THE UNIVERSITY OF RHODE ISLAND

University of Rhode Island DigitalCommons@URI

Graduate School of Oceanography Faculty Publications

Graduate School of Oceanography

2016

Source attribution of poly- and perfluoroalkyl substances (PFASs) in surface waters from Rhode Island and the New York Metropolitan Area

Xianming Zhang

Rainer Lohmann University of Rhode Island, rlohmann@uri.edu

See next page for additional authors

Follow this and additional works at: https://digitalcommons.uri.edu/gsofacpubs

The University of Rhode Island Faculty have made this article openly available. Please let us know how Open Access to this research benefits you.

This is a pre-publication author manuscript of the final, published article.

Terms of Use

This article is made available under the terms and conditions applicable towards Open Access Policy Articles, as set forth in our Terms of Use.

Citation/Publisher Attribution

Zhang, X., Lohmann, R., Dassuncao, C., Hu, X. C., Weber, A. K., Vecitis, C. D., & Sunderland, E. M. (2016). Source Attribution of Poly- and Perfluoroalkyl Substances (PFASs) in Surface Waters from Rhode Island and the New York Metropolitan Area. *Environ. Sci. Technol.*, *3*(9), 316-321.

Available at: http://dx.doi.org/10.1021/acs.estlett.6b00255

This Article is brought to you for free and open access by the Graduate School of Oceanography at DigitalCommons@URI. It has been accepted for inclusion in Graduate School of Oceanography Faculty Publications by an authorized administrator of DigitalCommons@URI. For more information, please contact digitalcommons@etal.uri.edu.

Authors

Xianming Zhang, Rainer Lohmann, Clifton Dassuncao, Xindi C. Hu, Andrea K. Weber, Chad D. Vecitis, and Elsie M. Sunderland

- 1 Source attribution of poly- and perfluoroalkyl substances (PFASs) in surface
- 2 waters from Rhode Island and the New York Metropolitan Area
- 3 Xianming Zhang^{$\dagger \ddagger *$}; Rainer Lohmann[§]; Clifton Dassuncao^{$\dagger \ddagger \ddagger}$ </sup>; Xindi C. Hu^{$\dagger \ddagger}$ </sup>; Andrea K. Weber^{\dagger};
- 4 *Chad D. Vecitis*^{\dagger}; *Elsie M. Sunderland*^{\dagger ‡}
- ⁵ [†] Harvard John A. Paulson School of Engineering and Applied Sciences, Harvard University,
- 6 Cambridge MA USA 02138
- 7 [‡] Department of Environmental Health, Harvard T.H. Chan School of Public Health, Harvard
- 8 University, Boston MA USA 02115
- 9 [§] Graduate School of Oceanography, University of Rhode Island
- 10 *Corresponding author:
- 11 Xianming Zhang, E-mail: xmzhang@seas.harvard.edu; Tel: 617-495-2893

12 TOC Art



14 Abstract

15 Exposure to poly and perfluoroalkyl substances (PFASs) has been associated with adverse health 16 effects in humans and wildlife. Understanding pollution sources is essential for environmental 17 regulation but source attribution for PFASs has been confounded by limited information on 18 industrial releases and rapid changes in chemical production. Here we use principal component 19 analysis (PCA), hierarchical clustering, and geospatial analysis to understand source 20 contributions to 14 PFASs measured across 37 sites in the Northeastern United States in 2014. 21 PFASs are significantly elevated in urban areas compared to rural sites except for 22 perfluorobutane sulfonate (PFBS), N-methyl perfluorooctanesulfonamidoacetic acid (N-23 MeFOSAA), perfluoroundecanate (PFUnDA) and perfluorododecanate (PFDoDA). The highest PFAS concentrations across sites were for perfluorooctanate (PFOA, 56 ng L^{-1}) and 24 perfluorohexane sulfonate (PFHxS, 43 ng L⁻¹) and perfluorooctanesulfonate (PFOS) levels are 25 26 lower than earlier measurements of U.S. surface waters. PCA and cluster analysis indicates three 27 main statistical groupings of PFASs. Geospatial analysis of watersheds reveals the first 28 component/cluster originates from a mixture of contemporary point sources such as airports and 29 textile mills. Atmospheric sources from the waste sector are consistent with the second 30 component, and the metal smelting industry plausibly explains the third component. We find this 31 source-attribution technique is effective for better understanding PFAS sources in urban areas.

