

# Detection, quantification, and isomer differentiation of per- and polyfluoroalkyl substances (PFAS) using MALDI-TOF with trapped ion mobility

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## Abstract

Per- and polyfluoroalkyl substances (PFAS) are a class of organic compounds that have attracted global attention for their persistence in the environment, exposure to biological organisms, and their adverse health effects. There is an urgent need to develop analytical methodologies for characterization of PFAS in various sample matrices. Matrix-assisted laser desorption/ionization mass spectrometry (MALDI-MS) represents a chromatography-free MS method that performs laser-based ionization and *in situ* analysis on samples. In the present study, we present PFAS analysis by MALDI-time-of-flight MS with trapped ion mobility (TIMS), which provides an additional dimension of gas phase separation based on the size-to-charge ratios. MALDI matrix composition and key instrument parameters were optimized to produce different ranges of calibration curves. Parts per billion range of calibration curves were achieved for a list of legacy and alternative perfluorosulfonic acids (PFSAs) and perfluorocarboxylic acids (PFCAs), while ion mobility spectrum filtering enabled parts per trillion (ppt) range of calibration curves for PFSAs. We also successfully demonstrated the separation of three perfluorooctanesulfonic acid (PFOS) structural isomers in the gas phase using TIMS. Our results demonstrated the new development of utilizing MALDI-TOF-MS coupled with TIMS for fast, quantitative, and sensitive analysis of PFAS, paving ways to future high-throughput and *in situ* analysis of PFAS such as MS imaging applications.

**Running title:** PFAS analysis with MALDI-TIMS-TOF MS.

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## 1 1. Introduction

2 Per- and polyfluoroalkyl substances (PFAS) are a class of persistent environmental contaminants  
3 that have been in use since the mid-20<sup>th</sup> century.<sup>1,2</sup> Due to their high stability and amphiphilic  
4 properties, PFAS have been used in various industrial processes and consumer products such  
5 as non-stick cookware, firefighting foams, and household cleaning products.<sup>3</sup> Over the past  
6 several decades, per- and polyfluorinated compounds have evolved and diversified, resulting in  
7 thousands of different chemical structures and functional differences.<sup>4</sup> PFAS have shown  
8 resistance to biological and chemical degradation due to their physicochemical properties<sup>5</sup>, which  
9 has led to accumulation of these compounds in the environment. Their persistence and  
10 accumulation in the environment often result in biomagnification within the food chain as you  
11 ascend trophic levels.<sup>6-10</sup> Research has shown that PFAS cause adverse health effects such as  
12 fertility issues, thyroid disease, and endocrine function.<sup>1,11</sup> There is a significant concern regarding  
13 the fate of PFAS in the environment and the long-term adverse effects of PFAS exposure in  
14 humans and other organisms, calling for an urgent need for the development of analytical  
15 methodologies for characterizing PFAS in a wide variety of samples.

16  
17 Many analytical advancements have been made to address the challenge of PFAS analysis.<sup>12,13</sup>  
18 One particularly important tool is mass spectrometry (MS). Current efforts in MS-based PFAS  
19 analysis focus on utilizing liquid or gas chromatography (LC/GC)-MS for qualitative and  
20 quantitative characterization of PFAS in complex matrices.<sup>14</sup> While chromatography-coupled MS  
21 methodologies enable sensitive detection and comprehensive structural elucidation, they usually  
22 require extensive sample preparation, cost longer time in analysis, and may be prone to  
23 contaminations during sample processing.<sup>15</sup> Furthermore, LC/GC-MS analysis requires the  
24 samples to be homogenized and causes the loss of spatial information. Matrix-assisted laser  
25 desorption/ionization (MALDI) is a laser-based “soft ionization” technique for chromatography-  
26 free, in situ MS analysis. Coupled with a time-of-flight (TOF) mass analyzer, MALDI-TOF-MS has  
27 been widely used for high-throughput and imaging MS applications<sup>16-18</sup>. Some studies have  
28 reported the application of MALDI-MS to profile the spatial distribution of PFAS in zebrafish and  
29 mouse tissues,<sup>19-22</sup> and detect PFAS in environmental water samples<sup>23</sup>, showcasing MALDI-MS  
30 as a promising tool for fast, in situ PFAS analysis.

31  
32 The structural diversity of PFAS and other xenobiotics, together with the complexity of biological  
33 and environmental matrices where PFAS are present, calls for the capability to identify PFAS  
34 signals from biological molecules and backgrounds. Chromatography, high-resolution and  
35 tandem MS, and Kendrick Mass Defect (KMD) analysis are among tools that are widely used.<sup>14,24</sup>  
36 In recent years, ion mobility spectrometry (IMS) has shown unique strength in the analysis of  
37 halogenated xenobiotics, including PFAS, due to its capability in separating ions by size, shape  
38 and charge state in the gas phase.<sup>25-27</sup> By coupling drift-time ion mobility spectrometry (DTIMS)  
39 to LC-MS, Foster, et al. demonstrated a unique trendline of collision cross section vs. *m/z* for  
40 halogenated xenobiotics compared to biological molecules, facilitating the identification of  
41 xenobiotics in a complex sample matrix.<sup>28</sup> Furthermore, DTIMS is capable of differentiating PFAS  
42 isobars and isomers.<sup>15</sup> Trapped ion mobility spectrometry (TIMS) is one of the most recent  
43 iterations of IMS and facilitates ion separation by using a carrier gas to pass ions through a  
44 mobility separation region against an opposing electric field gradient. Ions are eluted after

45 achieving separation by decreasing the opposing electric field; by tuning the ramping time, TIMS  
46 can achieve resolution of ~300.<sup>29</sup> Recently, the development of instrumentation has enabled a  
47 MALDI-TIMS-TOF configuration,<sup>30</sup> providing opportunities for utilizing the power of IMS with  
48 MALDI-MS for PFAS analysis.

