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Accessibility
Detection of Poly- and Perfluoroalkyl Substances (PFASs) in U.S. Drinking Water Linked to Industrial Sites, Military Fire Training Areas, and Wastewater Treatment Plants

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ABSTRACT: Drinking water contamination with poly- and perfluoroalkyl substances (PFASs) poses risks to the developmental, immune, metabolic, and endocrine health of consumers. We present a spatial analysis of 2013–2015 national drinking water PFAS concentrations from the U.S. Environmental Protection Agency’s (US EPA) third Unregulated Contaminant Monitoring Rule (UCMR3) program. The number of industrial sites that manufacture or use these compounds, the number of military fire training areas, and the number of wastewater treatment plants are all significant predictors of PFAS detection frequencies and concentrations in public water supplies. Among samples with detectable PFAS levels, each additional military site within a watershed’s eight-digit hydrologic unit is associated with a 20% increase in PFHxS, a 10% increase in both PFHpA and PFOA, and a 35% increase in PFOS. The number of civilian airports with personnel trained in the use of aqueous film-forming foams is significantly associated with the detection of PFASs above the minimal reporting level. We find drinking water supplies for 6 million U.S. residents exceed US EPA’s lifetime health advisory (70 ng/L) for PFOS and PFOA. Lower analytical reporting limits and additional sampling of smaller utilities serving <10000 individuals and private wells would greatly assist in further identifying PFAS contamination sources.

INTRODUCTION

Poly- and perfluoroalkyl substances (PFASs) make up a large group of persistent anthropogenic chemicals used in industrial processes and commercial products over the past 60 years. Widespread use and extreme resistance to degradation have resulted in the ubiquitous presence of these compounds in the environment. The 2011–2012 U.S. National Health and Nutrition Examination Survey reported detectable serum PFAS concentrations in virtually all individuals (97%). Human PFAS exposure has been linked to cancer, elevated cholesterol, obesity, immune suppression, and endocrine disruption. Health concerns in the early 2000s prompted manufacturers in Europe and North America to phase out production of some long-chain PFASs. Declines in
production of these compounds have been offset by increases in developing regions such as Asia. Limited available data suggest widespread exposure to replacement (short-chain) PFASs may also adversely affect human health.

Human PFAS exposure includes dietary sources, household dust, air, and drinking water. Exposure from drinking water is a serious concern because of the high aqueous solubility of many PFASs. Relatively low PFAS concentrations can lead to elevated exposures in the general population. Elevated PFAS concentrations in U.S. drinking water have been reported in numerous regions, especially near industrial sites that produce or use them. For example, perfluorooctanoic acid (PFOA) concentrations 190-fold higher than the lifetime health advisory (70 ng/L) recommended by the U.S. Environmental Protection Agency (US EPA) were measured in drinking water near a fluorochemical facility in Washington, WV, where PFOA was used in fluoropolymer production.

Many civilian airports and military fire training areas have been contaminated by PFASs contained in aqueous film-forming foams (AFFFs) that are widely used during firefighting training activities. Groundwater and surface waters surrounding these sites containing PFAS concentrations that are 3–4 orders of magnitude higher than the US EPA health advisory level for drinking water have been reported. Wastewater treatment plants (WWTPs) are another important PFAS source because these compounds are not removed by standard treatment methods, and labile precursors biodegrade, increasing concentrations in effluent relative to influent. Land application of approximately half of the biosolids generated by WWTPs may contribute to human exposure through subsequent contamination of water, food, livestock, and wildlife.

Understanding nationwide PFAS exposures from drinking water is important for identifying potentially vulnerable populations. However, previous studies have mainly focused on individual point sources of PFAS contamination and site-specific drinking water exposures. Here we develop a statistical framework for investigating whether increased PFAS concentrations in drinking water are associated with the number of point sources within a watershed (hereafter abbreviated HUC). We used publicly available drinking water concentration data for six PFASs from the US EPA’s third Unregulated Contaminant Monitoring Rule (UCMR3), including perfluorobutanesulfonic acid (PFBS), perfluorohexanesulfonic acid (PFHxS), perfluoroheptanoic acid (PFHpA), PFOA, perfluorooctanesulfonic acid (PFOS), and perfluorononanoic acid (PFNA) (Table S1). We discuss the utility of the UCMR3 database for identifying sources of PFASs to U.S. drinking water supplies, locations of vulnerable populations, and priorities for future monitoring.