32 Introduction

33 Exposure to poly- and perfluoroalkyl substances (PFASs) has been associated with many 34 negative health outcomes including compromised immune function, metabolic disruption, obesity, and altered liver function.¹ PFASs in surface waters are an emerging concern for U.S. 35 36 public water supplies and long-chain compounds bioaccumulate in aquatic food webs, posing health risks to seafood consumers.²⁻⁶ Production of PFASs and their precursors has shifted 37 dramatically over the last two decades toward shorter-chain and polyfluorinated species.⁷ 38 39 Diverse point sources and atmospheric deposition of some PFASs confounds understanding of 40 the dominant contributors to contamination in the aquatic environment. Regulatory databases such as the U.S. EPA's Facility Registry Survey (FRS)⁸ and the Toxic Release Inventory⁹ 41 42 presently contain limited to no information on magnitudes of PFASs released to the environment. 43 Multivariate statistical analyses based on chemical composition profiles can be a 44 powerful tool for diagnosing contamination sources, as illustrated for many other organic contaminants.¹⁰ Principal components analysis (PCA) provides information on 45 46 interrelationships among various chemicals and is useful for deriving common source 47 profiles. Two-way hierarchical clustering can be used as a confirmatory analysis of PCA by 48 generating a flexible number of subgroups of similar sites (those affected by a common 49 source type) without dictating the number of clusters *a priori*. Clustering of compounds 50 identifies chemicals that co-occur to form a unique signature. These techniques have not 51 been routinely applied to interpret PFAS contamination and show potential for interpreting sources in surface water and seawater.^{4,11} 52

Here we combine PCA and hierarchical clustering of PFAS profiles measured in surface
 waters from 37 rivers, streams and estuaries in the Northeastern United States with geospatial

55 analysis of potential sources. Few measurements are available for PFASs in U.S. surface waters 56 over the past five years and the importance of different sources is poorly understood. Source regions for air pollution are commonly identified using back trajectories.^{12,13} We apply an 57 58 analogous approach for identifying sources of aquatic pollution based on hydrological 59 distances within a watershed. The main objective of this study is to identify major sources of 60 surface water PFAS contamination in diverse watersheds using information on chemical 61 composition and geospatial analytical tools that consider surface hydrology. 62 Methods 63 Sample collection and analysis 64 We collected surface water samples from rivers/creeks and estuaries at approximately 1 65 m depth at 28 sites in the state of Rhode Island (RI) in June, 2014 and 9 sites the New York 66 Metropolitan Area (NY/NJ) in October, 2014 (Figure 1). A complete description of sampling 67 sites is provided in the Supporting Information (SI Table S1). Precipitation and flow rates in 68 rivers tend to be higher in June, potentially resulting in enhanced dilution and a low bias for 69 some PFASs measured in RI rivers compared to NY/NJ. 70 Samples were stored in one-liter pre-rinsed polypropylene bottles at -20 °C and thawed to 71 room temperature. Each sample was homogenized by shaking vigorously before subsampling 72 500 ml for the analysis of 21 PFASs. Each unfiltered sample was spiked with 20 µL of a 0.1 ng µL⁻¹ mass labeled PFAS mixture (Wellington; Guelph, Canada; individual compounds are listed 73 74 in Table S2) as internal standards for quantification. PFASs were extracted using an Oasis Wax 75 solid phase extraction (SPE) cartridge (6 mL, 150 mg sorbent) following the method of Taniyasu et. al.¹⁴ (see SI Section S1 for details). A nitrogen evaporator (ZIPVAP) was used to concentrate 76

77 the extract to 1 mL (methanol: water; v:v = 1:1).

