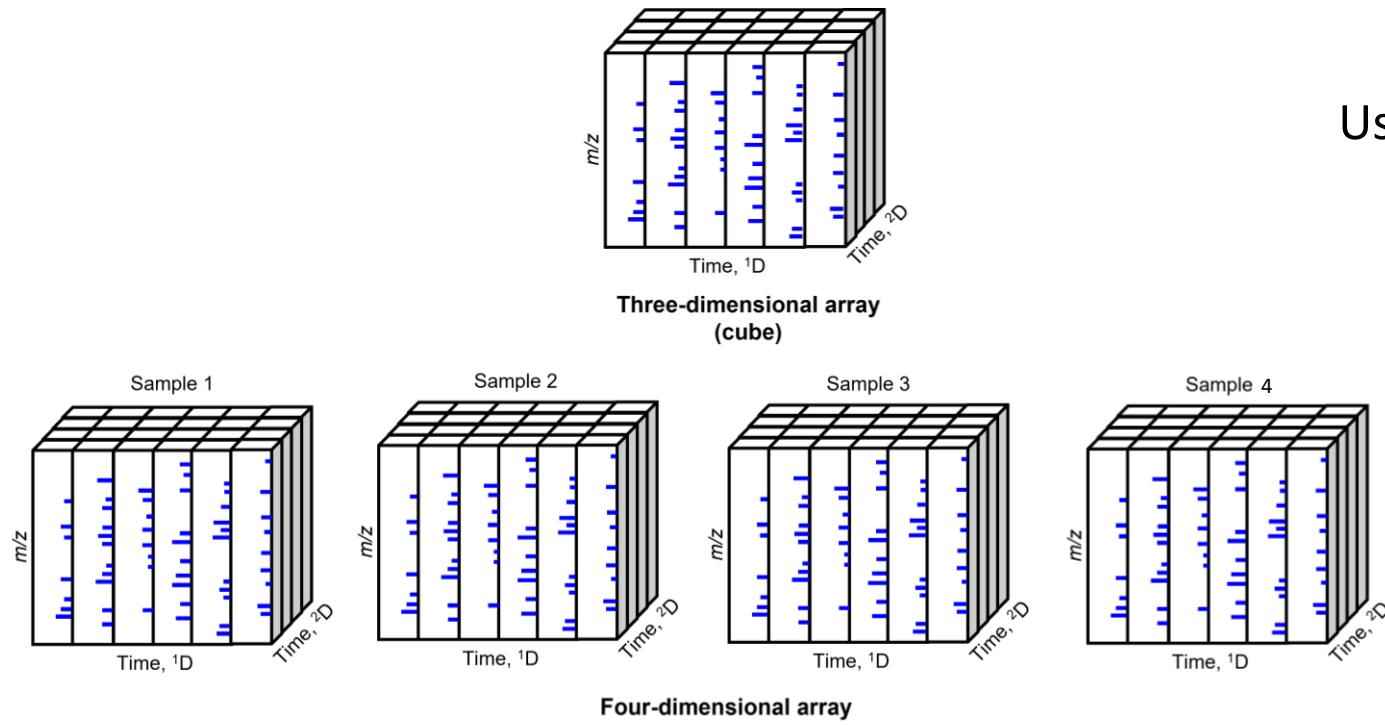


GC \times GC-TOFMS : Comprehensive Two-Dimensional Gas Chromatography with Time-of-Flight Mass Spectral Detection

- Mass spectra with electron impact ionization and unit mass resolution
→ peak identification
- Fast → 500 spectra / second
Column Two peak widths:
~ 50 ms to ~ 300 ms
- Adds another selective dimension
→ 3rd - order data
→ Provides informative data analysis



Methods of Handling GC×GC-TOFMS Data



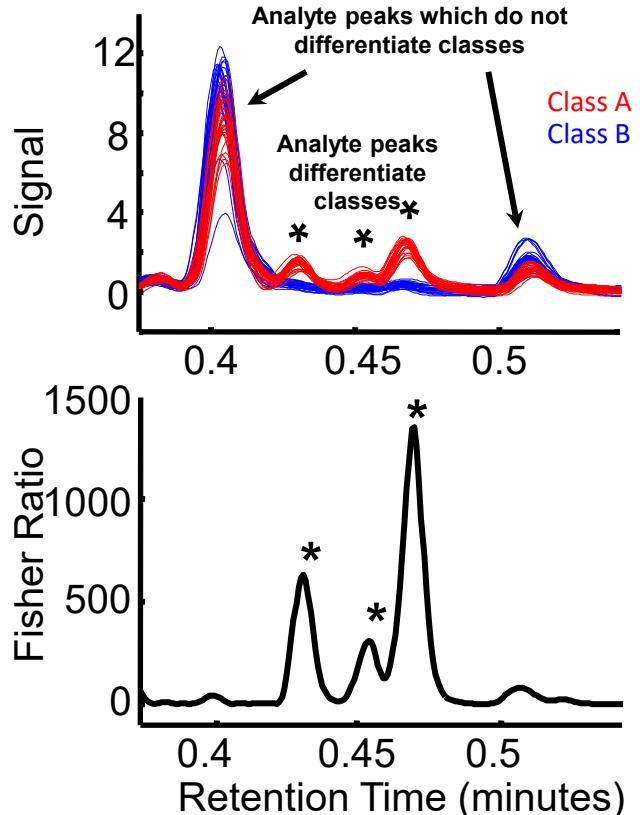
Use of a TOFMS with a high collection frequency can create a large amount of data
> 10^6 data points per chromatogram!

Data reduction strategies

- **Assemble peak tables**
- **Average the chromatographic signal (i.e., binning)**
- **Discovery-based Feature selection**

Tile-based Fisher Ratio (F-ratio) Analysis

Shown for 1D data....



.... applied to **Class A** samples
versus **Class B** samples

$$\text{Standard } F - \text{ratio} = \frac{\text{Between Class Variance}}{\sum(\text{Within Class Variance})}$$

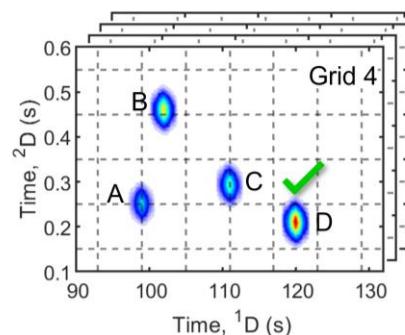
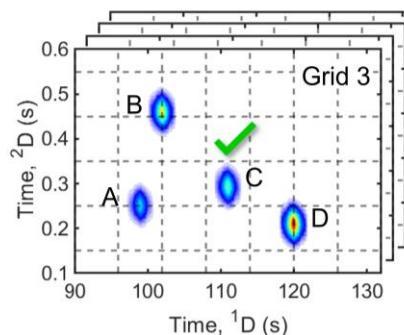
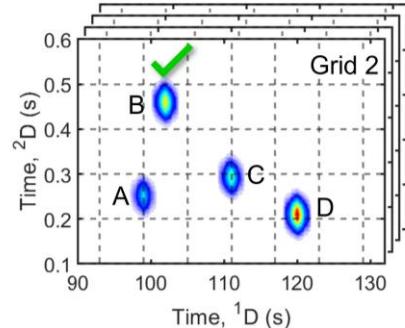
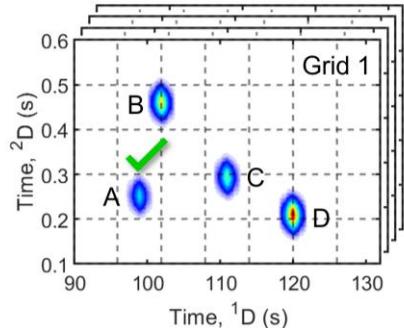
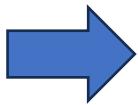
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F-ratio Analysis provides a ranked hitlist of analytes that are likely to be statistically different in concentration (p -value < 0.05) between sample classes.