49  
50 Herein, we establish a MALDI-TIMS-TOF method for quantitative analysis of several legacy and  
51 alternative PFAS molecules. Matrix screening and optimization was performed for negative  
52 ionization mode analysis for selected legacy and alternative perfluorocarboxylic acids (PFCAs)  
53 and perfluorosulfonic acids (PFSAs), which demonstrated that 1,5-diaminonaphthalene was best  
54 suited for PFAS analysis. Several key instrument parameters were tuned and optimized to  
55 maximize detection and sensitivity of PFAS analysis by MALDI-TIMS-TOF with detection and  
56 quantitation for PFCAs and PFSAs at ppb and ppt concentrations, respectively. Mobility filtering  
57 by TIMS improved S/N by reducing background signal in select PFSAs examined at ppt  
58 concentrations. Furthermore, we've firstly reported differentiation of PFAS structural isomers  
59 using TIMS. Our results demonstrated the new development of utilizing MALDI-TOF-MS coupled  
60 with TIMS for fast, quantitative, and sensitive analysis of PFAS, paving ways to future high-  
61 throughput and *in situ* analysis of PFAS such as MS imaging applications.

62

## 63 **2. Materials and Methods**

64

### 65 2.1 Materials

66 Perfluorohexanoic acid (PFHxA, >98%), perfluorodecanoic Acid (PFDA, 98%), perfluoro-3,7-  
67 dimethyloctanoic acid (PF-3,7-DMOA, 95%), perfluorobutanesulfonic acid (PFBS, 97%), and  
68 ammonium 2,3,3,3-tetrafluoro-2-(heptafluoropropoxy)propanoate (GenX, 95%), 2',5'-  
69 dihydroxyacetophenone (2',5'-DHAP, 97%), 2',6'-dihydroxyacetophenone (2',6'-DHAP, 99%), 9-  
70 aminoacridine (9-AA, >99.5%), norharmine (NRM, 98%) were purchased from Sigma-Aldrich  
71 Chemical (St. Louis, MO, USA). Sodium perfluoro-3-methylheptanesulfonate  
72 (Na3PMHpS, >98%), sodium perfluoro-6-methylheptanesulfonate (Na6PMHpS, >98%), and  
73 potassium 9-chlorohexadecafluoro-3-oxanonane-1-sulfonate (F53B (Major), >98%) were  
74 purchased from Wellington Laboratories (Guelph, Ontario, Canada). Perfluorooctanoic acid  
75 (PFOA, 96%), 1',8'-Bis(dimethylamino)naphthalene (DMAN), 1',5'-diaminonaphthalene  
76 (DAN, >98%), LC-MS grade acetonitrile, LC-MS grade water, and LC-MS grade methanol were  
77 purchased from Fisher Scientific (Pittsburgh, PA, USA). Perfluorohexanesulfonic acid  
78 (PFHxS, >95%), and perfluorooctanesulfonic acid (PFOS, >97%) were purchased from Synquest  
79 Labs (Alachua, FL, USA).

80

### 81 2.2 Sample Preparation

82 Chemical matrices were prepared to 1mL total volumes, and each matrix was vortexed prior to  
83 aliquoting. Matrices were dissolved as follows: DAN at 10 mg/mL in 7:3 acetonitrile:water, 9-AA  
84 at 10 mg/mL in 9:1 methanol:water, 2',5'-DHAP at 15 mg/mL in 8:2 acetonitrile:water, 2',6'-DHAP  
85 at 10 mg/mL in 1:1 acetonitrile:water, DMAN at 10 mg/mL in 100% acetonitrile and NRM at 10  
86 mg/mL in 7:3 methanol:water. All solvents are LC-MS grade. For each PFAS, two stock  
87 concentrations were prepared: 1 µg/mL (ppm) and 1 ng/mL (ppb). All PFAS excluding F53B  
88 (major) were prepared by dissolution in water to yield a 1mg/mL solution. Immediately after, a

89 1000-fold dilution was performed by transferring 10  $\mu\text{L}$  of the 1mg/mL solution and diluting to 10  
90 mL total volume in water to yield a 1  $\mu\text{g}/\text{mL}$  (ppm) solution. From this 1  $\mu\text{g}/\text{mL}$  solution, another  
91 1000-fold serial dilution was performed, again by transferring 10  $\mu\text{L}$  from the 1  $\mu\text{g}/\text{mL}$  solution and  
92 diluting to a final volume of 10 mL, yielding a 1 ng/mL (ppb) solution. A 1.2 mL methanol ampule  
93 containing 50  $\mu\text{g}/\text{mL}$  of F53B (major) was diluted 5-fold using water to a final volume of 6 mL to  
94 yield a 10  $\mu\text{g}/\text{mL}$  concentration. From this solution, a 10-fold dilution using water was performed  
95 to produce a 1  $\mu\text{g}/\text{mL}$  (ppm) solution. From this solution, a 1000-fold dilution using water was  
96 performed by transferring 10  $\mu\text{L}$  from the 1  $\mu\text{g}/\text{mL}$  solution and diluting to a final volume of 10 mL.  
97 All stock solutions were prepared in water and stored in polypropylene conical tubes. Standards  
98 were made immediately before each experiment and vortexed prior to aliquoting. Sub-ppm  
99 standards of 750, 500 and 250 ppb were prepared by serial dilution from a 1-ppm stock solution.  
100 A 100-ppb standard was prepared by parallel dilution from the 1-ppm stock, and the 50-ppb and  
101 10-ppb standards were prepared by serial dilution from the 100-ppb standard. Sub-ppb standards  
102 were prepared by serial dilution only from a 1-ppb sub-stock solution. Standards were diluted  
103 using a 1:1 methanol:water solution. Organic solvent was added to diluent to increase the rate of  
104 spot-drying and improve matrix crystal homogeneity on target plate.