78	Sample detection for 21 native PFASs (Tables S2, S3) was performed using an Agilent
79	6460 LC-MS/MS equipped with an online-SPE system (Agilent 1290 Infinity Flex Cube) in
80	dynamic multiple reaction mode (sample chromatogram in Figure S1). At least one negative
81	control (field or procedural blank) and one positive control (spiked with 2 ng of the 21 PFASs in
82	500 ml water) were included in every extraction batch. Whole method recovery tested using the
83	positive controls was 70-120% for all but 4 PFASs that ranged from 60-70%, which is
84	comparable to recoveries reported by previous studies. ^{3,14,15} . The 4 PFASs are perfluoropentanate
85	(PFPeA), perfluoroheptanate (PFHpA), N-methyl perfluorooctanesulfonamidoacetic acid
86	(MeFOSAA) and N-ethyl perfluorooctanesulfon-amidoacetic acid (EtFOSAA). Potential analyte
87	loss during sample preparation was corrected using internal standards spiked prior to sample
88	extraction. The limit of detection (LOD, Figure S2) was defined as equivalent to the blank plus
89	the concentration corresponding to a signal-to-noise ratio of three. Variability between duplicates
90	obtained at two sites was <20%. Concentrations of PFASs in five field blanks (HPLC grade
91	water) prepared following the sample preparation procedure were all below the LOD.
92	We quantified branched isomers for perfluorooctanate (PFOA), perfluorohexane
93	sulfonate (PFHxS), perfluorooctane sulfonate (PFOS), N-MeFOSAA and N-EtFOSAA using
94	calibration standards for the linear isomers, assuming the same instrumental response factor
95	(Table S3). Seven compounds namely perfluorododecane sulfonate (PFDS), 8:2 fluorotelomer
96	sulfonate (8:2 FTS), perfluorooctane sulfonamide (FOSA), and perfluorocarboxylates with more
97	than 12 carbon atoms) were detected in less than half of samples and were excluded from
98	additional statistical analysis (see Table S2 for details). For the 14 PFASs that had detection
99	frequencies of greater than 60% (Table S2), we used the Robust Regression on Order Statistics

100 approach for censored log-normally distributed environmental data described by Helsel¹⁶ to

101 assign values to samples with concentrations below the LOD.

102 Statistical and spatial analysis

We used principal components analysis (PCA) and hierarchical clustering to group sites with statistically distinct PFAS composition profiles. PCA was performed using MATLAB's Statistics Toolbox (MathWorks, Inc.) on normalized (z-score to remove the effect of concentration difference at different sites) PFAS concentration data. The inverse of variances of the data were used as variable weights and varimax rotation was applied to interpret the meaning of extracted principal components. Hierarchical Cluster analysis was conducted using the hclust function in the R statistical computing package (version 3.1.3).

110 We characterized the watershed for each freshwater sampling site using the U.S. 111 Geological Survey's (USGS) National Elevation Dataset (3 arc-second for site 15 and 16 and 1 112 arc-second for others) and the Hydrologic Tool in ArcGIS Pro 1.2 and ArcGIS online. Estuarine 113 sampling sites were excluded from the geospatial analysis due to the confounding influence of 114 tidal waters on potential source profiles. Population within each watershed was based on ESRI's U.S. Demographic Database.¹⁷ We used the USGS's StreamStats database (version 4)¹⁸ to 115 116 characterize water flow rates for each location and to compute mass flow (kg/yr) of PFASs at 117 each site and per-capita mass flows (kg/person/yr).

For all inland sites (non-estuarine), we acquired a list and geospatial data for plausible PFAS sources from the US EPA Facility Registry Service (FRS) database on facilities and sites subject to environmental regulation (see SI for the search criteria).⁸ These include airports, facilities for metal plating/coating, printing, sewage treatment, waste management (including landfills), and manufacturers of semiconductor, textile, paint/coating/adhesive, ink, paper, and

petroleum products. A caveat of this analysis is that not all facilities included in the FRS
database necessarily release PFASs and the database may not comprehensively include all
possible sources.

126 Hydrological distances of point sources from each sampling site were computed using the 127 ArcGIS Trace Downstream tool. Within each watershed, we defined an indicator for the impact 128 of potential point sources as a function of distance from sampling locations by assuming 129 exponential decay in the source signature¹⁹ (i.e., *impact* = $1/e^d$, where d = hydrological distance, 130 km). This approach provides additional information on plausible sources that complements 131 multivariate statistical analysis but cannot be considered a quantitative estimate of contributions 132 to sampling locations since magnitudes of PFAS discharges are not available.

