Additive Models Reveal Sources of Metals and Organic Pollutants in Norwegian Marine Sediments

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ABSTRACT: We characterized spatial patterns of surface sediment concentrations of seven polychlorinated biphenyls (PCBs), seven polycyclic aromatic hydrocarbons (PAHs), three chlorinated pesticides, and five metals in Norwegian waters and Skagerrak. In total, we analyzed 5036 concentrations of 22 chemical substances that were measured between 1986 and 2014 at 333 sampling sites by means of generalized additive models (GAMs). We found that GAMs with organic carbon content of the sediment and latitude and longitude as co-variates explained as ca. 75% of the variability of the contaminant sediment concentrations. For metals, a predominantly hotspot-driven spatial pattern was found, i.e., we identified historical pollution hotspots (e.g., Sørfjord in western Norway) for mercury, zinc, cadmium, and lead. Highest concentrations of PAHs and PCBs were found close to densely populated and industrialized regions, i.e., in the North Sea and in the Kattegat and Skagerrak. The spatial pattern of the PCBs suggests the secondary and diffusion nature of their sources. Atmospheric inputs are the main sources of pollution for most organic chemicals considered, but north of the Arctic circle, we found that concentrations of PAHs increased from south to north most likely related to a combination of coal-eroding bedrock and the biological pump. The knowledge acquired in the present research is essential for developing effective remediation strategies that are consistent with international conventions on pollution control.

1. INTRODUCTION

In the 1970s, it became evident that anthropogenic chemicals can remain in the environment for long periods and pose toxic risk to humans and wildlife. As a result, multiple international conventions and management programmes were initiated, such as the European Water Framework Directive (Directive 2000/60/EC),2 the Marine Strategy Framework Directive (Directive 2008/56/EC),3 the Oslo-Paris Convention for the Protection of the Marine Environment of the North-East Atlantic (OSPAR)4 and the Helsinki Convention (HELCOM) that pertains to the Baltic Sea. These conventions aim to prevent pollution by continuously reducing discharges, emissions and losses of hazardous chemicals to the environment with the ultimate goal of achieving concentrations in the marine environment that are near background values for naturally occurring chemicals and are close to zero for man-made synthetic chemicals.1,4 The OSPAR convention provides a framework for frequent and obligatory monitoring to assess the inputs, levels, trends and effects of prioritized contaminants such as polychlorinated biphenyls (PCBs), polycyclic aromatic hydrocarbons (PAHs), pesticides and metals. The results of these monitoring campaigns over time indicate that most of the primary sources and discharges of these priority chemicals have been discontinued or considerably reduced.5,6 However, because many chemicals are persistent and strongly particle-associated, contaminated sediments may still act as important secondary pollution sources mainly in coastal waters.8 After decades of primary emissions, considerable reservoirs of anthropogenic chemicals have accumulated in sediments. These reservoirs can be remobilized due to climate-induced changes in temperature and organic matter content or physical disturbances due to sand or gravel extractions and dredging activities.10 The latter means that, especially in coastal waters, sediments have shifted from being a sink of pollution to being a potential source of pollution.11,12 The terrestrial environment

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The Norwegian marine environment receives input of hazardous chemicals due to on- and offshore point sources and diffuse pollution. Several sites along the Norwegian coast have previously been identified as highly contaminated with organic contaminants and metals. Some of these chemicals are known for their toxic potential and may cause adverse ecological effects. Prior to defining the most appropriate remediation strategy, and in the scope of the above-mentioned conventions and management plans, an important first step is to analyze and quantify the pollution levels, preferably with complete databases integrating data from different Norwegian institutes covering the Norwegian marine waters and the Skagerrak region. Based on these data, we assessed the spatial patterns of the contaminant concentrations and aimed to relate these patterns to the main sources of pollution, i.e., natural sources, local industry and mining, or diffuse long-range transport.

2. MATERIALS AND METHODS

Data. The Norwegian Institute for Water Research (NIVA), the Institute of Marine Research (IMR), and the Geological Survey of Norway (NGU) have collected a long series of comprehensive marine biological, oceanographic, and contaminants field data. Concentrations of anthropogenic chemicals in Norwegian coastal and marine sediments were obtained from the Marine AREA database for Norwegian waters (MAREANO; Barents Sea and Norwegian Sea) mapping program and the Contaminants in Coastal Waters of Norway monitoring program (MILKYS; North Sea and other coastal waters), which is part of the Norwegian contribution to OSPAR’s Coordinated Environmental Monitoring Programme (CEMP).