<h1>Hit List</h1>		
Hit #	Fisher Ratio	Analyte
1	High	AAA

Tile-based Fisher Ratio (F-ratio) Analysis

Illustration of how 4-grid tile-based FRA works.

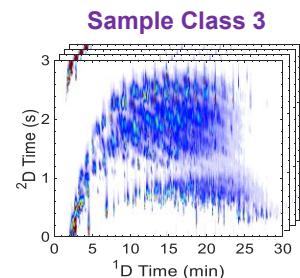
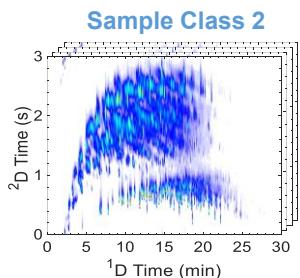
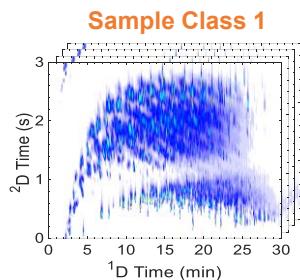
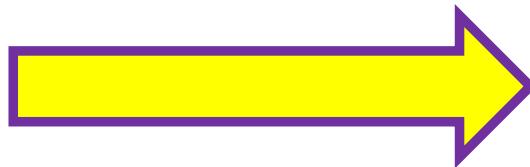


Hit List

Hit #	F-ratio	$({}^1t_R, {}^2t_R)$
1	High	(x, y)
N	Low	(x_N, y_N)

Software Development at UW

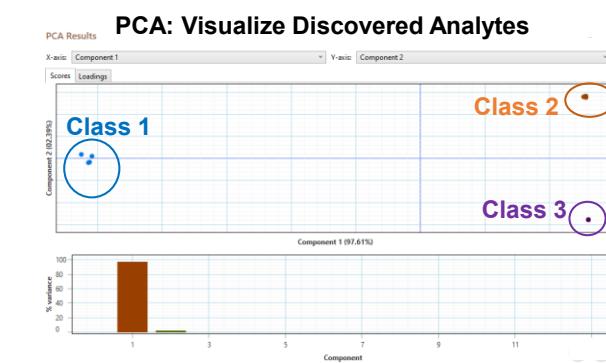
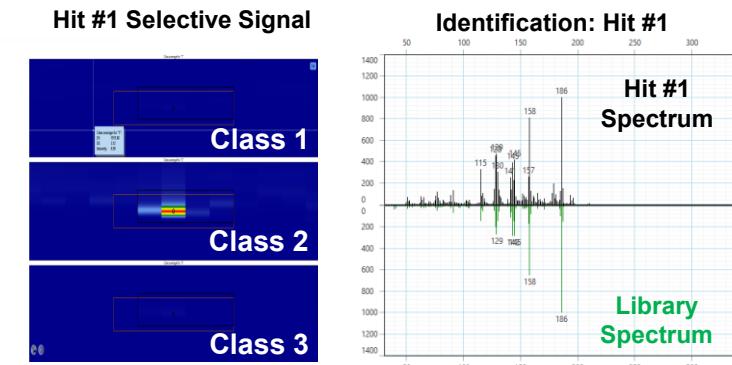
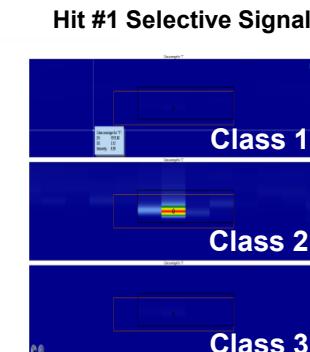
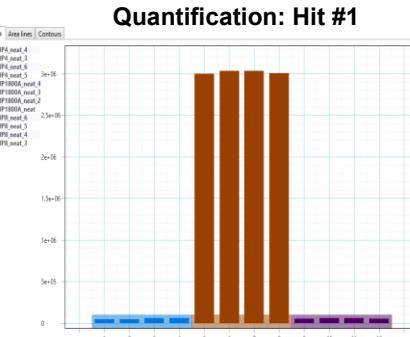
Tile-based
Fisher Ratio Analysis