**METHODS**

**Drinking Water Data.** Our analysis included analytical results for six PFASs in 36149 drinking water samples from the US EPA’s UCMR3 program collected between January 2, 2013, and December 9, 2015. Samples cover all 4064 public water supplies serving >10000 individuals. Data are also available for 800 public water supplies serving <10000 individuals, but this represents only a small fraction (0.5%) of the 144165 in this category. Minimum reporting levels (MRLs) for the six PFASs analyzed are listed in Table S1.

One limitation of the UCMR3 database is that national data on system intakes for public water supplies are classified, making it difficult to place them within a specific hydrological network. We therefore extracted the zip codes for areas served and aggregated data within eight-digit HUCs to capture the most detailed hydrologic information that exceeds the spatial resolution of PFAS data (zip code areas). We used the highest reported PFAS concentrations when multiple systems were located within a single zip code and/or when multiple zip code areas were located within the same HUC.

**PFAS Point Sources.** Our spatial analysis (Figure S1) included point source information for (a) 16 industrial sites listed in the US EPA’s 2010/2015 PFOA Stewardship Program (Table S2), (b) 8572 WWTPs, (c) 290 military fire training areas that contain 664 military fire training sites, and (d) 533 civilian airports that are compliant with Title 14 Code of Federal Regulations, Part 139 for personnel trained in the use of AFFF (hereafter termed “AFFF-certified airports”). PFASs produced and/or used vary across industrial sites, and not all compounds were associated with all sites. For example, a fluorochemical manufacturing facility in Decatur, AL, produced both PFOS and PFOA, while only PFOA was used in the manufacturing process of another fluorochemical production facility in Parkersburg, WV. We conducted a sensitivity analysis to examine the potential production misclassification bias by limiting industrial sites to include the ones that only produced or used each specific compound (Table S3). We used the Google Maps application program interface (API) to geocode coordinates based on addresses. Potentially important PFAS sources such as landfills, biosolids, and small industrial PFAS users could not be included in this analysis because comprehensive geospatial data are not available.

**Spatial and Statistical Analysis.** We used ArcMap 10.3.1 (ESRI) to explore statistical differences between the number of point sources in eight-digit HUCs with PFAS levels above and below the level of detection. We developed a multivariate spatial regression model for watersheds with detectable PFASs that adjusts for correlations and co-location among point sources. A natural log transformation was used to normalize the distribution of individual PFASs. PFNA and PFBS were excluded from the spatial regression analysis due to a low detection frequency (15 and 14 of 1601 watersheds, respectively). We used Moran’s I statistic to test for spatial dependence in the model residuals from an ordinary least-squares (OLS) regression and correct for spatial dependence in the final spatial regression model. Akaike Information Criterion was used to compare the OLS and spatial regression models, where a lower value implies a better model fit. A series of cross-validation tests were also completed to assess the predictive capacity and stability of the final set of models. The OLS and spatial regression models were constructed using GeoDa 1.6 software, and cross-validation was implemented in R version 3.1.3.

**RESULTS AND DISCUSSION**

**PFASs in U.S. Drinking Water.** PFASs were detected at or above the MRLs in 194 of 4864 public water supplies, serving 16.5 million residents in 33 different states, three American territories (American Samoa, Northern Mariana Islands, and Guam), and the Salt River Pima-Maricopa Indian Community. Drinking water from 13 states accounted for 75% of detections, including, by order of frequency of detection, California, New Jersey, North Carolina, Alabama, Florida, Pennsylvania, Ohio,
New York, Georgia, Minnesota, Arizona, Massachusetts, and Illinois (Figure 1). Detection frequencies for PFASs across the 4864 public water supplies were 2.2% for PFOA, 2.0% for PFOS, 1.7% for PFHpA, 1.1% for PFHxS, and <0.003% for others.

![Hydrologic unit codes (eight-digit HUCs) used as a proxy for watersheds with detectable PFOA and PFOS in drinking water measured in the US EPA’s UCMR3 program (2013–2015). Blank areas represent regions where no data are available.](Image)

Figure 1. Hydrologic unit codes (eight-digit HUCs) used as a proxy for watersheds with detectable PFOA and PFOS in drinking water measured in the US EPA’s UCMR3 program (2013–2015). Blank areas represent regions where no data are available.