The main advantage of using field data for sediment (as opposed to sampling the water column) is that concentrations found in this compartment represent exposure over time (e.g., weeks, months, or years) as the sediment integrates the historic pollution signal. In total, 30 488 concentrations at 420 sampling locations representing 22 chemicals were extracted from the databases. Based on the reported latitude and longitude, each sampling site was attributed to a geographical category (i.e., inner fjord, coastal, northern offshore, and southern offshore). Samples taken in fjords and bays were categorized as "inner fjord". Samples taken at open sea but close to the mainland were categorized as coastal. All other samples were classified as offshore; those above 66 °N were classified as northern offshore, and those below 66 °N were classified as southern offshore (Figure S1). Note that, except for a small number of sampling locations, all data have been collected on the Norwegian Continental Shelf (Figure S2–S4). For data exploration, the sediment concentrations were plotted against the above-mentioned geographical categories for each chemical. We used a Kruskal–Wallis statistical test and a posthoc Mann–Whitney U test to identify pairwise differences of the chemical concentrations between the four geographical categories. Chemicals included in the analysis are polychlorinated biphenyl 28 (PCB28), PCB52, PCB101, PCB118, PCB138, PCB153, PCB180, pp′-dichlorodiphenyltrichloroethylene (DDT), hexachlorobenzene (HCB), α-hexachlorocyclohexane (α-HCH), anthracene (ANT), benz[a]anthracene (BAA), benzo[a]pyrene (BAP), naphthalene (NAP), fluoranthene (FLU), phenanthrene (PHE), pyrene (PYR), mercury (Hg), cadmium (Cd), zinc (Zn), lead (Pb), nickel (Ni), and total organic carbon (TOC).

Prior to the modeling, three preprocessing steps were performed. First, we only selected samples from the sediment surface layer (0–1 cm) of the sediment fraction smaller than 2000 μm grain size. Next, if one site was sampled multiple times in the same year, we calculated the average concentration. For example, at sampling location 10S in Varangerfjorden (69°56′06″ N; 30°06′42″ E) concentrations of mercury have been quantified three times in the top layer of the sediment (i.e., 0.053; 0.043; and 0.052 mg kg⁻¹ dry weight) in 1994. Hence, we calculated the average contaminant concentration (0.049 mg kg⁻¹ dry weight) and used this in the further analysis. The selection of the sediment surface layer combined with the year-round averaging resulting in a data set composed of 5036 records representing yearly average sediment concentrations measured between 1986 and 2014 (Table S1) at 333 sampling sites (Figure S2–S4). Furthermore, in line with the approach of Borge et al. (2011), if the reported concentrations were below the limit of quantification, we replaced the concentration (e.g., < 0.05 μg kg⁻¹) by a random value. This random value was selected from a uniform distribution characterized by half of the limit of quantification (minimum value; e.g., 0.025 μg kg⁻¹) and the limit of quantification (maximum value; e.g., 0.05 μg kg⁻¹). A chemical-specific summary of the data set is available in Table S1.

Model. The overall objective of the present research was to infer spatial patterns of anthropogenic chemicals in marine sediments in Norwegian waters and to relate these patterns to different potential pollution sources. A main advantage of using field data is that the natural background variation of marine ecosystems (e.g., seasonal variation) is implicitly included in the results that are obtained. However, field studies may be subject to perturbations that affect the bioavailability of organic chemicals. In this context, statistical ecological modeling can be useful to quantify the impact of the natural variability on the ambient concentrations of contaminants. To do so, we opted for generalized additive models (GAMs) because of their ability to simultaneously capture the linear and non linear effects of variables. Additive modeling is a nonparametric alternative to the more conventionally used generalized linear models (GLMs). GAMs only differ from GLMs in that GAMs use nonparametric smooth functions of the co-variates, whereas GLMs use products between the regression coefficients and the co-variate. As such, GAMs can, as a data mining method, disentangle additive effects of different predictor variables on the response variables from the data without making strong assumptions about the shape of the response function. The GAM regression technique consists of fitting smooth additive functions for each co-variate included in the model structure.
Table 1. Summary of the Chemical-Specific Generalized Additive Models*4