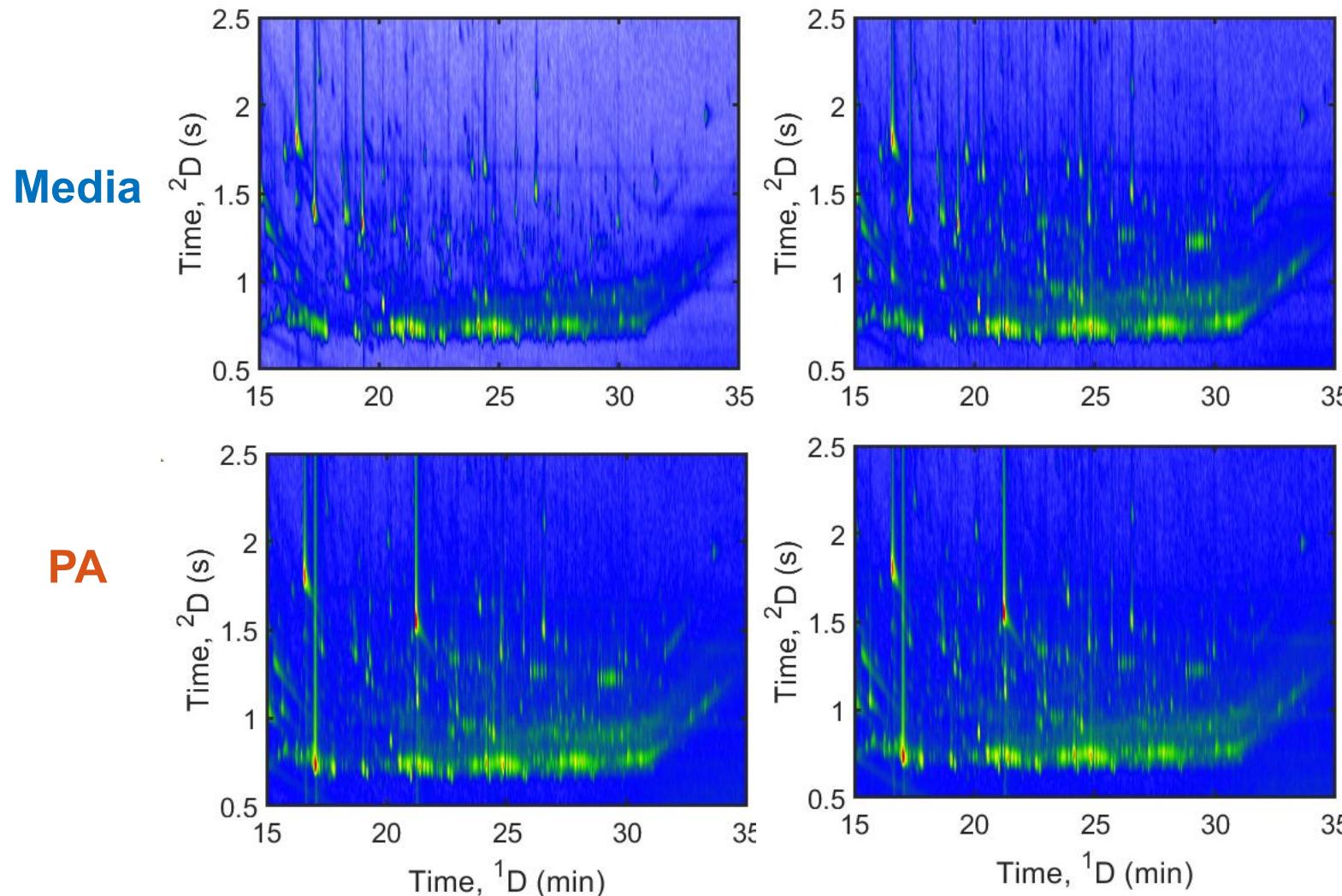
Apply
ChromaTOF Tile

Hit Table

#	Aug. ratio	Top mass	RT1	RT2	Masses	1	2	3
001	21025.54	186	1938.08	2.32	160	100	100	100
002	14000.00	186	1938.08	2.32	160	100	100	100
003	134310.79	209	1963.07	2.41	143	100	100	100
004	132793.51	46	1744.09	2.12	78	100	100	100
005	132793.51	46	1744.09	2.12	78	100	100	100
006	124204.93	163	1963.07	2.36	167	100	100	100
007	48721.71	145	1938.08	2.42	147	100	100	100
008	442616.67	81	1779.08	0.73	40	100	100	100
009	38975.00	245	1938.08	2.32	160	100	100	100
010	38975.00	245	1938.08	2.32	160	100	100	100
011	140631.52	214	2038.07	2.37	133	100	100	100
012	140631.52	240	1963.07	2.38	160	100	100	100
013	140631.52	240	1963.07	2.38	160	100	100	100
014	132793.51	58	147.08	1.81	52	100	100	100
015	132793.51	58	147.08	1.81	52	100	100	100
016	132793.51	58	147.08	1.81	52	100	100	100
017	38975.00	146	1938.08	2.35	173	100	100	100
018	280461.52	49	1455.91	1.60	173	100	100	100
019	280461.52	49	1455.91	1.60	173	100	100	100
020	280461.52	49	1455.91	1.60	173	100	100	100
021	280461.52	186	1938.08	2.37	152	100	100	100
022	27303.00	208	1981.08	2.30	152	100	100	100
023	27303.00	208	1981.08	2.30	152	100	100	100
024	26671.52	209	1963.07	2.34	152	100	100	100
025	25052.02	261	1767.09	1.09	109	100	100	100
026	25052.02	261	1767.09	1.09	109	100	100	100
027	25052.02	261	1767.09	1.09	109	100	100	100
028	25350.00	262	1973.08	1.54	128	100	100	100
029	25350.00	262	1973.08	1.54	128	100	100	100
030	25350.00	262	1973.08	1.54	128	100	100	100
031	251343.51	208	1932.08	2.39	160	100	100	100
032	251343.51	208	1932.08	2.39	160	100	100	100
033	25796.00	58	279.08	2.07	78	100	100	100
034	25796.00	58	279.08	2.07	78	100	100	100
035	22031.74	68	133.98	1.85	68	100	100	100
036	22031.74	68	133.98	1.85	68	100	100	100
037	21993.52	49	144.99	2.06	49	100	100	100
038	21993.52	49	144.99	2.06	49	100	100	100
039	21775.96	75	1553.91	1.40	173	100	100	100
040	21775.96	172	1980.07	2.51	177	100	100	100



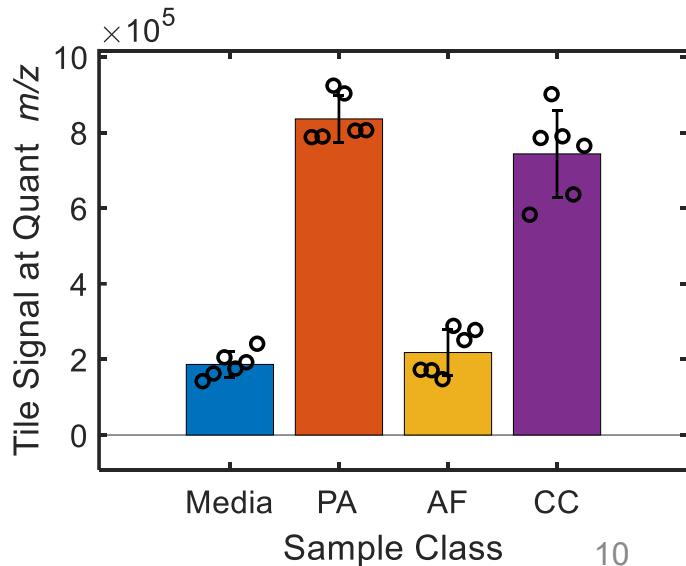
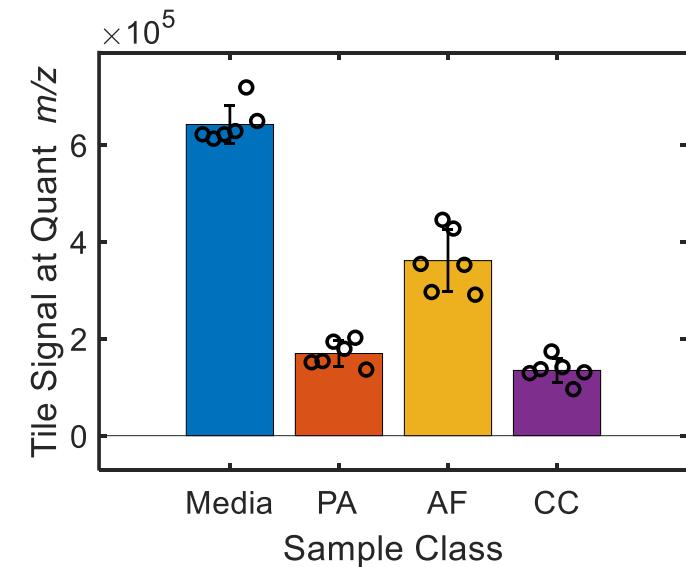
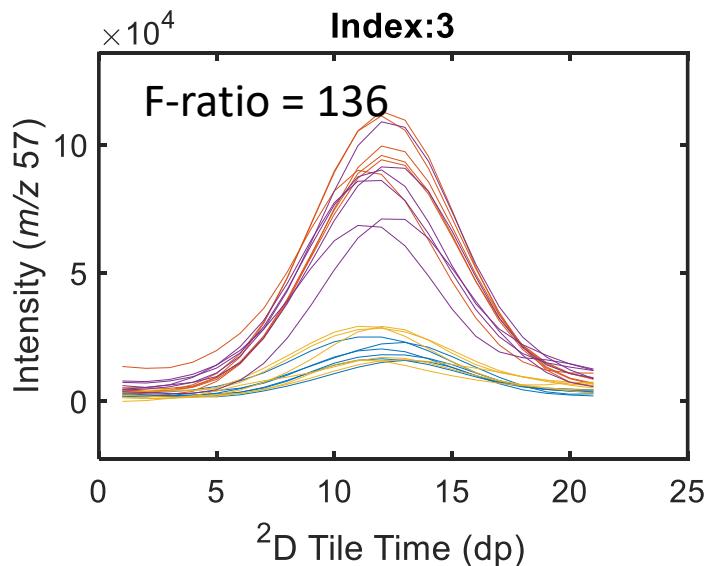
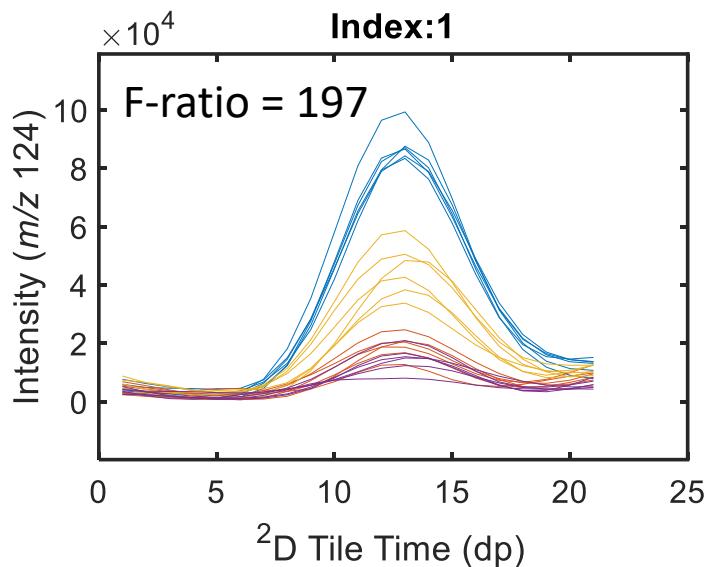
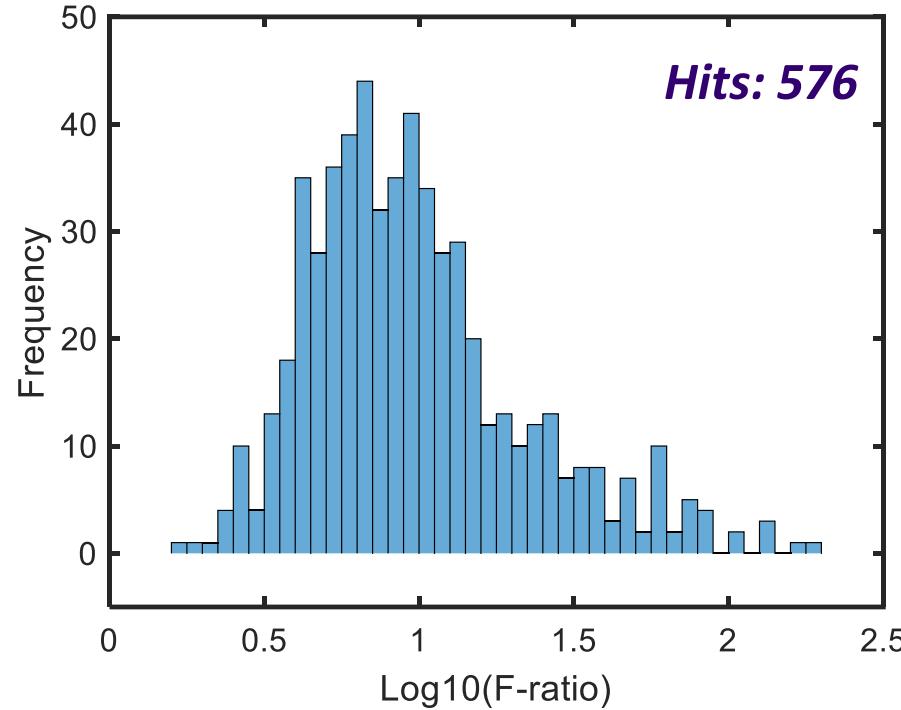
Comparative Analysis of GC \times GC-TOFMS Data