Many detectable PFAS concentrations in the UCMR3 database are above chronic drinking water and water quality standards for other regions (i.e., surface water European Union, PFOS, <1 ng/L; drinking water Sweden, sum of seven PFASs, <90 ng/L; groundwater State of New Jersey, PFNA, <10 ng/L; drinking water State of Vermont, sum of PFOS and PFOA, <20 ng/L). A recent analysis developed a benchmark dose for immunotoxicity in children and suggested a drinking water limit of approximately 1 ng/L for PFOS and PFOS. Data from rodents that measured sensitive end points such as mammary gland development support a similar level.

Six million people were served by 66 public water supplies that have at least one sample at or above the US EPA’s 2016 health advisory for PFOS and PFOA (70 ng/L individually or combined). Concentrations ranged as high as 349 ng/L for PFOA, 1800 ng/L for PFOS, and 56 ng/L for PFNA.

The detection frequency in drinking water sourced from groundwater was more than twice that from surface water (Table S4). Long-chain PFASs (PFHxS, PFOS, PFOA, and PFNA) were more frequently detected in groundwater, and short-chain compounds (PFHpA and PFBS) were detected more frequently in surface waters. This may be due to both the original mode of environmental release (as an aerosol, application to soil, and aqueous discharge) and the inverse relationship between PFAS mobility and chain length. The MRLs (10–90 ng/L) in the UCMR3 database are up to 2 orders of magnitude higher than the limit of quantitation in most published studies, and more than 10 times higher than the drinking water limit (1 ng/L) suggested by human and animal studies. Because PFASs are detectable in virtually all parts of the environment, we infer that the large fraction of samples below reporting limits (Table S4) is driven in part by high MRLs.

### Sources Surrounding Locations with Detectable PFASs

Our analysis indicates point sources are significantly more abundant in HUCs with detectable PFASs [two-sided t test, p < 0.05 (Table 1 and Figure S2)]. This includes drinking water samples from 1601 of the 2158 total U.S. HUCs. For example, HUCs with detectable PFOA levels (8% of the total) have more industrial sites, military fire training areas, AFFF-certified airports, and WWTPs than those with concentrations below detection. These trends can be observed across all.

### Table 1. Mean Abundance of Point Sources within Eight-Digit Hydrologic Unit Codes (HUCs) with Drinking Water PFAS Concentrations above and below the Method Reporting Limit in the UCMR3 Program