<table>
<thead>
<tr>
<th>Chemical</th>
<th>n</th>
<th>TOC</th>
<th>Longitude and Latitude</th>
<th>AIC</th>
<th>$R^2$</th>
<th>Deviance Explained</th>
<th>Figure</th>
</tr>
</thead>
<tbody>
<tr>
<td>Hg</td>
<td>362</td>
<td>$1.4 \times 10^{-15}$</td>
<td>$&lt;-2.0 \times 10^{-18}$</td>
<td>94.1</td>
<td>0.70</td>
<td>72.1%</td>
<td>Figure S12</td>
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<tr>
<td>Cd</td>
<td>362</td>
<td>$7.3 \times 10^{-12}$</td>
<td>$2.7 \times 10^{-18}$</td>
<td>160.6</td>
<td>0.41</td>
<td>45.7%</td>
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<tr>
<td>Ni</td>
<td>322</td>
<td>$&lt;2.0 \times 10^{-16}$</td>
<td>$&lt;-2.0 \times 10^{-18}$</td>
<td>$-345.5$</td>
<td>0.76</td>
<td>78.0%</td>
<td>Figure S14</td>
</tr>
<tr>
<td>Pb</td>
<td>362</td>
<td>$&lt;2.0 \times 10^{-16}$</td>
<td>$&lt;-2.0 \times 10^{-18}$</td>
<td>$-149.6$</td>
<td>0.73</td>
<td>74.8%</td>
<td>Figure S15</td>
</tr>
<tr>
<td>Zn</td>
<td>362</td>
<td>$&lt;2.0 \times 10^{-16}$</td>
<td>$&lt;-2.0 \times 10^{-18}$</td>
<td>$-302.4$</td>
<td>0.81</td>
<td>82.5%</td>
<td>Figure S16</td>
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<tr>
<td>NAP</td>
<td>243</td>
<td>$&lt;2.0 \times 10^{-16}$</td>
<td>$&lt;-2.0 \times 10^{-18}$</td>
<td>37.9</td>
<td>0.77</td>
<td>79.6%</td>
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<tr>
<td>FLU</td>
<td>242</td>
<td>$&lt;2.0 \times 10^{-16}$</td>
<td>$&lt;-2.0 \times 10^{-18}$</td>
<td>132.0</td>
<td>0.84</td>
<td>85.7%</td>
<td>Figure S22</td>
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<tr>
<td>PHE</td>
<td>243</td>
<td>$&lt;2.0 \times 10^{-16}$</td>
<td>$&lt;-2.0 \times 10^{-18}$</td>
<td>8.1</td>
<td>0.84</td>
<td>85.7%</td>
<td>Figure S23</td>
</tr>
<tr>
<td>ANT</td>
<td>241</td>
<td>$&lt;2.0 \times 10^{-16}$</td>
<td>$&lt;-2.0 \times 10^{-18}$</td>
<td>53.8</td>
<td>0.84</td>
<td>85.4%</td>
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</tr>
<tr>
<td>PYR</td>
<td>243</td>
<td>$&lt;2.0 \times 10^{-16}$</td>
<td>$&lt;-2.0 \times 10^{-18}$</td>
<td>$-39.4$</td>
<td>0.86</td>
<td>87.3%</td>
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<td>BAA</td>
<td>243</td>
<td>$&lt;2.0 \times 10^{-16}$</td>
<td>$&lt;-2.0 \times 10^{-18}$</td>
<td>8.0</td>
<td>0.80</td>
<td>81.7%</td>
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<tr>
<td>BAP</td>
<td>243</td>
<td>$&lt;2.0 \times 10^{-16}$</td>
<td>$&lt;-2.0 \times 10^{-18}$</td>
<td>86.1</td>
<td>0.74</td>
<td>76.9%</td>
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<tr>
<td>$\sum_{PAH}$</td>
<td>241</td>
<td>$&lt;2.0 \times 10^{-16}$</td>
<td>$&lt;-2.0 \times 10^{-18}$</td>
<td>$-58.6$</td>
<td>0.85</td>
<td>86.5%</td>
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<td>PCB28</td>
<td>113</td>
<td>$1.9 \times 10^{-12}$</td>
<td>$1.4 \times 10^{-14}$</td>
<td>25.1</td>
<td>0.80</td>
<td>84.2%</td>
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</tr>
<tr>
<td>PCB52</td>
<td>113</td>
<td>$1.6 \times 10^{-12}$</td>
<td>$6.9 \times 10^{-15}$</td>
<td>$-22.1$</td>
<td>0.79</td>
<td>83.9%</td>
<td>Figure S27</td>
</tr>
<tr>
<td>PCB101</td>
<td>113</td>
<td>$&lt;2.0 \times 10^{-13}$</td>
<td>$1.3 \times 10^{-13}$</td>
<td>22.5</td>
<td>0.73</td>
<td>75.0%</td>
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<tr>
<td>PCB118</td>
<td>113</td>
<td>$&lt;2.0 \times 10^{-13}$</td>
<td>$8.8 \times 10^{-16}$</td>
<td>0.1</td>
<td>0.78</td>
<td>79.9%</td>
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<tr>
<td>PCB153</td>
<td>91</td>
<td>$&lt;2.0 \times 10^{-12}$</td>
<td>$2.4 \times 10^{-15}$</td>
<td>17.3</td>
<td>0.74</td>
<td>75.4%</td>
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<tr>
<td>PCB180</td>
<td>113</td>
<td>$&lt;2.0 \times 10^{-13}$</td>
<td>$3.7 \times 10^{-16}$</td>
<td>6.7</td>
<td>0.80</td>
<td>81.7%</td>
<td>Figure S31</td>
</tr>
<tr>
<td>$\sum_{PCB}$</td>
<td>91</td>
<td>$&lt;2.0 \times 10^{-13}$</td>
<td>$2.1 \times 10^{-16}$</td>
<td>99.2</td>
<td>0.52</td>
<td>53.7%</td>
<td>Figure S32</td>
</tr>
<tr>
<td>DDTPP</td>
<td>96</td>
<td>$9.3 \times 10^{-10}$</td>
<td>$1.8 \times 10^{-14}$</td>
<td>91.7</td>
<td>0.71</td>
<td>77.2%</td>
<td>Figure S34</td>
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<tr>
<td>HCB</td>
<td>113</td>
<td>$7.0 \times 10^{-13}$</td>
<td>$5.4 \times 10^{-13}$</td>
<td>60.4</td>
<td>0.69</td>
<td>72.0%</td>
<td>Figure S35</td>
</tr>
<tr>
<td>$\alpha$-HCH</td>
<td>113</td>
<td>$1.7 \times 10^{-13}$</td>
<td>$1.6 \times 10^{-13}$</td>
<td>65.9</td>
<td>0.71</td>
<td>76.1%</td>
<td>Figure S33</td>
</tr>
</tbody>
</table>