### 105 2.3 MALDI-TIMS-TOF Analysis

106 MALDI parameters in TIMS-mode were optimized to maximize PFAS signal intensity from 150  
107  $m/z$  to 700  $m/z$ . Laser parameters used in data collection include: M5 defocused MS dried droplet  
108 laser focusing, a laser frequency of 10,000Hz, and 25 laser shots per burst. Ion optics parameters  
109 were also tuned, including a collision cell energy of 10.0 eV, a collision RF of 750.0 Vpp, an ion  
110 transfer time of 50.0  $\mu\text{s}$ , a pre pulse storage time of 5  $\mu\text{s}$ , a TIMS funnel 1 RF of 300.0 Vpp, a  
111 TIMS funnel 2 RF of 200.0 Vpp, a Multipole RF of 200.0 Vpp, a Deflection 1 Delta of -70.0V, a  
112  $\Delta t_4$  (Accumulation Trap  $\rightarrow$  Funnel 1 In) of -25.0 V, a  $\Delta t_6$  (Ramp Start  $\rightarrow$  Accumulation Exit) of -  
113 25.0 V, and a collision cell in voltage of -140.0 V.

114  
115  
116 For MALDI-TIMS-TOF analysis, spotting technique was used. All spotting was performed by  
117 transferring 0.5  $\mu\text{L}$  of analyte onto a ground steel Bruker MTP target plate, followed by overlaying  
118 a drop of 0.5  $\mu\text{L}$  matrix solution onto the analyte drop. Samples were allowed to dry at ambient  
119 temperature, and the formation of crystals was observed under a Zeiss Stemi 305  
120 stereomicroscope. For matrix screening, spots were prepared in technical replicates of  $n=3$  and  
121 randomly sampled 20 times at different locations on the spot for a total of 500 shots. Isomer  
122 differentiation spots were randomly sampled 8 times for a total of 200 shots. Sub-ppm and sub-  
123 ppb standards were spotted in technical replicates of  $n=5$  and randomly sampled 20 times for a  
124 total of 500 shots at laser intensities of 20% and 50%, respectively. The spots were analyzed in  
125 negative-ion mode from 150  $m/z$  to 1100  $m/z$ .

### 126 2.4 Data Analysis

127  
128 Mass spectra were analyzed using Bruker Compass DataAnalysis Version 6.1 and processed  
129 using Microsoft Excel 2019. Spectrum Data was copied from DataAnalysis and pasted into a  
130 Microsoft Excel template. Data in Microsoft Excel was filtered to identify  $m/z$  that belonged to  
131 PFAS. **Figure 1** and **Table 1** contains data analysis from unnormalized, raw peak intensity in the  
132 mass spectra. In **Figure 1**, for easier visualization, logarithmic transformation was performed due

133 to there being several orders of magnitude differences in analyte raw intensities across matrices.  
134 S/N ratio was calculated based on raw spectra. The average and standard deviation of the  
135 baseline intensities adjacent to analyte peaks were determined by sampling 10 background  
136 intensities within the range  $\pm 3 \times$  (the width at the analyte's peak base). Signal-to-noise ratio (S/N)  
137 of the analyte peak was established by subtracting the mean baseline intensity surrounding  
138 analyte peaks from the observed analyte intensity and dividing by the standard deviation of the  
139 baseline signal. Standard deviation of S/N ratios for technical replicates were calculated using the  
140 STDEV.S function in Excel.

141  
142 For **Table 1**, we estimated LOD/LOQ values from raw mass spectra (sub-ppm) and mobility-  
143 filtered mass spectra (sub-ppb). For sub-ppm range, the raw peak intensities from the mass list  
144 of raw mass spectra were used to represent analyte signals. For sub-ppb range, mobility filtering  
145 was performed to extract a filtered analyte peak intensity. Extracted ion mobility (EIM) of the  
146 analyte was performed from the raw mobility- $m/z$  spectrum by filtering the mobilogram at analyte  
147  $m/z \pm 10$  ppm. The mobility peak (Compounds) corresponding to the analyte was identified by the  
148 DataAnalysis and then selected to show the mass spectrum filtered by the range of the Compound  
149 peak. The filtered intensity of analyte was used to represent mobility-filtered analyte peak  
150 intensity. With the analyte peak intensity, linear regressions were performed on their averages of  
151 the technical replicates; standards with average S/N < 10 were not included for linear regression.  
152 Estimated LOD and LOQ intensities were established as 3 standard deviations and 10 standard  
153 deviations above background noise present in blanks, respectively. Estimated LOD and LOQ  
154 concentrations were calculated by dividing the LOD and LOQ intensities to slopes of the best fit  
155 lines produced from analyte peak intensities. Slopes produced by intensities from multi-day  
156 calibration curve trials, LOQ, LOD and S/N ratios were averaged with variation described as  
157 relative standard deviation percentage (RSD%).

158  
159 We then performed total ion current (TIC) normalization to help reduce variations across  
160 replicates and build calibration curves using the standards with average S/N > 10 that were  
161 identified from raw spectra. TIC was calculated by summing all intensities in the mass list found  
162 by DataAnalysis in each raw mass spectrum. TIC normalization was performed by dividing analyte  
163 peak raw intensity (sub-ppm, **Figure 2**) or mobility-filtered analyte peak intensity (sub-ppb, **Figure**  
164 **3**) by the TIC. TIC normalized analyte signals were averaged across technical replicates using  
165 the AVERAGE function, and standard deviation of the replicate sample sets were calculated using  
166 the STDEV.S function. Relative standard deviation percentage (RSD%) was calculated by  
167 dividing the standard deviation of replicates by the mean and converted to a percentage value.