~ 500 peaks detected in the TIC chromatogram

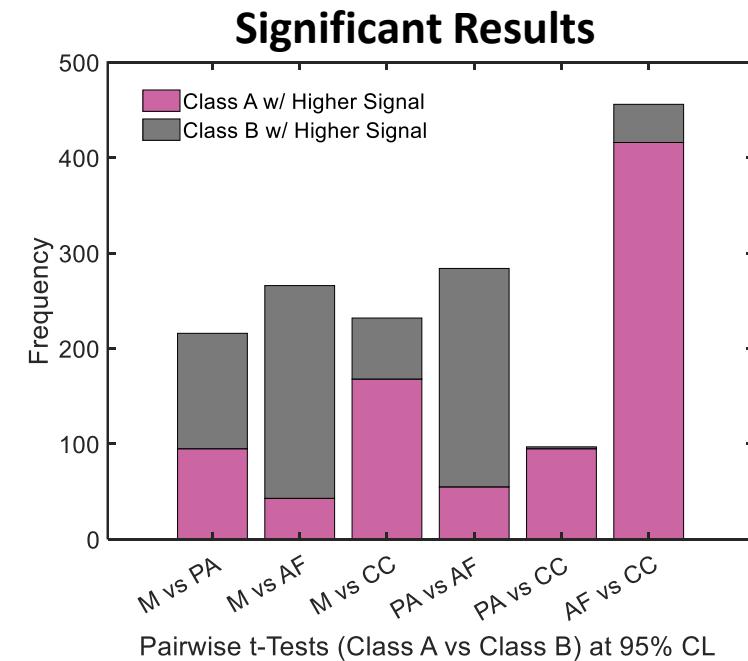
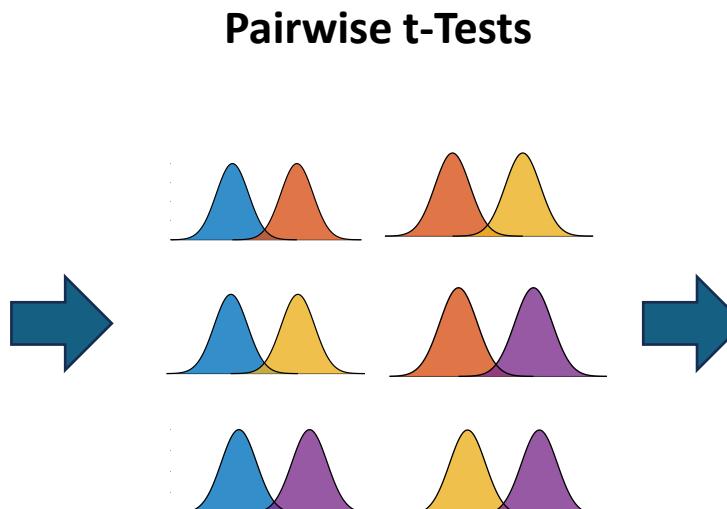
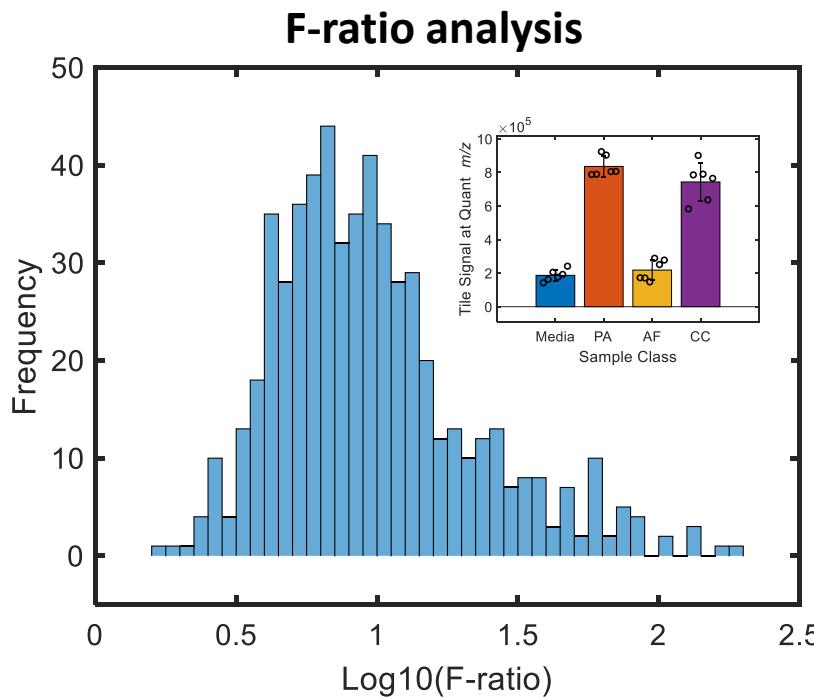
The media sample appears to have less overall signal