133 **Results and Discussion**

134 Concentrations and spatial patterns

135 Figure 1 shows the compound specific composition and concentrations of PFASs 136 measured in surface water samples as part of this work. Sampling sites in NY/NJ had much 137 greater population density in upstream watersheds (10 to 43-fold) compared to RI but the highest 138 concentrations of most PFASs were measured near the city of Providence, RI (Figure 1, Figure 139 S2). The range of measured PFAS concentrations reported here are similar or lower than U.S. surface waters from other regions collected between 2000-2009 (Table S4).^{2,20-24} 140 141 All sites had detectable PFOA and PFNA and over 90% contained detectable PFHxS, 142 PFOS, PFDA, and 6:2 FtS (Table S2, S3, Figure S2). The highest individual PFAS 143 concentration across sites was PFOA (56 ng L⁻¹) at Site 31 (Passaic River, NJ). The highest concentrations of PFHxS (43 ng L⁻¹) and PFNA (14 ng L⁻¹) were measured at Site 5 (Mill Cove, 144 RI). The maximum PFOS concentration (27 ng L⁻¹) was measured at Site 2 (Woonasquatucket 145

River, RI) within the City of Providence, RI. This is much lower than maximum levels reported
in earlier studies of US surface waters that range between 43-244 ng L⁻¹ (Table S4) and reflects
the continued decline in environmental PFOS burdens in North America following elimination of
production in 2002.^{25,26}

150

151 Measured PFAS concentrations in urban regions were significantly higher (Wilcoxin rank 152 sum test, p<0.017) than rural sites for all compounds except PFBS, N-MeFOSAA, PFUnDA and 153 PFDoDA (Figure S3). Sites 1-11 in RI and Sites 29-37 in NY/NJ are all urban areas, defined by 154 population densities of greater than 1000 individuals per square mile (2590 km²), and population 155 densities of greater than 500 individuals per square mile in surrounding census blocks.²⁷ We did 156 not find a statistically significant correlation between total population in each upstream 157 watershed and PFAS concentrations measured at each sampling site (p=0.12 to 0.95 across 158 compounds). We derived per-capita discharges (Figure S4) using a similar approach as Pistocchi and Loos.²⁸ Highest median per-capita discharges (ug person⁻¹ dav⁻¹) across compounds, in 159 160 decreasing order, were for PFOA (27), PFHxA (14), PFHpA (10), PFOS (9), PFHxS (7), and 161 PFNA (5) (Figure S4). These are lower than previously reported in Europe ca. 2007 (e.g., PFOA: 82 μ g person⁻¹ day⁻¹, PFOS: 57 μ g person⁻¹ day⁻¹).²⁸ 162

163 Source identification

Both hierarchical clustering and PCA identified three distinct groupings of PFASs (Figure 2a, b). The first component/cluster explains 46% of variability in the PCA and includes two major end products of the fluorochemical manufacturing industry (PFOA, PFNA), and a mix of other compounds: PFBS, PFHxS, PFHxA, PFDA. Site 5 (Mill Cove, RI) contains the highest summed PFASs across all sites and is dominated by this mixture of PFASs. PCA results suggest Site 5 is statistically similar to the Pawcatuck River, RI sampling locations (Sites 20, 19) and the

170	Passaic River, NJ (Site 31). However, these sites are grouped separately in the hierarchical
171	clustering analysis (Figure 2b), suggesting some differences in source contributions.
172	Geospatial analysis of the watersheds for Sites 5, 19, 20 and 31 reveals a mixture of
173	potential sources (Figure S5). For Site 5, the greatest source impact as a function of distance
174	within the watershed is from T.F. Green Airport, the largest public airport in Rhode Island. Prior
175	work indicates uses of AFFF at modern airports release diverse PFASs to downstream aquatic
176	environments, including the compounds identified as part of the first PCA/cluster. ^{4,29-31} For Sites
177	19 and 20, textile mills in the upstream watersheds have the highest impact as a function of
178	distance (Table S5). PFASs are used for water resistant coating in textiles and washing and
179	disposal of wastewater at textile mills provides a vector for their entry to the aquatic
180	environment. For Site 31, PCA scores suggest a mix of components 1-3 (Figure 2 c, d). This site
181	also clusters differently than Sites 19 and 20 (Figure 2b). The FRS database indicates the
182	watershed of Site 31 (Figure S5) contains diverse industrial sources that must account for this
183	profile including metal plating, printing, a landfill, petroleum and coal products manufacturing.
184	Overall, we conclude that the first PCA component and cluster of PFASs (PFOA, PFNA PFBS,
185	PFHxS, PFHxA, PFDA) represents a mixture of contemporary sources including airports and
186	textile mills.
187	The second component/cluster explains 19% of the variability in PFASs and includes two
188	long-chain PFASs (PFUnDA and PFDoDA) and two precursors to PFOS (N-MeFOSAA and N-
189	EtFOSAA) (Figure 2). PFUnDA and PFDoDA mainly originate from fluorotelomer alcohols or

190 other fluototelomer based products.³² Both N-MeFOSAA and N-EtFOSAA are intermediate