168

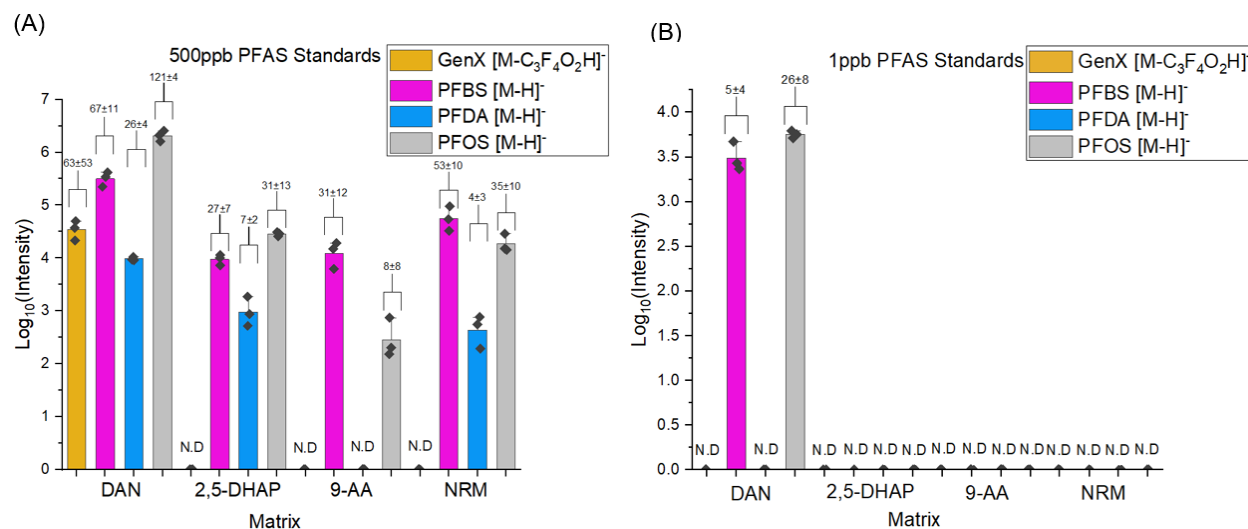
### 169 **3. Results and Discussion**

170

171 **Matrix screening showed DAN as a suitable matrix for all selected PFAS.** Matrix selection is  
172 essential for efficient energy transfer from laser to analyte and should be capable of ionizing  
173 analytes by protonation or deprotonation.<sup>31</sup> Several commonly used matrices (DMAN, 2,6-DHAP,  
174 DAN, 2,5-DHAP, 9-AA, NRM) for negative ion mode analysis were chosen for the detection of  
175 PFCAs and PFSA. Preliminary investigation into these matrices revealed that DMAN and 2,6-  
176 DHAP had poor vacuum stability and were prone to sublimation, rendering them unusable in this



177 study. Four PFAS were chosen as representatives for both PFCA and PFSA families: GenX,  
 178 PFDA, PFBS and PFOS. As shown in **Figure 1A**, DAN was the only matrix to produce signals  
 179 that could be attributed to all four PFAS and produced the highest intensities and S/N ratios across  
 180 all PFAS screened. When comparing spectra of each PFAS across matrices, PFCAs showed  
 181 significant fragmentation when DAN was used as the matrix whereas PFSA showed minimal  
 182 fragmentation. Similarly reported by Cao et al., PFDA produced signals at  $m/z$  468.9696 [M-  
 183 CO<sub>2</sub>H]<sup>-</sup> and  $m/z$  430.9728 [M-2F-CO<sub>2</sub>H]<sup>-</sup> in addition to  $m/z$  512.9594 [M-H]<sup>-</sup> (**Figure S1A**).<sup>22</sup>  
 184 Notably, the [M-2F-CO<sub>2</sub>H]<sup>-</sup> peak appeared to be the highest among all PFDA peaks. NRM  
 185 induced some fragmentation with PFDA, yielding only [M-CO<sub>2</sub>H]<sup>-</sup> and [M-H]<sup>-</sup> signals to be  
 186 detected (**Figure S1B**). 2,5-DHAP only yielded molecular ion peaks and no fragmentation ions  
 187 were detected, which may be related to the low intensity of its molecular ion peak (**Figure S1C**),  
 188 and 9-AA did not produce observable PFDA peaks (**Figure S1D**). Contrary to PFDA, GenX  
 189 showed no [M-H]<sup>-</sup>, [M-CO<sub>2</sub>H]<sup>-</sup>, or [M-2F-CO<sub>2</sub>H]<sup>-</sup> related peaks. Instead, fragmentation  
 190 occurred at the ether present in the fluorinated chain to produce a signal at  $m/z$  184.9837 which  
 191 corresponds to the formula [C<sub>3</sub>F<sub>7</sub>O]<sup>-</sup> (**Figure S1E**). To further investigate the detection sensitivity  
 192 of these matrices, concentrations for each PFAS were reduced to 1ppb. Illustrated by **Figure 1B**,  
 193 PFCAs were not detected across all matrices, however, both PFASs were still detectable with  
 194 DAN. Thus, DAN was chosen as the chemical matrix for PFAS analysis for the subsequent  
 195 experiments.  
 196



197 **Figure 1.** Matrix screening for (A) 500 ppb and (B) 1 ppb of PFAS solution. Different colored bars  
 198 represent the average signal intensities of ions from a specific PFAS, and solid diamonds indicate  
 199 individual replicates (n=3). Error bars represent standard deviation. Numbers above each bar  
 200 indicate the S/N ratio of the signal of interest from raw spectra (average ± SD). Regions labeled  
 201 N.D represent nondetectable signals with S/N < 3. The Y-axis is plotted in logarithmic scale.  
 202

203 **Multi-day calibration curves demonstrated PFAS quantification at different ranges.** Matrix  
 204 optimization data demonstrated parts per billion detection and quantification for select PFCA and  
 205 PFSA. To investigate the quantitative capability of MALDI-TIMS-TOF detection, calibration curves  
 206 of sub-ppm and sub-ppb ranges for PFAS compounds were prepared in technical replicates (n=5),  
 207