Tile-based Fisher Ratio (F-ratio) Analysis



Followed Up: Pairwise t-Test (95% CL)

- F-ratio analysis discovers analytes that are statistically different across all sample classes.
- Pairwise t-tests were formed to find out which classes differ.
- Significance set at $p < 0.05$ (95% confidence level).
- For 4 classes, this results in 6 pairwise comparisons.

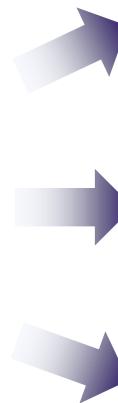


Aim and Methodology

- To discover analytes that are **statistically different** in the headspace across all sample classes using comprehensive two-dimensional gas chromatography with time-of-flight mass spectrometry (GC \times GC-TOFMS) and chemometrics
- To explore appropriate experimental design approaches for generating high quality data and enabling high throughput experimentation

Feature Selection Methods

Tile-based Analysis



Fisher Ratio

Coefficient of Variation

Fold Change

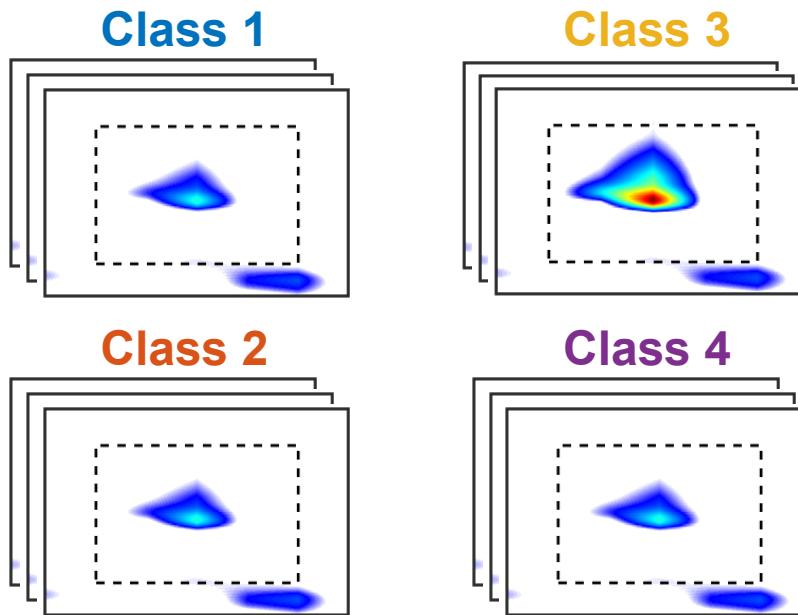
Experimental Design Considerations Under Practical Constraints

- Replicates may not always be available due to sample, time, and/or cost limitations
- Replicates may not be reproducible: consistency across replicates is required
- Highly precision required during sample preparation for device fabrication
- Time sensitivity of sample preparation and data collection
 - SPME-GC \times GC-TOFMS dataset collection required 24-hour spacing between the completion of sample preparation and the start of volatile absorption.

*Experimental Design Approaches for High-Quality Data
via High Throughput Experimentation*

Tile-Based Feature Selection Metrics

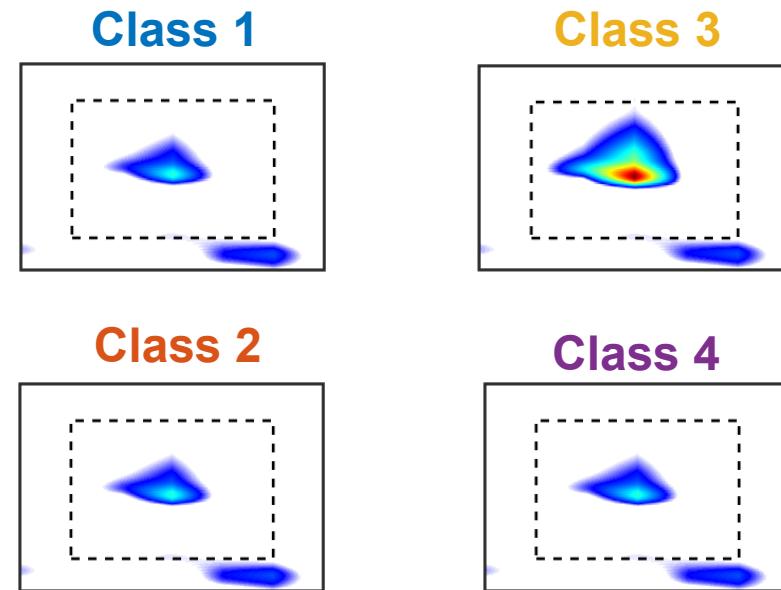
Tile-Based F-ratio Analysis



$$F - ratio = \frac{Between\ Class\ Variance}{\sum(Within\ Class\ Variance)}$$

Replicates may not always be available due to sample, time, and/or expense limitations

Tile-Based Coefficient of Variation Analysis



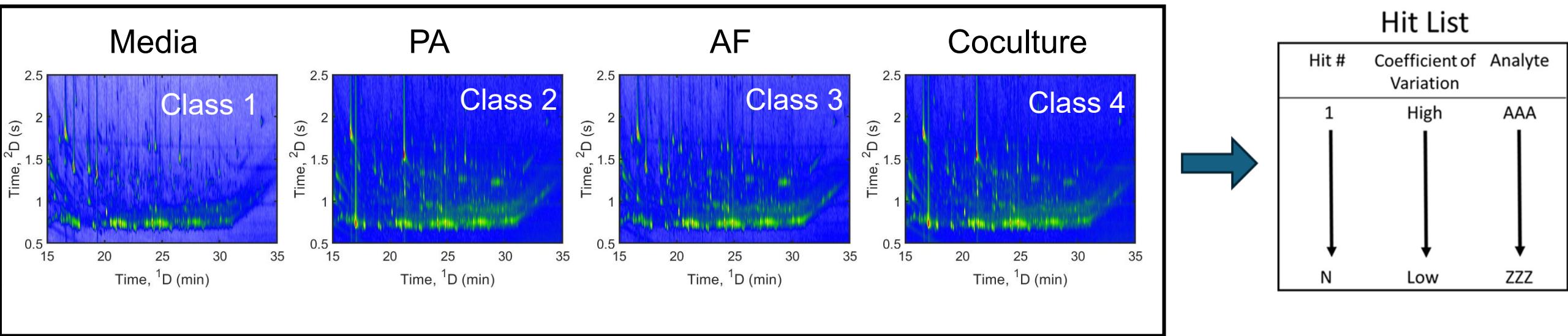
$$CV = \frac{STD}{Mean}$$

Overcomes issues associated with pixel-based subtraction plots

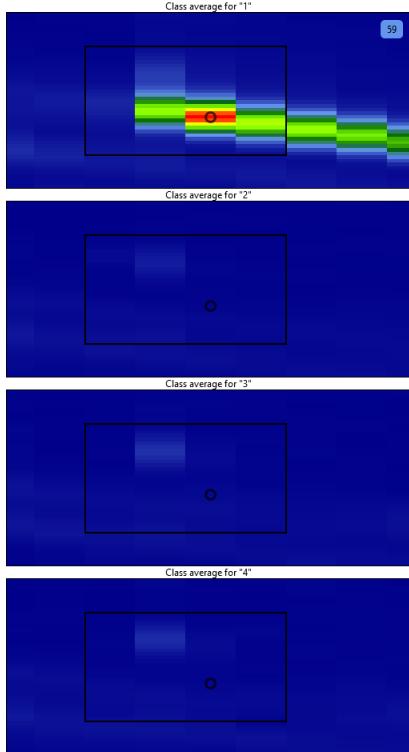
Tile-Based Coefficient of Variation Analysis