*Each model included the total organic carbon content of the sediment (TOC) and an interaction term between latitude and longitude (eq 1). The $p$-values are reported to indicate the contribution of the co-variates to the model. Models were chemical-specific and based on a number of samples (n). They were assessed based on the Akaike Information Criterion (AIC), the adjusted correlation coefficient ($R^2$), and residual diagnosticsough the range of fitted values. One model was built for each chemical (22 in total) and for the sum of seven PCBs (i.e., PCB28, PCB52, PCB101, PCB118; PCB138; PCB153; PCB180; $pp'$-dichlorodiphenyltrichloroethane (DDTTPP); hexachlorobenzene (HCB); $\alpha$-hexachlorocyclohexane ($\alpha$-HCH); anthracene (ANT); benzo[a]anthracene (BAA); benzo[a]pyrene (BAP); naphthalene (NAP); fluorene (FLU); phenanthrene (PHE); pyrene (PYR); mercury (Hg); cadmium (Cd); zinc (Zn); lead (Pb); nickel (Ni); the sum of seven polycyclic aromatic hydrocarbons ($\sum_{PAH}$); and the sum of seven polychlorinated biphenyls ($\sum_{PCB}$).

These smooth functions are linear combinations of a finite number of cubic spline functions, with the smoothness of the function estimated by balancing between the model fit and the model complexity.53

In ecological research, GAMs have been often used. For example, Stenseth et al. (2006)26 used GAMs to study seasonal plankton dynamics under a complex scenario of physical and biogeochemical variability. More recently, Feng et al. (2015)25 used GAMs to analyze large-scale satellite derived data to characterize correlation patterns between environmental forcing and marine biomass. Furthermore, GAMs have been used to assess the risks of agricultural pesticides in small streams in Germany,27 to model the spatial distribution of deep-sea coral,28 to map the distribution of Baltic Sea flounder,29 and to infer trends from pollution data.30,31 In the present research, the chemical-specific GAMs incorporated TOC and geo-spatial information as:

$$\log_{10}([\text{chemical X}]) = b_0 + b_{x_1}([\text{TOC}]) + f_{x_2}(\text{latitude, longitude}) + \epsilon_x \quad (1)$$

where [chemical X] is the concentration of chemical X measured at different locations in Norwegian marine waters. A chemical-specific intercept is $b_x$ and $f_{x_2}$ ($i = 1$ or 2) are chemical-specific smooth functions (cubic splines with a maximum of four degrees of freedom for TOC; $f_{x_2}$) describing the partial additive effect of the predictor variable on the chemical concentrations. We included a smooth function on the interaction effect between latitude and longitude ($f_{x_2}$) to capture the geo-spatial variability of the concentration of chemical X. A chemical-specific normal distribution and independent noise term is $\epsilon_x$ (eq 1).

Model selection was based on the residual diagnostics as in Zuur et al. (2009).31 Homogeneity was concluded if the variability of the residuals was similar across the range of fitted values. One model was built for each chemical (22 in total) and for the sum of seven PCBs (i.e., PCB28, PCB52, PCB101, PCB118, PCB138, PCB153, and PCB180) and of seven PAHs (i.e., ANT, BAA, BAP, NAP, FLU, PHE, and PYR). Normality was assessed using a quantile—quantile plot and a histogram of the model residuals.32 For each of these chemical-specific models, we report the amount of deviance explained, the Akaike Information Criterion (AIC), and the significance level of the covariates (Table 1). The spatial pattern of each chemical was deduced directly from the latitude—longitude interaction term.

**Meta-Analysis.** The chemical-specific models (as described in the previous paragraph) were used to perform a meta-analysis to investigate which type of pollution (i.e., primary or secondary sources) was driving the environmental fate of the organic chemicals (i.e., PAHs, PCBs and pesticides). To do so, for each hydrophobic chemical we extracted the corresponding intercept ($b_x$) of the additive model and plotted these chemical-specific intercepts (17 in total) against the corresponding log$_{10}$ organic carbon—water partition coefficients (log KOC) and the corresponding molecular weights (MW). For five metals, the
same extraction was done as for the organic chemicals and the chemical-specific intercepts were plotted against the corresponding MW. The intercept of an additive model is a first indication of the magnitude of the concentrations, i.e., the higher the intercept, the higher the concentration of the corresponding chemical. If primary point sources have a dominant impact on the observed concentrations, then it is hypothesized that no clear relationship is found between the intercepts and KOC or MW as this suggests that the chemicals present in the different environmental compartments are not yet in equilibrium. Hence, we hypothesize that, in case of primary sources, no clear relation will be found between the chemical-specific intercepts against the corresponding log KOC and MW. However, if secondary sources dominate the concentrations of the chemicals, then clear intercept−KOC and intercept−MW relationships are expected across chemicals that belong to the same group (i.e., PAHs, PCBs, and pesticides). This type of meta-analysis is similar to the approach followed by Everaert et al. (2014) and Berrojalbiz et al. (2011). All statistical analyses were performed in R, and the additive models were fitted with the mgcv library.