208 and the calibration curve was freshly made for repeated analysis for at least three times across  
209 multiple days. The results were summarized in **Table 1** and illustrated in **Figure 2 and 3**. Of the  
210 calibration curves analyzed, one GenX calibration curve was excluded due to erroneous standard  
211 preparation, where significantly different peak intensities and linearity were observed. It should  
212 be noted that the calibration curves for sub-ppm and sub-ppb are not directly comparable due to  
213 the different laser intensities used for different concentration ranges. From **Table 1**, it is clear that  
214 the slopes of calibration curves (sensitivity) and LOD/LOQ greatly differ among PFAS types. For  
215 sub-ppm range, PFASs showed better detection with orders of magnitude higher sensitivities and  
216 lower LOD/LOQ values compared to PFCAs. We observed that among different chain lengths of  
217 legacy PFCAs and PFSA series, PFOA and PFOS showed the best detection with highest slope  
218 and smallest LOD/LOQ. This may be attributed to the TIMS parameters we optimize that prefer  
219 the middle part of the calibration range (150-1100  $m/z$ ). The reproducibility of sub-ppm and sub-  
220 ppb calibration curves with multi-day experiments are evaluated by the slopes and their RSD%  
221 values (**Table 1**). Most of the RSD% of slopes determined from multi-day experiments were  
222 around 30%, which was expected considering the semi-quantitative nature of MALDI-MS  
223 analysis. It should be noted that normalization by TIC did not seem to reduce the RSD% of slope  
224 across days (**Table S1**). Reproducibility may be improved by using internal standards and  
225 improving the homogeneity of matrix crystals.<sup>32</sup> Spotting method for matrix application may cause  
226 heterogenous crystal formation; a spray-coating or sublimation application technique can be used  
227 to improve crystal heterogeneity.

228  
229 The individual sub-ppm calibration curves for PFCAs were plotted in **Figure 2A-D**. Across all  
230 legacy PFCAs tested (PFHxA, PFOA, and PFDA),  $[M-2F-CO_2H]^-$  peaks yielded the highest  
231 signal intensity, followed by  $[M-CO_2H]^-$  peaks with  $[M-H]^-$  peaks producing the least signal  
232 intensity (**Figure S1A**). Knowing this,  $[M-2F-CO_2H]^-$  signals were used as the signal for  
233 quantification in calibration curves with the exception of GenX whose only signal was an ion with  
234 the putative fragment  $[M-C_3F_4O_2]^-$ . Such fragmentation may be attributed to the metastable ions  
235 formed from MALDI ionization. As fragmentation results in a series of ions at different  $m/z$ , it may  
236 also contribute to the decreased sensitivity of PFCAs compared to PFASs, which primarily formed  
237 molecular ions. Shown in **Table 1** and **Figure 2E-H**, PFASs demonstrated greater sensitivity and  
238 lower LOD/LOQ compared to PFCAs. All PFSA indicated potential for parts per trillion limits of  
239 detection based on LOD/LOQs calculated based on the calibration curves. As observed in GenX  
240 where fragmentation occurred at the site of the ether present in the fluorinated chain,  
241 fragmentation at this site in F53B (Major) was also observed at  $m/z$  350.9446 (**Figure S1F**). In  
242 contrast to GenX, this fragment ion was of low intensity, with the dominant ion peak being the  $[M-$   
243  $H]^-$  signal.

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**Table 1.** Sensitivity, limit of detection/quantification, and S/N ratios from multi-day experiments.

<b>Sub-ppm calibration curves. Values shown as average (RSD%)#</b>						
		Slope (peak intensity vs. concentration)	Calculated LOD (ppb)*	Calculated LOQ (ppb)*	Lowest standard with average S/N≥10 (ppb)**	S/N of the lowest concentration tested (10 ppb)
PFCA	GenX	31.5 (60.3%)	7 (40.7%)	22 (40.7%)	50	5 (11.8%)
	PFHxA	8.45 (31.3%)	50 (46.2%)	165 (46.2%)	250	3 (27.4%)
	PFOA	103 (29.1%)	3 (29.9%)	8 (29.9%)	50	7 (18.0%)
	PFDA	50.8 (66.4%)	5 (87.3%)	17 (87.3%)	10	10 (49.4%)
PFSA	PFBS	528 (27.7%)	0.96 (106%)	3.22 (106%)	10	13 (26.7%)
	PFHxS	652 (19.1%)	0.19 (22.3%)	0.64 (22.3%)	10	30 (50.8%)
	PFOS	977 (36.2%)	0.08 (30.4%)	0.25 (30.4%)	10	36 (14.5%)
	F53B	750 (49.7%)	0.06 (68.1%)	0.18 (68.1%)	10	28 (49.7%)
<b>Sub-ppb calibration curves. Values shown as average (RSD%)#</b>						
		Slope (peak intensity vs. concentration)	Calculated LOD (ppt)*	Calculated LOQ (ppt)*	Lowest standard with average S/N≥10 (ppt)**	S/N of the lowest concentration tested (100 ppt)
PFSA	PFBS	27.2 (38.9%)	34 (101%)	112 (101%)	100	21 (25.1%)
	PFHxS	36.3 (30.6%)	24 (58.4%)	82 (58.4%)	100	20 (25.3%)
	PFOS	97.6 (21.8%)	6 (47.5%)	20 (47.5%)	100	20 (4.91%)
	F53B	30.7 (27.9%)	22 (97.6%)	74 (97.6%)	100	23 (34.4%)