191 degradation products from the volatile parent compound N-alkyl

192 perfluorooctansulfonamidoethanol (FOSE) with PFOS as the final degradation product. This

193 profile is most pronounced at Site 3 along the Woonasquatucket River in RI and is also evident 194 at Site 1 (Slack's Tributary, RI) and Site 6 (Buckeye Brook, RI). For Site 3, the largest source 195 impact based on distance is from a wastewater treatment plant 1 km upstream. No industrial 196 facilities exist upstream of Sites 1 and 6. Landfill/waste management facilities are located within 197 2 km of all three sites but are not hydrologically connected to the sampling locations (Figure S5). 198 Both landfills and wastewater treatment plants are known atmospheric sources of fluorotelomer alcohols and FOSE.³³ Concentrations of N-MeFOSAA, PFUnDA and PFDoDA were not 199 200 spatially variable at most sites and only slightly elevated at Site 3, consistent with an atmospheric 201 input pathway. We thus infer that this component is most likely attributable to sources from the 202 waste sector.

203 The third component explains 15% of the variability in PFASs and includes PFPeA, 204 PFOS, and 6:2 FTS. This component is most pronounced at Site 2 along the Woonasquatucket 205 River, within the City of Providence, RI. GIS analysis of the watershed at this site reveals the 206 presence of 14 metal coating/plating industries upstream (Figure 2d, Table S5, Figure S5). PFOS 207 was historically used as a mist/fume control agent in metal plating, in surface coatings and as the major component in AFFFs for fighting petroleum related fire.^{25,26,34} Some PFOS applications 208 209 such as metal plating have been replaced by less stable fluorotelomer based chemicals such as 6:2 FtS,³⁵ which will eventually degrade into PFPeA and PFHxA (yields of 1.1% and 1.5% in 210 activated sludge).³⁶ It is likely that PFHxA is not included in the cluster because other direct 211 sources can contribute one order of magnitude more PFHxA than PFPeA.^{37,38} We conclude that 212 213 the distinct PFAS profile at Site 2 is can be explained by the metal plating industry.

214 Implications

215 Multivariate statistical tools such as PCA and hierarchical clustering of PFAS profiles 216 combined with data on hydrological proximity of potential sources are useful for identifying 217 sources of surface water contamination. We find aquatic transport pathways (hydrological 218 distance and river flow directions) are critical for source identification. This contrasts many other 219 persistent organic pollutants that are primarily transported atmospherically, allowing sources within a radius surrounding the sampling sites to be linked to concentrations.³⁹ We conclude that 220 221 the approach demonstrated here for RI and NY/NJ has potential for diagnosing PFAS source 222 contributions in urbanized regions with elevated concentrations and lacking specific information 223 on the magnitude of PFAS discharges from diverse industries. Background PFAS concentrations 224 at most rural sites in this study contain a mix of diverse source signatures that are not statistically 225 distinguishable using these methods. This analysis could be refined in future applications by 226 analyzing additional emerging short-chain PFASs and precursors to develop more unique 227 chemical signatures for specific industries (i.e., those contributing to the first component/cluster).

228 Supporting Information

Supporting Information Available: Details on analytical methods, data analyses, supporting
figures and tables. This material is available free of charge via the Internet at http://pubs.acs.org.

231 ACKNOWLEDGMENTS

We acknowledge financial support for this study from the Smith Family Foundation, the Harvard
John A. Paulson School of Engineering and Applied Sciences TomKat fund and the Harvard
NIEHS Center Grant (P30ES00002). X.Z. was partially supported by a Postdoctoral Fellowship
from the Natural Sciences and Engineering Research Council of Canada (PDF-437949-2013).
We thank Pete August (URI), Wenlu Zhao, Minggang Cai (Xiamen Univ.), Kirk Barrett

- 237 (Manhattan College) for assistance with sample collection and GIS data. We thank Linda Green
- and Elizabeth Herron, University of Rhode Island Watershed Watch program
- 239 (<u>http://web.uri.edu/watershedwatch</u>) leaders as well as the many URI Watershed Watch
- volunteer water quality monitors who collected water samples. RL acknowledges support from
- the Hudson River Foundation (HRF 2011-5) and SERDP (ER-2538).