3. RESULTS AND DISCUSSION

Data Exploration. The sediment concentration of metals, chlorinated compounds, and PAHs were dependent on the sampling site (Figures 1 and S5–S6 and Table S2). Median concentrations of heavy metals were on average 2- to 4-fold higher inside fjords than in offshore sampling locations (p < 1.8 × 10−3; Figure 1; Table S3–S7). For PAHs, we found that median concentrations of NAP, PHE, and BAP were not significantly different at the four geographical sampling locations (p > 0.20; Figure S6). However, concentrations of BAA, ANT, FLU, and PYR and of the chlorinated compounds were higher in fjords than in offshore sampling sites (p < 0.04; Table S8–S26). For example, in the inner fjord sampling locations median concentrations of ANT were 2.11 (first quartile = 1.57; third quartile = 6.95) μg kg−1 dry sediment. In the southern offshore sampling locations median concentrations were about three times lower (i.e., 0.76 (0.40–1.43) μg kg−1 dry sediment; p = 9.0 × 10−3; Table S2). Similar differences were found for PCB28, i.e., median concentrations in inner fjord sampling locations were 0.16 (0.14–0.36) μg kg−1 dry sediment, which is about 5-fold higher than in southern offshore sampling locations with median concentrations of 0.03 (0.02–0.09) μg kg−1 dry sediment (Table S2). It is important to note that for some specific PAHs (e.g., PHE and NAP) relatively high concentrations were found in northern offshore sampling sites at levels that are comparable or even higher than the concentrations observed in coastal sampling locations (Table S2 and Figure S6).

Although we acknowledge that it is compound-specific, overall, we found on average 2- to 4-fold higher concentrations of chemicals close to urbanized and industrialized sites than in rural less-populated areas. This pattern is likely to be related with point sources inside the fjords and diffuse catchment inputs paired with the fjord hydrology (see further). In addition, TOC is considerably higher in fjords (i.e., 17.2 (13.8–19.6) g 100 g−1 dry sediment) and along the coast (13.7 (7.38–20.5) g 100 g−1 dry sediment) than in offshore sampling locations (6.54 (3.47–12.4) g 100 g−1 dry sediment and 7.98 (5.35–12.4) g 100 g−1 dry sediment; Table S2 and S27 and Figure 1F), hence resulting in higher binding capacities of the sediment for chemicals. These findings corroborate with those of Everaert et al. (2014) in the North Sea and Gomez-Gutierrez et al. (2007) in the Mediterranean Sea.

Figure 1. Surface-sediment concentrations of cadmium (A), zinc (B), lead (C), nickel (D), mercury (E), and total organic carbon (F) divided according to the geographical location of the sampling sites. Each sampling site was attributed to a geographical category (i.e., inner fjord, coastal, offshore (North), and offshore (South)). Samples taken inside fjords and bays were categorized as "inner fjord". Samples taken at open sea but close to the main land were categorized as coastal. All other samples were classified as offshore, those above 66°N were considered northern offshore, and those below 66°N were considered southern offshore.
Chemical-Specific Models. GAMs with TOC and an interaction term between latitude and longitude as co-variates explained about 75% of the variability of the log-transformed sediment concentrations (Table 1). However, for cadmium, the variability explained was only 45.7%. The latter is related to the concentrations that were reported as 0.05 μg kg⁻¹ dry weight, which was the limit of quantification. Based on the straight line of residuals in the residual plots (Figure S7), we suspect that most of those concentrations were close to, but different from, the limit of quantification of cadmium. For all other chemicals under study, the residuals of the models were randomly distributed, followed a normal distribution and the predicted values plotted against the observed values were on a straight line parallel with the first diagonal (Figures S7–S10). Because the residuals do not show a particular pattern or bias the models are valid and useful to infer the chemical-specific spatial trends. Furthermore, the meta-analysis of the intercepts will give valuable information on the nature of the sources of the chemicals. For all the chemicals tested, we found a significant positive relationship between the TOC and the sediment concentrations of the chemical (p < 1.7 × 10⁻⁸; Table 1; Figure S11). As such, GAMs including TOC as a co-variates do take into account the spatial gradient of TOC (Figure 1F). This finding is in line with Ma et al. (2015), who found a positive correlation between the concentration of POPs and TOC in marine sediments. Indeed, many POPs are hydrophobic (log Kow > 5) and have an affinity for fatty tissues and organic material. As for POPs, metals showed a clear affinity for marine sediments with relatively high content of TOC (p < 7.3 × 10⁻¹¹; Table 1), corroborating with the results of Palanques et al. (1995) in the Gulf of Cadiz (Spain). Except for PCB180 (p = 2.1 × 10⁻¹¹; Table 1), the smoother of the interaction term between longitude and latitude of the sampling locations contributed in each model (p < 1.3 × 10⁻¹); Table 1).