<table>
<thead>
<tr>
<th>compound</th>
<th>major industrial sites</th>
<th>military fire training areas</th>
<th>AFFF-certified airports</th>
<th>WWTPs</th>
</tr>
</thead>
<tbody>
<tr>
<td>PFBS</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>&lt;90 ng/L</td>
<td>(n = 1587)</td>
<td>0.01</td>
<td>0.15</td>
<td>0.29</td>
</tr>
<tr>
<td>&gt;90 ng/L</td>
<td>(n = 14)</td>
<td>0.21</td>
<td>0.71</td>
<td>0.50</td>
</tr>
<tr>
<td>p-value</td>
<td>&lt;0.05</td>
<td>0.0206</td>
<td>0.105</td>
<td>0.148</td>
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<tr>
<td>PFHxS</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>&lt;30 ng/L</td>
<td>(n = 1507)</td>
<td>0.01</td>
<td>0.13</td>
<td>0.27</td>
</tr>
<tr>
<td>&gt;30 ng/L</td>
<td>(n = 94)</td>
<td>0.06</td>
<td>0.60</td>
<td>0.63</td>
</tr>
<tr>
<td>p-value</td>
<td>&lt;0.001</td>
<td>0.056</td>
<td>&lt;0.001</td>
<td>&lt;0.001</td>
</tr>
<tr>
<td>PFHpA</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>&lt;10 ng/L</td>
<td>(n = 1509)</td>
<td>0.01</td>
<td>0.13</td>
<td>0.26</td>
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<tr>
<td>&gt;10 ng/L</td>
<td>(n = 92)</td>
<td>0.09</td>
<td>0.57</td>
<td>0.67</td>
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<tr>
<td>p-value</td>
<td>&lt;0.001</td>
<td>0.016</td>
<td>&lt;0.001</td>
<td>&lt;0.001</td>
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<tr>
<td>PFOA</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>&lt;20 ng/L</td>
<td>(n = 1473)</td>
<td>0.01</td>
<td>0.13</td>
<td>0.26</td>
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<tr>
<td>&gt;20 ng/L</td>
<td>(n = 128)</td>
<td>0.05</td>
<td>0.52</td>
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<tr>
<td>p-value</td>
<td>&lt;0.001</td>
<td>0.038</td>
<td>&lt;0.001</td>
<td>&lt;0.001</td>
</tr>
<tr>
<td>PFOS</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>&lt;40 ng/L</td>
<td>(n = 1487)</td>
<td>0.01</td>
<td>0.13</td>
<td>0.26</td>
</tr>
<tr>
<td>&gt;40 ng/L</td>
<td>(n = 114)</td>
<td>0.05</td>
<td>0.54</td>
<td>0.57</td>
</tr>
<tr>
<td>p-value</td>
<td>&lt;0.001</td>
<td>0.064</td>
<td>&lt;0.001</td>
<td>&lt;0.001</td>
</tr>
<tr>
<td>PFNA</td>
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<td></td>
<td></td>
</tr>
<tr>
<td>&lt;20 ng/L</td>
<td>(n = 1586)</td>
<td>0.01</td>
<td>0.15</td>
<td>0.28</td>
</tr>
<tr>
<td>&gt;20 ng/L</td>
<td>(n = 15)</td>
<td>0.13</td>
<td>1.13</td>
<td>1.13</td>
</tr>
<tr>
<td>p-value</td>
<td>&lt;0.001</td>
<td>0.366</td>
<td>0.014</td>
<td>0.008</td>
</tr>
</tbody>
</table>

The mean abundance is calculated as the mean number of point sources within HUCs with PFASs above or below the level of detection. Only the major industrial sites participating in the US EPA’s 2010/2015 PFOA Stewardship Program were included. Wastewater treatment plant. Two-sample t-test p-values.
indicate increasing concentrations with an increasing abundance of point sources within the same hydrologic unit. PFOA, PFHxS, and PFHpA, and is statistically significant for the spatial error term characterizing spatial influence. Results have been transformed to reflect expected changes in drinking water concentrations per increase in the abundance of different sources. Positive coefficients in the results indicate increasing concentrations with an increasing abundance of point sources within the same hydrologic unit. p-values for the spatial error regression model. The spatial error term is used to incorporate spatial autocorrelation structures into a linear regression model.

PFAFs. Similarly, HUCs with point sources have higher detection frequencies for PFAFs (Table S5). For example, the presence of a military fire training area within a HUC increases the frequency of detection of at least one PFAS from 10.4% to 28.2%. One caveat is that imprecise information about public water supply intakes can cause misclassification bias. Systems that draw water upstream from point sources, such as Minneapolis and St. Paul in Minnesota, may not actually be affected as indicated by the aggregated spatial analysis.

Results of the Spatial Regression Model. Spatial regression modeling explains 38–62% of the variance in drinking water concentrations for the four PFASs considered (Table 2). Each additional industrial site within a HUC is associated with an 81% increase in PFOA (p < 0.001), which is the strongest statistical association across compounds and point sources. Increasing PFOS concentrations are positively associated with the number of industrial sites, but this relationship is not statistically significant (p = 0.124). The small number of sites that have manufactured or used PFOS likely accounts for the lack of a statistically significant relationship.

The number of military fire training areas within each HUC is positively associated with increasing levels of all PFOS, PFOA, PFHxS, and PFHpA, and is statistically significant for PFHxS (p = 0.045) and PFOS (p = 0.007). Each additional military fire training area within the same HUC is associated with a 20% increase in PFHxS (p = 0.002), a 10% increase in PFHpA (p = 0.155), a 10% increase in PFOA (p = 0.111), and a 35% increase in PFOS (p < 0.001). AFFFs typically contain relatively high concentrations of PFOS and PFHxS and their polyfluorinated precursors compared to the concentrations of other perfluorinated carboxylates, which is consistent with these statistical results.