242 References

243 Grandjean, P.; Andersen, E. W.; Budtz-Jorgensen, E.; Nielsen, F.; Molbak, K.; Weihe, P.; 1. 244 Heilmann, C., Serum vaccine antibody concentrations in children exposed to perfluorinated 245 compounds. Jama-J Am Med Assoc 2012, 307, 391-397.

- 246 2. Post, G. B.; Louis, J. B.; Lippincott, R. L.; Procopio, N. A., Occurrence of Perfluorinated
- 247 Compounds in Raw Water from New Jersey Public Drinking Water Systems. Environ. Sci.
- 248 Technol. 2013, 47, 13266-13275.
- 249 Castiglioni, S.; Valsecchi, S.; Polesello, S.; Rusconi, M.; Melis, M.; Palmiotto, M.; 3.
- 250 Manenti, A.; Davoli, E.; Zuccato, E., Sources and fate of perfluorinated compounds in the
- 251 aqueous environment and in drinking water of a highly urbanized and industrialized area in Italy. 252 J. Hazard. Mater. 2015, 282, 51-60.
- 253 de Solla, S. R.; De Silva, A. O.; Letcher, R. J., Highly elevated levels of perfluorooctane 4. 254 sulfonate and other perfluorinated acids found in biota and surface water downstream of an 255
- international airport, Hamilton, Ontario, Canada. Environ. Int. 2012, 39, 19-26.
- 256 Hurley, S.; Houtz, E. F.; Goldberg, D.; Wang, M.; Park, J.; Nelson, D. O.; Reynolds, P.; 5.
- 257 Bernstein, L.; Anton-Culver, H.; Horn-Ross, P., Preliminary Associations between the Detection 258 of Perfluoroalkyl Acids (PFAAs) in Drinking Water and Serum Concentrations in a Sample of
- 259 California Women. Environ. Sci. Technol. Letters 2016.
- 260 Happonen, M.; Koivusalo, H.; Malve, O.; Perkola, N.; Juntunen, J.; Huttula, T., 6.
- 261 Contamination risk of raw drinking water caused by PFOA sources along a river reach in south-262 western Finland. Sci. Total Environ. 2016, 541, 74-82.
- 263 7. Wang, Z. Y.; Cousins, I. T.; Scheringer, M.; Hungerbuhler, K., Fluorinated alternatives to 264 long-chain perfluoroalkyl carboxylic acids (PFCAs), perfluoroalkane sulfonic acids (PFSAs) and 265 their potential precursors. Environ. Int. 2013, 60, 242-248.
- US EPA, Facility Registry Service (FRS). http://www.epa.gov/enviro/epa-frs-facilities-266 8. state-single-file-csv-download (accessed Nov 2015). 267
- 268 9. U.S. EPA, Toxics Release Inventory (TRI) Program. https://http://www.epa.gov/toxics-269 release-inventory-tri-program (accessed May 2016).
- 270 Johnson, G. W.; Ehrlich, R.; Full, W.; Ramos, S., Principal components analysis and 10. 271 receptor models in environmental forensics. Academic Press, San Diego, CA: 2002.
- 272 Benskin, J. P.; Ahrens, L.; Muir, D. C. G.; Scott, B. F.; Spencer, C.; Rosenberg, B.; 11.
- 273 Tomy, G.; Kylin, H.; Lohmann, R.; Martin, J. W., Manufacturing Origin of Perfluorooctanoate
- 274 (PFOA) in Atlantic and Canadian Arctic Seawater. Environ. Sci. Technol. 2012, 46, 677-685.
- 275 Zhang, X.; Meyer, T.; Muir, D. C.; Teixeira, C.; Wang, X.; Wania, F., Atmospheric 12.
- 276 deposition of current use pesticides in the Arctic: snow core records from the Devon Island Ice 277 Cap, Nunavut, Canada. Environ Sci Process Impacts 2013, 15, 2304-11.
- 278 Westgate, J. N.; Wania, F., On the construction, comparison, and variability of airsheds 13.
- 279 for interpreting semivolatile organic compounds in passively sampled air. Environ. Sci. Technol. 280 2011, 45, 8850-7.
- 281 14. Taniyasu, S.; Kannan, K.; So, M. K.; Gulkowska, A.; Sinclair, E.; Okazawa, T.;
- 282 Yamashita, N., Analysis of fluorotelomer alcohols, fluorotelorner acids, and short- and long-
- 283 chain perfluorinated acids in water and biota. J. Chromatogr. A 2005, 1093, 89-97.
- 284 D'eon, J. C.; Crozier, P. W.; Furdui, V. I.; Reiner, E. J.; Libelo, E. L.; Mabury, S. A., 15.
- 285 Perfluorinated Phosphonic Acids in Canadian Surface Waters and Wastewater Treatment Plant
- 286 Effluent: Discovery of a New Class of Perfluorinated Acids. Environ. Toxicol. Chem. 2009, 28,