Spatial Patterns and Sources. The interaction term between longitude and latitude was an important contributor to fit of the chemical-specific models (Table 1) and revealed the spatial patterns of the concentrations of the chemicals as it quantified the nonlinear influence of the position of the sampling location on the chemical concentrations. Dissimilar spatial patterns and sources were found for each group of chemicals. In the meta-analysis of the chemical-specific models (Figure 2A–C), we found a clear relationship between the chemical-specific intercept and the corresponding MW and log KOC. For both PAHs and PCBs, we found a linear trend between the intercept and the MW (r = 0.60 and r = 0.76, respectively) and between the intercept and the log KOC (r = 0.75 and r = 0.76, respectively) (Figure 2A - 2B). The latter suggests that mainly secondary sources contribute to the fate of the organic chemicals as the magnitude of the chemical-specific intercepts show a relationship (suggesting equilibrium conditions) with the chemical characteristics of the substances (see further resources). For metals, in spite of the fact that the amount of TOC plays an important role in their concentrations, i.e., the more TOC, the higher the concentrations of metals (p < 7.3 × 10⁻¹¹; Table 1), the meta-analysis showed no uniform relation between the MWs and the intercepts of the chemical-specific models (Figure 2C). Because there was no clear relationship between the intercepts and the corresponding MWs (Figure 2C), it is suggested that mainly direct primary sources contribute to their environmental concentrations. Indeed, for metals a predominantly hotspot driven spatial pattern was found (Figures S12–S16), with higher than average metal concentrations in fjords and bays (Figure 1). Both natural factors (i.e., metal rich ores or bedrock in combination with erosion) as well as anthropogenic sources (i.e., mining and metal industries) are hypothesized to contribute to the elevated metal concentrations at particular sites. Among the anthropogenic sources, we count the hydroelectric power plants that were installed and energy-demanding industries like smelters and metallurgical plants that were developed in the past century. Prior to (inter)national legislation, untreated waste products from these industries were often discharged directly into the marine environment. Fjords often have a shallow sill near the mouth that limits water exchange with open marine waters, resulting in longer water retention times in fjords than in open waters. As such, recipient waters from a
point-discharge within a fjord might have a relatively high contaminant concentration (either in the dissolved phase or associated with particles) that can persist for a long time and increase contaminant loading to bottom sediments.\textsuperscript{41,43} In addition, contaminants delivered by rivers and freshwater runoff are also affected by the "salting out" phenomenon, whereby increased salinity in receiving fjord waters causes increased partitioning of contaminants into the particulate phase, which can lead to enhanced sedimentation of these contaminants.\textsuperscript{44} For several metals, such as mercury (Figure S12), cadmium (Figure S13), lead (Figure S15), and zinc (Figure S16) the most dominant feature of the spatial pattern is a hotspot in Sørfjorden (in western Norway), where there has been important local input from a zinc smelter.\textsuperscript{18,19} In Sørfjorden, the jasore waste of metallurgical activities was discharged in the fjord until 1986 and was long-considered to be one of the most metal-polluted fjords in the world.\textsuperscript{18} It appears that the metal pollution in this region mainly originates from (historical) primary sources and is still residing in the sediment of the corresponding fjords and bays. A study by Temara et al. (1998)\textsuperscript{84} found that within Sørfjorden, concentrations of lead, cadmium, and zinc were consistently higher (1.7 to 8.3 times) in the brittle star \textit{Asterias rubens} sampled close to the smelter than in those sampled further away. Remedial actions in Sørfjorden have included the disposal of metal-containing sludge in mountain caverns and the containment of contaminated sediment by the use of sheet pilings or capping. Highest concentrations of nickel have been found along the northern part of the Norwegian coast (Figure S14), corroborating the findings of Reimann et al. (2000) and Sandager et al. (2011).\textsuperscript{46,47} and in offshore sampling locations north-west of Trondheim, where there is an important deposition zone of suspended solids to the sediments (Figures S13–S14).\textsuperscript{39} In other regions, such as for example the Trondheim fjord, Faust et al. (2014)\textsuperscript{39} have demonstrated the importance of greenstone bedrock for the geochemical composition of the surface sediments. In this region, there is evidence that the bedrock is an important natural source for Zn and Ni. Because Zn and Ni are often associated with inorganic fine grain part of the sediment (i.e., clay and silt fractions),\textsuperscript{46} it is likely that the high Zn and Ni concentrations in and around the Trondheim fjord mainly originate from a mixture of natural and anthropogenic pollution sources.