We applied ChromaTOF Tile (Coefficient of Variation method) to simultaneously to obtain a single hitlist that relates all analytes in all samples in terms of up and down regulations of their concentrations.

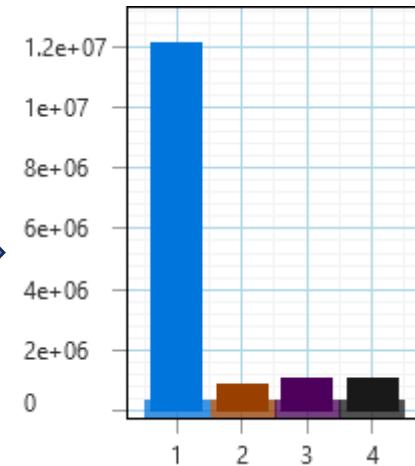
This required only one replicate from each sample to obtain this comprehensive hitlist of relative concentration for each analyte.



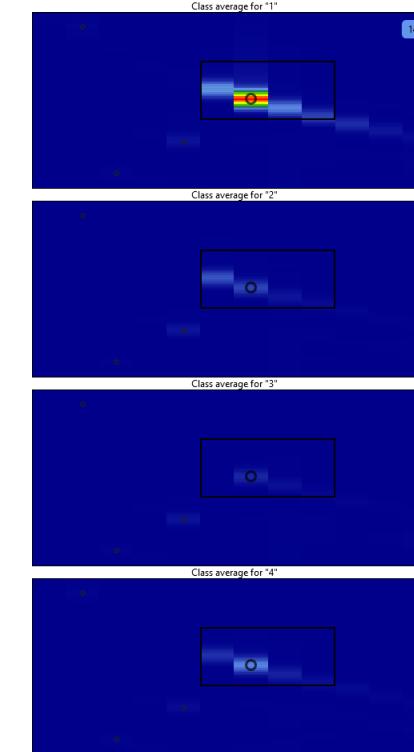
Example Hits with Highest Signals in Class 1 – Media



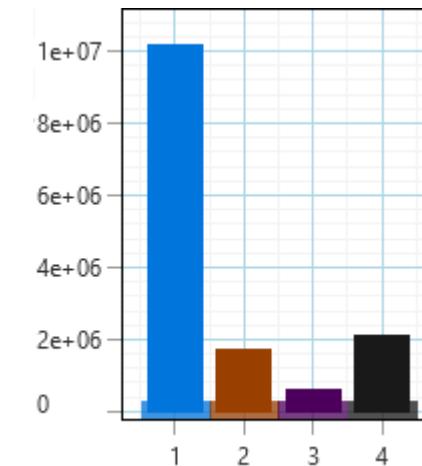
'2-Cyclobutyl-2-propanol'
MV=741 , CV = 1.48



20250708_Set1_A_Media
20250708_Set1_B_MonoPA01
20250708_Set1_C_MonoAF1607
20250708_Set1_D_Coculture

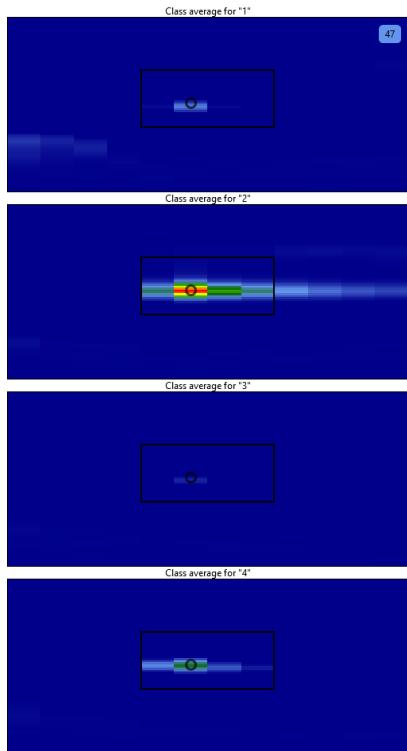


'4-Piperidinone, 2,2,6,6-tetramethyl-'
MV = 821 , CV = 1.2

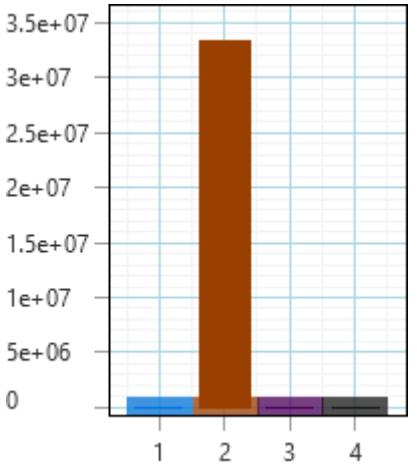


20250708_Set1_A_Media
20250708_Set1_B_MonoPA01
20250708_Set1_C_MonoAF1607
20250708_Set1_D_Coculture

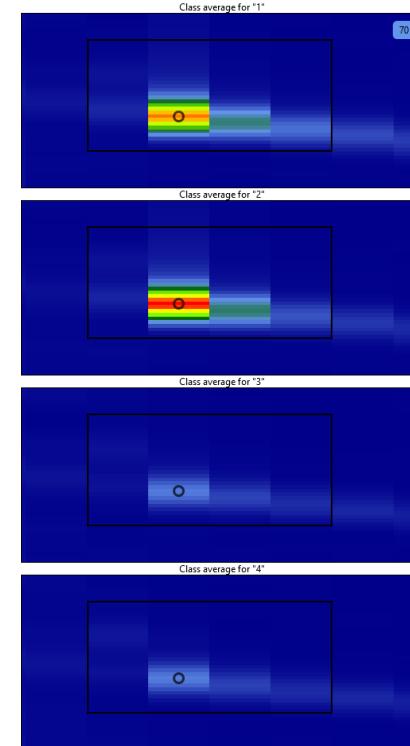
Example Hits with Highest Signals in Class 2 – PA monoculture



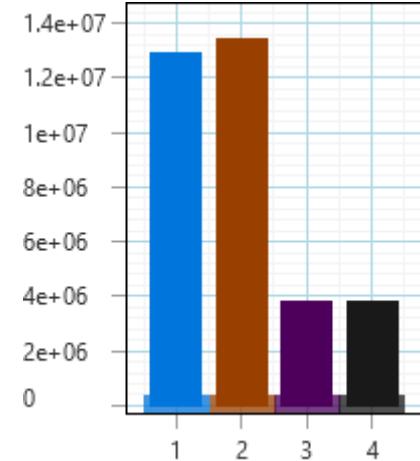
'Pentaborane(11)'
MV = 733 , CV = 2



20250708_Set1_A_Media
20250708_Set1_B_MonoPA01
20250708_Set1_C_MonoAF1607
20250708_Set1_D_Coculture