We find a small but significant increase in PFOS and PFOA (2%; p < 0.01) with each additional WWTP within the same HUC. This is consistent with the greater abundance but smaller quantities of PFASs released by WWTPs. Similarly, results of Valsecchi et al. show PFAS releases from WWTPs are important but less significant than those from fluorochemical manufacturing facilities in Italy. The number of WWTPs may also be a proxy for other population-driven PFAS sources. The number of AFFF-certified airports is not significantly associated with PFAS concentrations in the current data set. This may reflect misclassification bias because the certification used to identify airports indicates eligibility but not actual use of AFFF. The UCMR3 database contains limited data for smaller drinking water systems where localized reports of contamination from airports have been most abundant.

Current Data Limitations and Future Monitoring Efforts. The UCMR3 database has several limitations that restrict its predictive power for identifying U.S. drinking water supplies likely to contain elevated levels of PFASs. Classification of geospatial data on intakes for public water supplies limits the spatial resolution of the current data set and associated statistical models to a radius of 50 km (median radius of watersheds). Many of the impacted drinking water systems are groundwater systems, and contaminated groundwater plumes are often much smaller than 50 km. Geospatial data are lacking for many potentially important PFAS point sources such as a wide range of industries, landfills, biosolids application, and other AFFF-impacted sites where relatively smaller volumes of AFFF were released.

Data on PFAS releases from smaller industrial facilities (e.g., plastics, textiles, paper, and lubricants) are usually withheld as confidential business information, and little information about airborne emissions is available for characterizing the importance of atmospheric releases and potential long-range transport. For example, biosolids application resulted in one of the largest PFAS drinking water contamination events in Europe but could not be included in this analysis because U.S. use data are not available on a national scale.

Sources not included in our spatial analysis are represented by the highly significant lambda (λ) coefficients (Table 2). Areas with high model residuals (greater than 1.5 standard deviation) mean that current information about sources cannot fully explain the high observed PFAS concentrations. The map of model residuals (Figure S3) can thus be used to guide high-priority sampling regions in future work.

We found a statistically greater abundance of point sources in watersheds with detectable PFASs, including AFFF-certified airports. However, multivariate spatial regression models did not show a significant association between AFFF-certified airports and concentrations of PFASs in nearby drinking water.

<table>
<thead>
<tr>
<th>Table 2. Spatial Regression Models for Drinking Water PFAS Concentrations as a Function of the Abundance of Point Sources</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>compound</strong></td>
</tr>
<tr>
<td>----------------</td>
</tr>
<tr>
<td>PFHxS</td>
</tr>
<tr>
<td></td>
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<tr>
<td>PFHpA</td>
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<tr>
<td></td>
</tr>
<tr>
<td>PFOA</td>
</tr>
<tr>
<td></td>
</tr>
<tr>
<td>PFOS</td>
</tr>
<tr>
<td></td>
</tr>
</tbody>
</table>

Only the major industrial sites participating in US EPA’s 2010/2015 PFOA Stewardship Program were included. MFTA = military fire training area. WWTP = wastewater treatment plant. Coefficient for the spatial error term characterizing spatial influence. Results have been transformed to reflect expected changes in drinking water concentrations per increase in the abundance of different sources. Positive coefficients in the results indicate increasing concentrations with an increasing abundance of point sources within the same hydrologic unit. p-values for the spatial error regression model. The spatial error term is used to incorporate spatial autocorrelation structures into a linear regression model.
Other studies have reported elevated PFAS concentrations in groundwater wells adjacent to AFFF-certified airports. Small drinking water systems and private wells may be disproportionately affected by PFAs originating from AFFF use at civilian airports, but representative data for these small drinking water systems are not included in the UCMR3 program. Approximately 44.5 million U.S. individuals rely on private drinking water wells, and 52 million individuals rely on smaller public water supplies (<10000 served). The UCMR3 program includes 0.5% testing incidence for smaller public water supplies and no testing of private wells, meaning that information about drinking water PFAS exposures is therefore lacking for almost one-third of the U.S. population.

**ASSOCIATED CONTENT**

Supporting Information
The Supporting Information is available free of charge on the ACS Publications website at DOI: 10.1021/acs.estlett.6b00260.

Additional tables and figures (PDF)

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NOTE ADDED AFTER ASAP PUBLICATION

There was a text error in the Results and Discussion section in the version published ASAP August 9, 2016; the corrected version was published ASAP August 11, 2016.