Metals such as mercury and lead are more strongly associated with the organic carbon content and have mainly anthropogenic sources (Figure S11).\textsuperscript{48} We identified a pollution hotspot of lead in Sørfjorden, and elevated concentrations of lead in Skagerak and along the coast of southern Norway. The elevated concentrations in these areas are directly related to the industrial applications of lead and the high degree of human activity. Elevated lead concentrations in southern Norway likely reflect a combination of local and regional inputs as well as long-range inputs from Europe (both historical and current) delivered through atmospheric deposition and rivers/runoff. Some of the organic lead antifouling agents have been manufactured locally, but that is considered to be a minor source.\textsuperscript{30} Hotspots of cadmium were found in Sørfjorden and close to the Lofoten archipelago (Figure S13), which is consistent with the findings of Julshamn et al.\textsuperscript{30} who observed high cadmium concentrations in edible crabs in Vestfjorden. Cadmium is found naturally in minerals and soils and is transported with both water and air,\textsuperscript{51} but cadmium has also been used in various industrial processes such as protecting steel against corrosion.\textsuperscript{59} In 2007, the emission of cadmium was estimated to 1.4 tons,\textsuperscript{53} with the highest discharges in Sørfjorden, Årdalsfjorden, and Orkdalsfjorden. Furthermore, it was estimated that an additional 3.5 tons of cadmium reach Norway through atmospheric long-range transport,\textsuperscript{52} which mostly affected the southern part of Norway.\textsuperscript{53} In 2009, the emission from closed mines to Norwegian waters were estimated to 0.3 tons and is considered the most-important known point source. In the same year, emissions from 63 industrial plants amounted to a comparable quantity.\textsuperscript{52} Overall, the interpretation of the meta-analysis (Figure 2C), that the concentrations of metals in Norwegian marine waters is mainly driven by primary point-sources and, to a limited extent, by diffuse long-range atmospheric transport, is reflected in their hotspot-like spatial patterns (Figures S12–S16) and is supported by scientific literature.

For the chlorinated pesticides, PCBS, and PAHs mainly diffuse, secondary sources modulate their concentrations because clear intercept–MW (Figure 2A) and intercept–log \(K_{OC}\) (Figure 2B) relations were found across chemicals that belong to the same group of pollutants. Multiple biogeochemical processes affect the fate of the organic chemicals in the marine environment.\textsuperscript{54} The chemical-specific spatial distributions that were inferred based on the GAMs (Figures S17–S35) have evolved over the past years and decades through complex interactions between a myriad of biogeochemical processes. In this context, we identified signs of these biogeochemical processes in the spatial patterns that were inferred, suggesting that they have played a role in the fate of organic chemicals in Norwegian marine sediments. Without having the intention to disentangle the interactions of the biogeochemical processes that were involved (which was not possible based on the models developed in the present research), these signs are discussed below. Overall, we found that mainly secondary and diffuse sources contribute to the fate of the organic chemicals as the magnitude of the chemical-specific intercepts show a relationship with the chemical characteristics of the substances, suggesting equilibrium conditions (Figure 2A,B).

Between northern Norway (main land) and Svalbard, the concentrations of PAHs gradually increase from south to north (Figures S17–S24). The presence of PAHs here is, first of all, related to diffuse air–water exchange in atmospheric and ocean currents from Central and Western Europe because at these high latitudes, there are few direct anthropogenic sources.\textsuperscript{34–56} In addition to the diffuse pollution via long-range atmospheric and aqueous transport,\textsuperscript{56} the erosion of coal-bearing bedrock also plays an important role (Figures S17–S24).\textsuperscript{55,58} The latter is a natural process and is considered a primary source of PAHs in the region. Furthermore, as the shallow Barents Sea bank (Svalbard Bank) is one of the most productive Arctic seas,\textsuperscript{59,60} the role of the biological pump in the fate of the organic chemicals must be considered. By means of the biological pump mechanism, organic chemicals are taken up by marine phytoplankton, and these plankton-bound chemicals are subsequently transferred to deeper water and sediment.\textsuperscript{61} Especially during, and directly after, the phytoplankton bloom, the flux of organic chemicals to sediment is high.\textsuperscript{61} Note that this process mainly affects PAHs with high MWs and that more than 90% of the PAHs with two or three aromatic rings that reach the surface water via diffuse air–water exchange are degraded before they reach the sediment.\textsuperscript{62–64}
As in the present research, both two- and three-ring PAHs as well as high-MW PAHs have elevated concentrations in the region between North Norway and Svalbard, and the impact of coal-eroding bedrock seems more important than the biological pump mechanism. Unfortunately we lack data on the PCB concentrations in this region. These PCB data could have served as an independent validation tool with few authochtonous inputs in this region. Therefore, we hypothesize that the air–sediment transport as mediated by the OC settling fluxes (biological pump) are most important for high-MW PAHs and only to a lesser extent for low-MW PAHs for which diffuse air–water exchange and coal-eroding bedrock are expected to be the main sources.

In the southern part of the study area, the highest concentrations of PAHs were observed in the North Sea and the Skagerrak (Figures S17–S24), which relates to continuous diffuse atmospheric inputs from households and industry. In the North Sea, as of 1969, oil and gas exploitation platforms have been developed in Norwegian marine waters and via sea and wind currents these regions receive considerable inputs of PAHs from these platforms and from Western and Central Europe. In this context, Boitsov et al. (2013) compared the PAH composition and sources in the North Sea and Skagerrak (IMR monitoring data) with those of the southwestern part of the Barents Sea (MAREANO data). It was highlighted that pristine areas of the Barents Sea have lower levels of PAHs than the Skagerrak. The elevated PAH levels in the Skagerrak were mainly attributed to a mixture of sources such as extensive offshore petroleum industry in the North Sea, intense ship traffic, larger populations, and industrial activities.