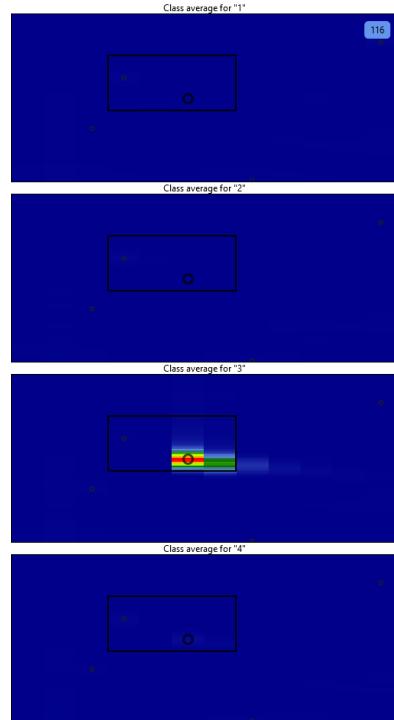


'Heptanal'
MV = 881 , CV = 0.64



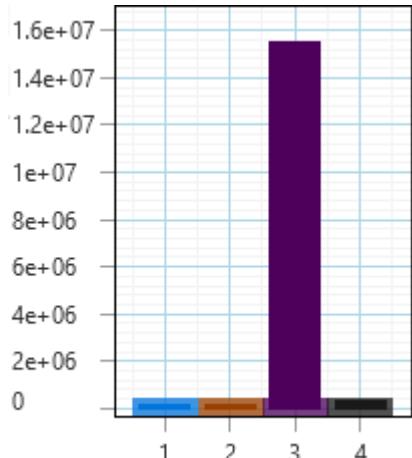
20250708_Set1_A_Media
20250708_Set1_B_MonoPA01
20250708_Set1_C_MonoAF1607
20250708_Set1_D_Coculture

Example Hits with Highest Signals in Class 3 – AF monoculture

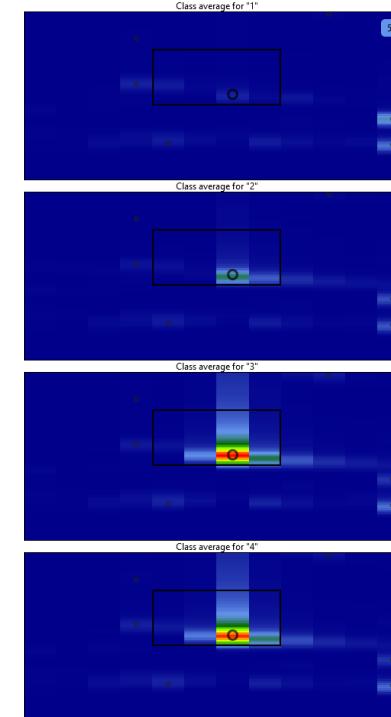


'3(2H)-Thiophenone, dihydro-2-methyl-'

MV = 818, CV = 1.9

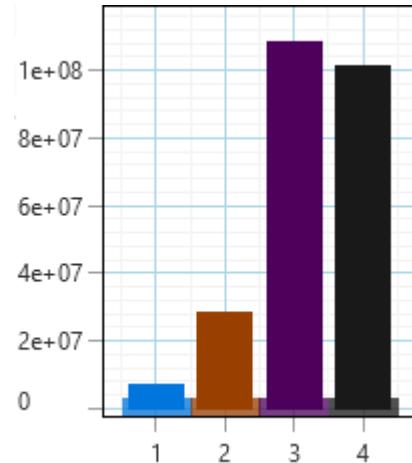


Legend:
20250708_Set1_A_Media
20250708_Set1_B_MonoPA01
20250708_Set1_C_MonoAF1607
20250708_Set1_D_Coculture



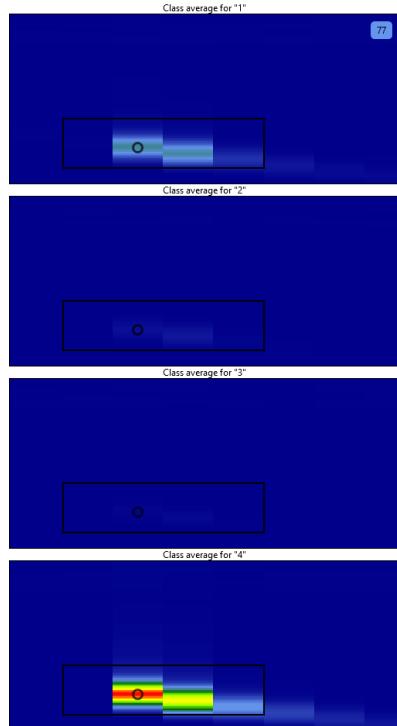
'1-Octen-3-ol'

MV = 837, CV = 0.83

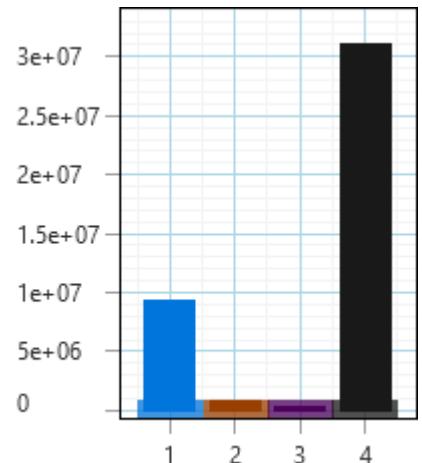


Legend:
20250708_Set1_A_Media
20250708_Set1_B_MonoPA01
20250708_Set1_C_MonoAF1607
20250708_Set1_D_Coculture

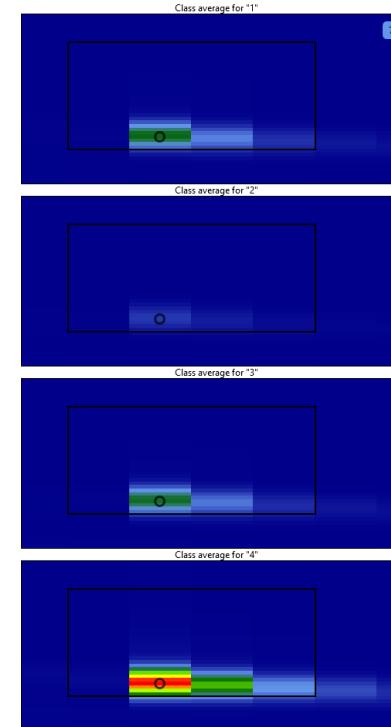
Example Hits with Highest Signals in Class 4 - Coculture



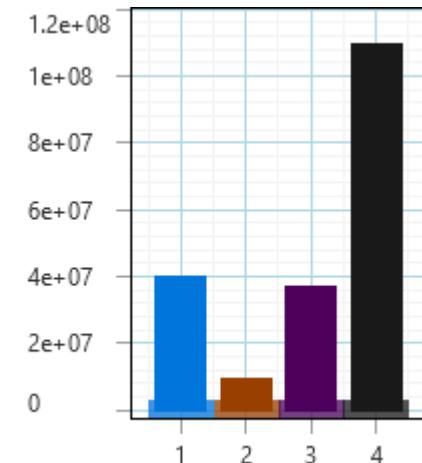
'1H,1H,2H-Perfluoro-1-octene'
MV = 798, CV = 1.39