- 287 2101-2107.
- 288 16. Helsel, D. R., Nondetects and data analysis : statistics for censored environmental data.
- 289 Wiley-Interscience: Hoboken, N.J., 2005; p xv, 250 p.
- 290 17. ESRI Demographics (2012) Population density in the United States.
- 291 <u>http://www.esri.com/data/esri_data</u> (accessed May 2016).
- 292 18. U.S. Geological Survey StreamStats (version 4). <u>http://ssdev.cr.usgs.gov/streamstats/</u>
- 293 (accessed May 2016).
- 294 19. Sakurai, T.; Serizawa, S.; Kobayashi, J.; Kodama, K.; Lee, J. H.; Maki, H.; Zushi, Y.;
- 295 Sevilla-Nastor, J. B.; Imaizumi, Y.; Suzuki, N.; Horiguchi, T.; Shiraishi, H., Temporal trends for
- 296 inflow of perfluorooctanesulfonate (PFOS) and perfluorooctanoate (PFOA) to Tokyo Bay, Japan,
- estimated by a receptor-oriented approach. Sci. Total Environ. 2016, 539, 277-85.
- 298 20. Hansen, K. J.; Johnson, H. O.; Eldridge, J. S.; Butenhoff, J. L.; Dick, L. A., Quantitative
- characterization of trace levels of PFOS and PFOA in the Tennessee River. *Environ. Sci. Technol.* 2002, *36*, 1681-1685.
- 301 21. Nakayama, S.; Strynar, M. J.; Helfant, L.; Egeghy, P.; Ye, X. B.; Lindstrom, A. B.,
- Perfluorinated compounds in the Cape Fear Drainage Basin in North Carolina. *Environ. Sci. Technol.* 2007, *41*, 5271-5276.
- 304 22. Konwick, B. J.; Tomy, G. T.; Ismail, N.; Peterson, J. T.; Fauver, R. J.; Higginbotham, D.;
- Fisk, A. T., Concentrations and patterns of perfluoroalkyl acids in Georgia, USA surface waters near and distant to a major use source. *Environ. Toxicol. Chem.* **2008**, *27*, 2011-2018.
- 307 23. Nakayama, S. F.; Strynar, M. J.; Reiner, J. L.; Delinsky, A. D.; Lindstrom, A. B.,
- Determination of Perfluorinated Compounds in the Upper Mississippi River Basin. *Environ. Sci. Technol.* 2010, 44, 4103-4109.
- 24. Lasier, P. J.; Washington, J. W.; Hassan, S. M.; Jenkins, T. M., Perfluorinated chemicals
- in surface waters and sediments from northwest Georgia, USA, and their bioaccumulation in Lumbriculus variegatus. *Environ. Toxicol. Chem.* **2011**, *30*, 2194-201.
- 25. Paul, A. G.; Jones, K. C.; Sweetman, A. J., A first global production, emission, and environmental inventory for perfluorooctane sulfonate. *Environ. Sci. Technol.* **2009**, *43*, 386-392.
- 316 26. Armitage, J. M.; Schenker, U.; Scheringer, M.; Martin, J. W.; MacLeod, M.; Cousins, I.
- 317 T., Modeling the Global Fate and Transport of Perfluorooctane Sulfonate (PFOS) and Precursor
- 318 Compounds in Relation to Temporal Trends in Wildlife Exposure. *Environ. Sci. Technol.* 2009,
- *43*, 9274-9280.
- 320 27. US Census Bureau (1995) Urban And Rural Definitions.
- 321 <u>http://www.census.gov/population/censusdata/urdef.txt</u> (accessed June 2016).
- Pistocchi, A.; Loos, R., A Map of European Emissions and Concentrations of PFOS and
 PFOA. *Environ. Sci. Technol.* 2009, *43*, 9237-9244.
- 324 29. Ahrens, L.; Norström, K.; Viktor, T.; Cousins, A. P.; Josefsson, S., Stockholm Arlanda
- 325 Airport as a source of per- and polyfluoroalkyl substances to water, sediment and fish.
- 326 *Chemosphere* **2015**, *129*, 33-38.
- 327 30. Awad, E.; Zhang, X. M.; Bhavsar, S. P.; Petro, S.; Crozier, P. W.; Reiner, E. J.; Fletcher,
- 328 R.; Tittemier, S. A.; Braekevelt, E., Long-Term Environmental Fate of Perfluorinated
- 329 Compounds after Accidental Release at Toronto Airport. *Environ. Sci. Technol.* 2011, 45, 8081330 8089.
- 331 31. Filipovic, M.; Woldegiorgis, A.; Norstrom, K.; Bibi, M.; Lindberg, M.; Osteras, A. H.,
- 332 Historical usage of aqueous film forming foam: A case study of the widespread distribution of