For chlorinated compounds such as PCB28, PCB52, PCB101, PCB118, PCB138, and α-HCH a gradient, with decreasing concentrations from the Southeast to the Northwest of the study area, was found (Figures S25–S31 and S33). The sediment concentrations of these compounds decreased in parallel to the Norwegian coastline and the highest concentrations were identified in the North Sea and Skagerrak. For the chlorinated compounds, riverine inputs and point sources are most important in fjords and coastal systems (Figure S5). However, for most marine regions, and certainly those located offshore, atmospheric deposition is the main source of pollution with PCBs and chlorinated pesticides. For these compounds, there are mainly historical sources, and even though there is long-range transport, this is a slow process historically. Sediments integrate this historical signal, and once the PCBs and chlorinated pesticides are captured in sediments they may not be remobilized, except in shallow areas. In addition, as for PAHs, the transport to sediments due to settling particles (biological pump) will be favored for the more hydrophobic congeners, for example PCB180. In this context, it is not unexpected that the spatial pattern of congener PCB180 differs from the other congeners, with low concentrations in the Skagerrak but a clear hotspot close to Varanger in Northwest Norway (Figure S32). It is hypothesized that the PCB180 concentrations in this region are mainly originating from (historical) pollution from Russian smelter activity, which, in combination with the relatively high productivity of the Barents Sea at the Kola peninsula, has resulted in elevated concentration of PCB180 in the sediment in this part of the study area (Figure S32).

Relevance for Management. In the meta-analysis of the chemical-specific models (Figure 2), we found a clear relationship between the chemical-specific intercept and the corresponding MW and log $K_{OC}$. The latter suggests that mainly secondary sources contribute to the fate of the organic chemicals as the magnitude of the chemical-specific intercepts show a relationship (suggesting equilibrium conditions) with the chemical characteristics of the substances. Therefore, in the discussion of which remediation strategy might be most efficient in the long run, decision makers should focus on both local sources and conditions as well as the chemical characteristics of the targeted substances. The main reason to do so is that PAHs and PCBs (and all organic chemical in general) partition between different environmental compartments (i.e., air, water, soil, sediment, and biota) and aim for a mass balance in equilibrium. Cleaning one compartment will not be successful if the remaining compartments can still act as potential sources. Furthermore, local conditions such as ship traffic in harbors may cause propeller-induced erosion and has a negative effect on environmental conditions by exposing and remobilizing sediments from previous decades. In some cases, severe (historical) pollution capping of the sediment seems to be a good option and has already been shown as successful. However, urban runoff in Norwegian fjords might also be a small, but significant, contributor to pollution levels. Although it is clear that selecting the most-optimal remediation strategy is not straightforward, the present research summarizes the pollution levels of 22 priority hazardous substances in the Norwegian marine waters. The overlap of these maps with the quantitative input of chemicals (for example, via urban runoff or atmospheric deposition) at the different sites will give policy-makers valuable information with which to select the most-efficient remediation strategy.

Relevance for Monitoring. The present research used 5036 points of data for 22 chemicals sampled between 1986 and 2014 at 333 sampling sites. Despite the size of the database, it is difficult to infer time trends because no locations are sampled on a frequent basis. Few locations were sampled three or four times between 1986 and 2014, and many more were only visited once. Taking into account possible seasonal changes of sediment concentrations and cyclic interannual changes due to climatic oscillations, this is too few to characterize the temporal trends of the concentrations. Furthermore, we selected at each sampling location the top 1 cm layer. However, sedimentation rates may largely vary in our study area, with high sedimentation rates at some locations (e.g., order of magnitude of centimeters per year) and low sedimentation rates in other areas (e.g., the order of magnitude of millimeters per year). The latter means that a slice of sediment of one locations may represent multiple years or even decades, whereas the same slice from other locations may integrate only few years. Hence, caution is needed when inferring time trends from this type of data. So, in spite of the large integrated data set, unbiased time trends could not be inferred. A potential solution is the use of dated sediment cores, as available in the MAREANO, program that may document past changes in the inputs and concentrations of hazardous substances.

**ASSOCIATED CONTENT**

**Supporting Information**

The Supporting Information is available free of charge on the ACS Publications Web site at: The Supporting Information is available free of charge on the ACS Publications website at DOI: 10.1021/acs.est.7b02964.
Additional information including an overview of the original data set and of the data set after preprocessing; median concentrations of 22 substances categorized per geographical area, tables showing the nonparametric Kruskal–Wallis statistical tests for each substance, maps showing the locations and depths of the sampling sites, box-plots with concentrations of 22 substances categorized per geographical area, residual diagnostics to assess model fit, the smoother of TOC, and spatial patterns for each substance. (PDF)

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■ REFERENCES


(43) Saloranta, T. M.; Armitage, J. M.; Haario, H.; Naes, K.; Cousins, I. T.; Barton, D. Modeling the effects and uncertainties of contaminated sediment remediation scenarios in a Norwegian fjord by