20250708_Set1_A_Media
20250708_Set1_B_MonoPA01
20250708_Set1_C_MonoAF1607
20250708_Set1_D_Coculture



'2,4-Dimethyl-1-heptene'
MV = 915, CV = 0.87



20250708_Set1_A_Media
20250708_Set1_B_MonoPA01
20250708_Set1_C_MonoAF1607
20250708_Set1_D_Coculture

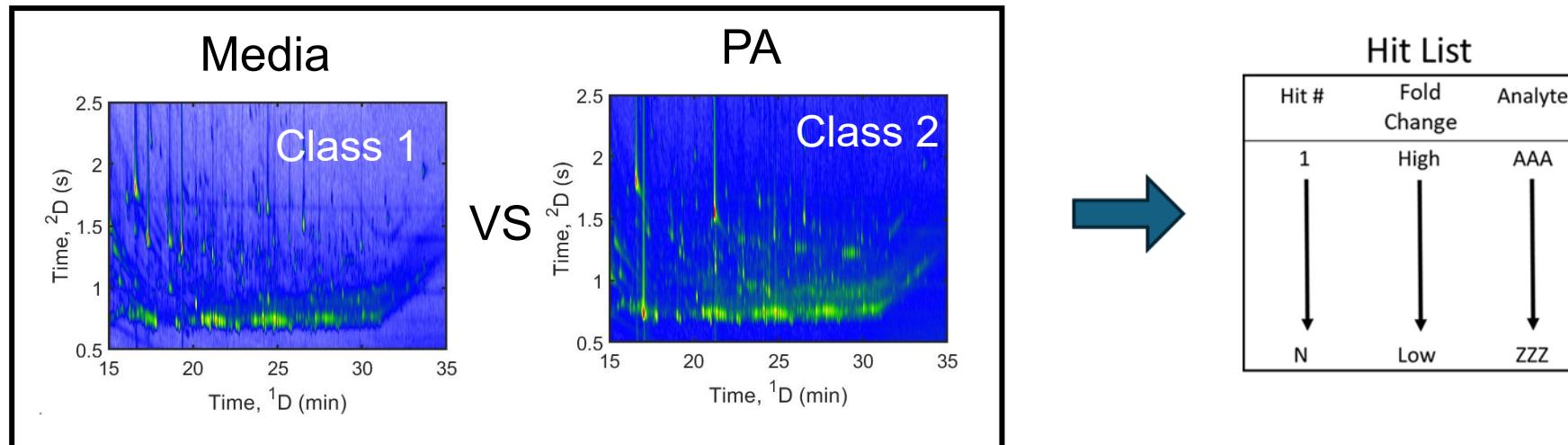
Tile-based Pairwise Analysis

Non-Targeted Discovery Analysis Comparing Two Chromatograms

Utilize a tile-based relative change method that discovers analytes that are different in concentration between two samples with only one replicate.

$$\text{Fold Change} = \left(\frac{\text{Class 1 Tile Signal Sum}}{\text{Class 2 Tile Signal Sum}} \right)$$

Discover any hits with specified fold change (FC) threshold.



Conclusions and Next Steps

- Tile-based F-ratio analysis discovered 576 sample class distinguishing analytes.
- Additional pairwise t-tests can further determine which sample classes differ.
- Under constrained experimental conditions, the tile-based coefficient of variation (CV) method is more suitable for achieving high-quality data in high-throughput experimentation.
- Appropriate feature selection methods (F-ratio, CV, and FC) should be selected based on the specific experimental design.
- Further analyte identification can be performed using the chemometric tool PARAFAC.

Acknowledgements

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Synovec Lab Members

- Rachel Halvorsen
- Austin Dobrecevich
- Wenjing Ma
- Haylee Meissner
- Cassandra Padilla
- Peri Abdigali
- Valencia Parker
- Jungho Ahn
- Stephanie Nguyen

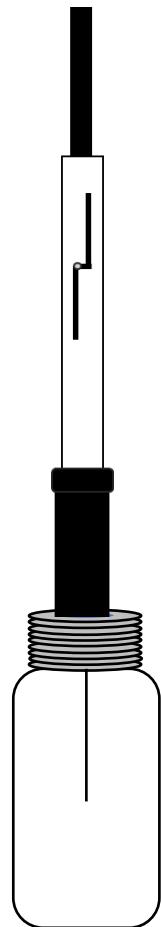
Synovec Lab

Gas Chromatography, Liquid Chromatography,
and Mass Spectrometry, with Multi-Dimensional
Data Analysis



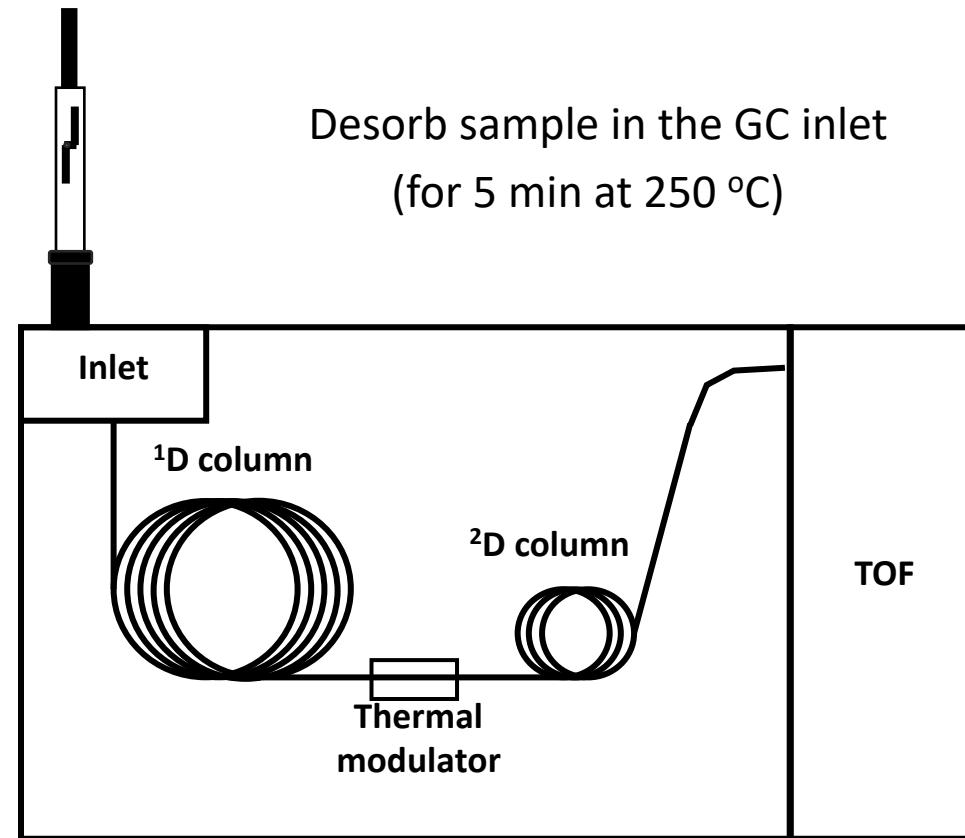
Thank you for listening! Any questions?

SPME Tool



Expose SPME to sample
(for 30 min at 37 °C)

Desorb sample in the GC inlet
(for 5 min at 250 °C)



Condition Workflow



Blank Run with Only Fiber



Run Sample Workflow