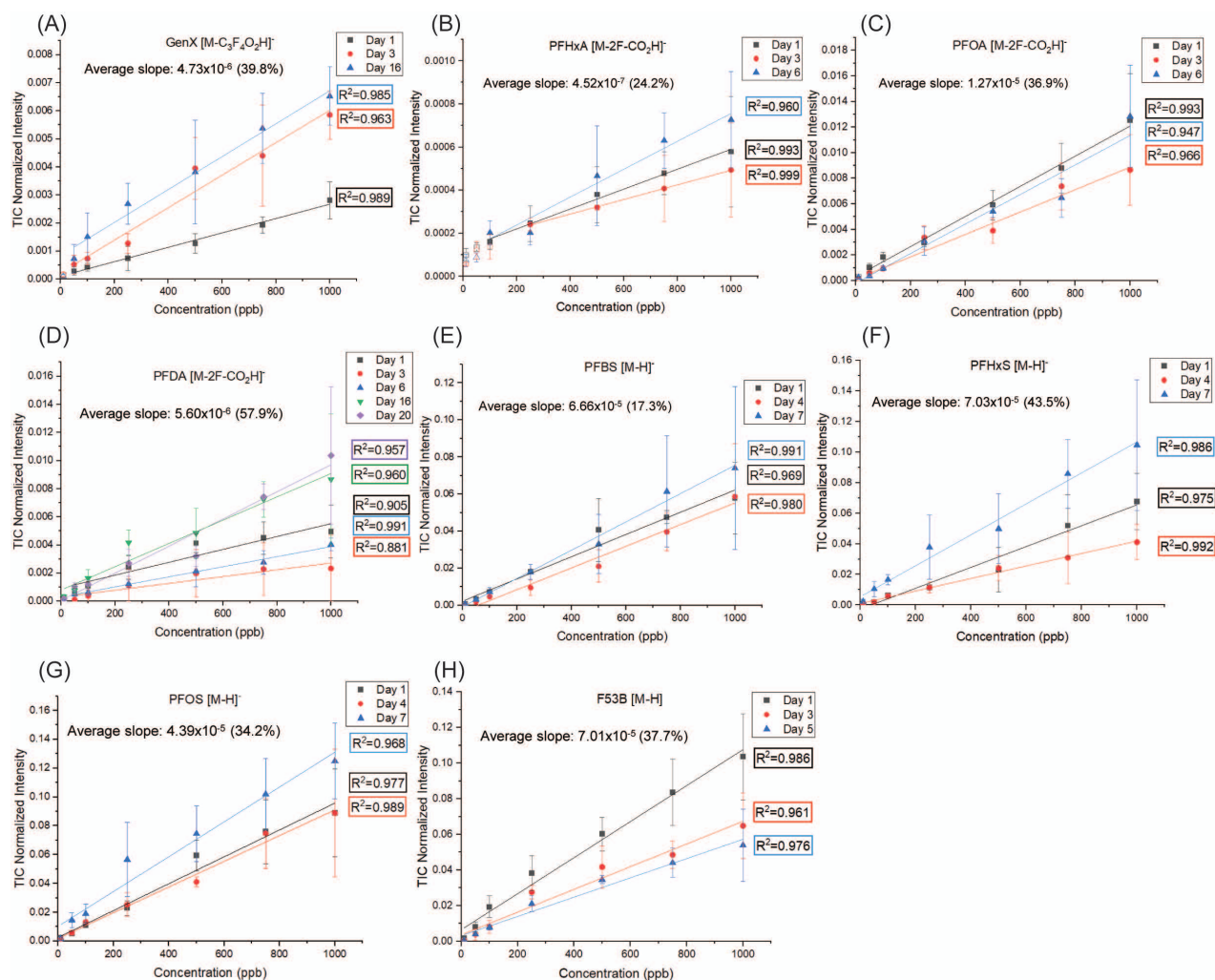
254 # Average and RSD calculated from triplicates of calibration curves from multiple days. For sub-ppm range, raw peak  
255 intensities of analytes were used. For sub-ppb range, mobility-filtered peak intensities of analytes were used.

256 \* LOD was calculated using 3\*blank SD/slope and LOQ was calculated using 10\*blank SD/slope.

257 \*\* Lowest calibrant with an average S/N≥10 measured across all multi-day experiments.

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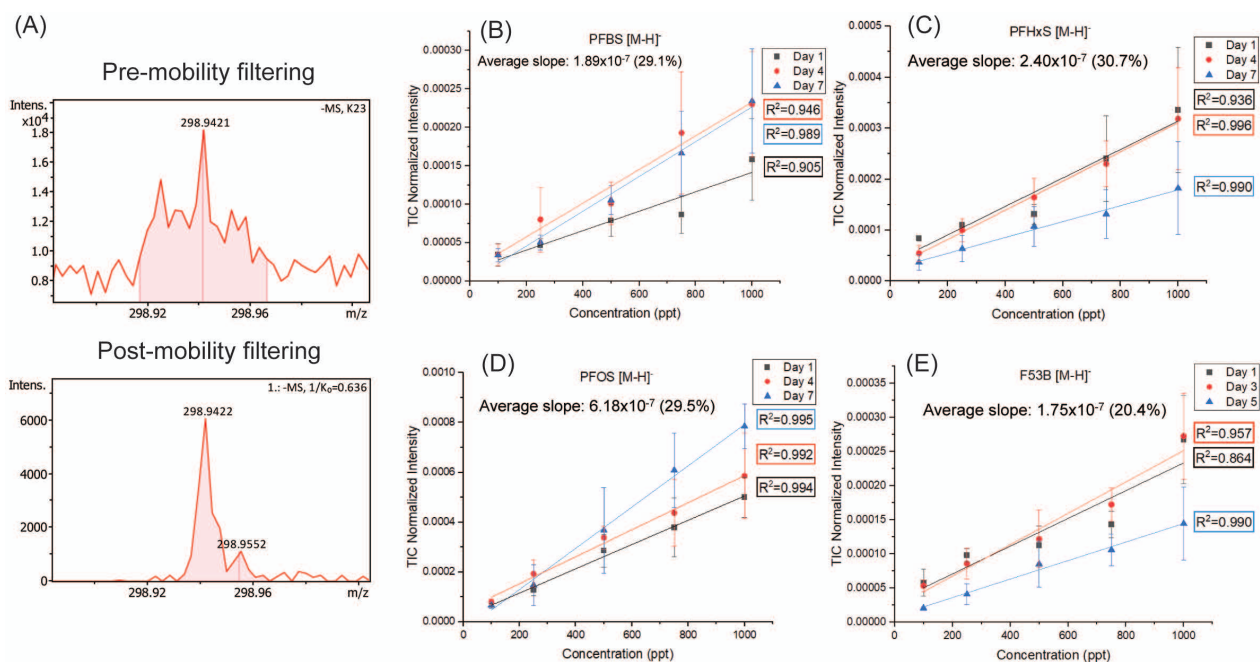