- perfluoroalkyl acids from a military airport to groundwater, lakes, soils and fish. *Chemosphere*
- **2015**, *129*, 39-45.
- 335 32. Government of Canada, Chemical Substances: Long-Chain (C9-C20)
- 336 Perfluorocarboxylic Acids (PFCAs), Their Salts, and Their Precursors
- 337 <u>http://www.chemicalsubstanceschimiques.gc.ca/challenge-defi/summary-sommaire/pfcaps-</u>
- 338 <u>apfcsp-eng.php</u> (accessed Nov 2015).
- 339 33. Ahrens, L.; Shoeib, M.; Harner, T.; Lee, S. C.; Guo, R.; Reiner, E. J., Wastewater
- 340 Treatment Plant and Landfills as Sources of Polyfluoroalkyl Compounds to the Atmosphere.
- 341 Environ. Sci. Technol. 2011, 45, 8098-8105.
- 342 34. Moody, C. A.; Field, J. A., Perfluorinated surfactants and the environmental implications
 343 of their use in fire-fighting foams. *Environ. Sci. Technol.* 2000, *34*, 3864-3870.
- 344 35. Persistent Organic Pollutants Review Committee of UNEP (2012) Technical paper on the
- 345 identification and assessment of alternatives to the use of perfluorooctane sulfonic acid, its salts,
- 346 perfluorooctane sulfonyl fluoride and their related chemicals in open applications
- 347 (UNEP/POPS/POPRC.8/INF/17/Rev.1).
- 348 36. Wang, N.; Szostek, B.; Buck, R. C.; Folsom, P. W.; Sulecki, L. M.; Gannon, J. T., 8-2
- Fluorotelomer alcohol aerobic soil biodegradation: Pathways, metabolites, and metabolite yields.
 Chemosphere 2009, 75, 1089-1096.
- 351 37. Dauchy, X.; Boiteux, V.; Rosin, C.; Munoz, J. F., Relationship Between Industrial
- 352 Discharges and Contamination of Raw Water Resources by Perfluorinated Compounds: Part II:
- 353 Case Study of a Fluorotelomer Polymer Manufacturing Plant. Bull. Environ. Contam. Toxicol.
- **2012**, *89*, 531-536.
- 355 38. Butenhoff, J. L.; Kennedy, G. L., Jr.; Frame, S. R.; O'Connor, J. C.; York, R. G., The 356 reproductive toxicology of ammonium perfluorooctanoate (APFO) in the rat. *Toxicology* **2004**, 357 *196*, 95-116.
- 358 39. Khairy, M.; Muir, D.; Teixeira, C.; Lohmann, R., Spatial Distribution, Air-Water
- 359 Fugacity Ratios and Source Apportionment of Polychlorinated Biphenyls in the Lower Great
- 360 Lakes Basin. Environ. Sci. Technol. 2015, 49, 13787-97.
- 361

- 362 Figure Captions
- **Figure 1**. Concentrations of PFASs measured in surface waters from Rhode Island and the New
- 364 York Metropolitan Area. Full names of individual compounds are listed in Table S2. N-
- 365 MeFOSAA and N-EtFOSAA are not shown but were detected in ~70% of the samples at

366 concentrations <1 ng/L.

- 368 Figure 2. Multivariate statistical analysis of surface water data. Panel (A) shows loadings of
- 369 principal components analysis (PCA) and Panels (C) and (D) show score plots for three
- 370 components across sampling sites. Panel (B) compares PCA results to hierarchical clustering of
- 371 compounds and sites. Sites with statistically distinct PFAS profiles are indicated on plots (C)
- and (D) and highlighted on the hierarchical clustering diagram. The three principal components
- together explain 80% of the variance in PFAS composition.

374 Figure 1