260  
 261 **Figure 2.** Sub-ppm calibration curves for (A-D) PFCAs and (E-H) PFSA. Data points represent  
 262 average normalized intensity from the five replicates in each experiment, and error bar represents  
 263 standard deviation. Solid shapes indicate data points with average S/N of raw spectra over 10  
 264 and used for linear regression of calibration curves. Hollow shapes represent detectable data  
 265 points with S/N less than 10 and not included for linear regression. Averages and RSD% for multi-  
 266 day TIC normalized slopes are presented below each calibration curve.  
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269 **Ion mobility spectrum filtering enabled quantification at sub-ppb (ppt) range.** Based on the  
 270 S/N and calculated LOD/LOQ of PFSA calibrants, we further explore the detection of PFSA in  
 271 the parts per trillion (ppt) range. To improve the lower limits of detection and quantitation, laser  
 272 intensity was increased from 20% to 50%, and mobility filtering was performed to reduce  
 273 background signal in the mass spectra. Perfluorinated molecules were reported to separate in the  
 274 CCS-*m/z* 2D space due to the high mass of fluorine atoms, and our TIMS-MS spectra matched  
 275 the trends reported in DTIMS (**Figure S2**).<sup>28</sup> Results showed that filtering the mass spectrum by  
 276 the mobility range of analytes of interest successfully decreased the backgrounds and increased  
 277 detection sensitivity (**Figure 3A, Figure S3**). Notably, upon mobility filtering of blanks (**Figure**  
 278 **S3**), it was found that PFAS-associated signal intensities were still present, indicating that TIMS

279 can be used to identify cross-contamination that would otherwise be indistinguishable from non-  
 280 mobility filtered intensities. With that, we performed mobility filtering for all sub-ppb mass spectra  
 281 of PFSA to build calibration curves from the multi-day experiments. Shown in **Figure 3B-E** and  
 282 **Table 1**, calibration curves were built in the ppt range, with limits of detection at low parts per  
 283 trillion levels. We noticed that the slopes were comparable for PFOS and F53B (Major) at sub-  
 284 ppm concentrations, but at sub-ppb concentrations F53B (Major) yielded relatively lower slopes  
 285 compared to PFOS. It is suspected that as laser intensity increased from 20% to 50%,  
 286 fragmentation at the ether present in the fluorinated chain of F53B (Major) increased, thereby  
 287 decreasing the intensity and subsequently, the sensitivity of the  $[M-H]^-$  base peak.  
 288

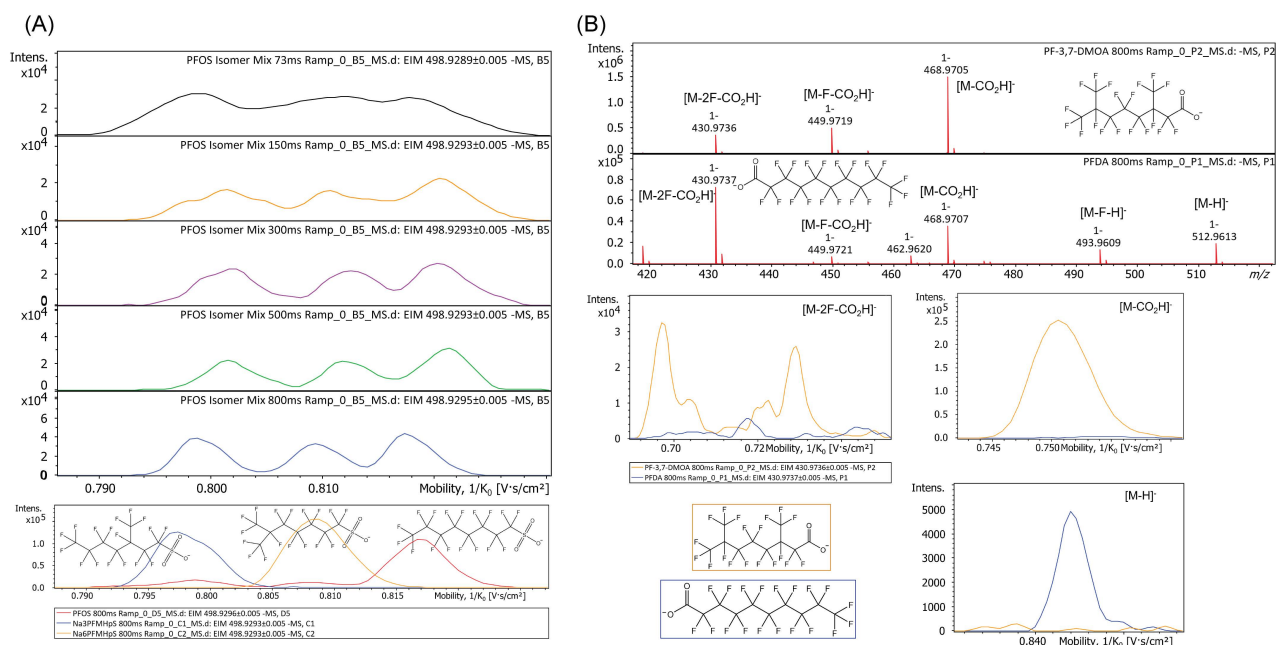


289 **Figure 3.** Sub-ppb calibration curves for PFASs. (A) An example of comparison between pre- and  
 290 post-mobility filtering for PFBS  $[M-H]^-$  ion. (B-E) Calibration curves for PFBS, PFHxS, PFOS and  
 291 F53B. Data points represent average normalized intensity from the five replicates in each  
 292 experiment, and error bar represents standard deviation. Solid shapes indicate data points with  
 293 average S/N of raw spectra over 10 and used for linear regression of calibration curves. Hollow  
 294 shapes represent data points with S/N less than 10 and not included for linear regression.  
 295 Averages and RSD% for multi-day TIC normalized slopes are presented below each calibration  
 296 curve.  
 297

298  
 299 **Ion mobility and fragmentation patterns allowed differentiation of PFAS structural isomers.**  
 300 It was demonstrated by Spraggins et al. that TIMS could differentiate isobaric lipid species.<sup>30</sup>  
 301 Isomer differentiation by TIMS was evaluated using three PFSA isomers and two PFCA isomers:  
 302 PFOS, sodium perfluoro-3-methylheptanesulfonate (Na3PMHpS) and sodium perfluoro-6-  
 303 methylheptanesulfonate (Na6PMHpS), PFDA and perfluoro-3,7-dimethyloctanoic acid (PF-3,7-  
 304 DMOA). Improved isomer separation was achieved by reducing the mobility window to 0.65-0.90  
 305

306 and increasing ramp time to allow greater separation by TIMS. It should be noted that reducing  
307 the mobility window compromised mobility calibration as the detection window no longer contains  
308 enough calibrants in the Agilent Tune Mix. As shown in **Figure 4A**, PFOS isomer differentiation  
309 was sufficiently achieved using 300ms ramp time and further improved when increased to 800ms.  
310 Upon increasing ramp time, it was discovered that the non-analytical standard linear PFOS  
311 compound contained isomeric impurities (**Figure 4A**, red trace at the bottom mobilogram). These  
312 impurities can affect the quality of analytical testing with MS and affect downstream toxicological  
313 assays. Therefore, these factors should be taken into consideration when procuring perfluorinated  
314 compounds for research purposes. PFOS isomers were measured at each ramp time individually  
315 to confirm peak identity in the mixture. As expected, Na<sub>3</sub>PMHpS had the smallest measured ion  
316 mobility given it was the least bulky ion, followed by Na<sub>6</sub>PMHpS and lastly, PFOS. Two-peak  
317 resolution ( $R_{pp}$ ) was calculated to measure the degree of separation between Na<sub>3</sub>PMHpS and  
318 Na<sub>6</sub>PMHpS, as well as Na<sub>6</sub>PMHpS and PFOS. Maximum  $R_{pp}$  for Na<sub>3</sub>PMHpS/Na<sub>6</sub>PMHpS and  
319 Na<sub>6</sub>PMHpS/PFOS was achieved using an 800ms ramp time and was calculated to be 0.907 and  
320 0.787, respectively. Isomer differentiation was achieved from a 1-ppm mixture of all PFOS at  
321 equal concentrations of 333-ppb for each isomer. At low concentrations, isomer differentiation  
322 became significantly more challenging due to sensitivity issues.

323  
324 Isomer differentiation with TIMS was also explored to differentiate PFDA and PF-3,7-DMOA, a  
325 branched isomer of linear PFDA. Different from PFSA like PFOS, PFCAs have thus far  
326 demonstrated significant fragmentation when ionized with DAN, which allowed for multiple ion  
327 peaks to be utilized for differentiation by mobility filtering. Illustrated by **Figure 4B**, PFDA and PF-  
328 3,7-DMOA yielded significantly different fragmentation patterns which prevented direct  
329 comparison of ion mobilities due to differences in signal intensity. Contrary to all linear legacy  
330 PFCA examined thus far, PF-3,7-DMOA's dominant ion signal appeared to be  $[M-CO_2H]^-$  rather  
331 than  $[M-2F-CO_2H]^-$ . Due to the significant difference in ion fragment formation, EIM filtering was  
332 not applicable, however, utilizing these "diagnostic fragment" intensities can allow for identification  
333 of isomers based on a ratio of intensities,<sup>33</sup> which will be explored in future investigations.  
334



335  
 336 **Figure 4.** Isomer differentiation with TIMS for (A) PFOS and (B) PFDA isomers. Structures of  
 337 different isomers are shown. (A) The top five show mobilograms with increasing ramp time from  
 338 73 to 800 ms for the separation of three PFOS structural isomers. Bottom panel shows the  
 339 mobilogram of three individual PFOS isomer standards at 800-ms ramp time, marked with  
 340 different colors and corresponding structures. (B) The top two mass spectra show different  
 341 fragmentation patterns of two PFDA structural isomers. The extracted mobilograms of selected  
 342 fragments and molecular ions are individually shown with colors marking different isomers.

343  
 344

#### 345 4. Conclusions

346 We developed a MALDI-TIMS-TOF MS method for the analysis of a list of PFAS that allowed for  
 347 parts per billion and parts per trillion detection and quantification of select PFCAs and PFSA,  
 348 respectively. The timsTOF fleX instrument was optimized for analysis of PFAS in relevant  $m/z$   
 349 and mobility ranges. Matrix screening at negative mode indicated DAN (1',5'-  
 350 diamionaphthalene) to be the optimal matrix suitable for all PFSA and PFCAs tested in this  
 351 study, although other options may achieve less fragmentations of PFCAs. We generated multi-  
 352 day calibration curves; without the usage of internal standards, we achieved good linearity in the  
 353 concentration range we tested and showed reproducibility that reflects the semi-quantitative  
 354 nature of MALDI-MS analysis. By filtering the mass spectra with the mobilities of specific PFAS  
 355 analytes, we were able to decrease background and increase the sensitivity of detection. Finally,  
 356 we achieved differentiation of structural PFOS isomers using TIMS and also showed different  
 357 fragmentation patterns of PFDA isomers, which may be used as diagnostic patterns for isomer  
 358 differentiation. Our results showed the potential of MALDI-TIMS-TOF MS analysis as a promising  
 359 method for in situ, high-throughput analysis of PFAS with low detection limit and capability of  
 360 isomer differentiation.



- 361  
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365
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