Oil Spill Detection Using Fluorometric Sensors: Laboratory Validation and Implementation to a FerryBox and a Moored SmartBuoy

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A large part of oil spills happen near busy marine fairways. Presently, oil spill detection and monitoring are mostly done with satellite remote sensing algorithms, or with remote sensors or visual surveillance from aerial vehicles or ships. These techniques have their drawbacks and limitations. We evaluated the feasibility of using fluorometric sensors in flow-through systems for real-time detection of oil spills. The sensors were capable of detecting diesel oil for at least 20 days in laboratory conditions, but the presence of CDOM, turbidity and algae-derived substances substantially affected the detection capabilities. Algae extract was observed to have the strongest effect on the fluorescence signal, enhancing the signal in all combinations of sensors and solutions. The sensors were then integrated to a FerryBox system and a moored SmartBuoy. The field tests support the results of the laboratory experiments, namely that the primary source of the measured variation was the presence of interference compounds. The 2 month experiments data did not reveal peaks indicative of oil spills. Both autonomous systems worked well, providing real-time data. The main uncertainty is how the sensors’ calibration and specificity to oil, and the measurement depth, affects oil detection. We recommend exploring mathematical approaches and more advanced sensors to correct for natural interferences.

Keywords: oil spill, flow-through system, fluorometric sensors, Baltic Sea, natural interferences, sensor selectivity

1. INTRODUCTION

Oil spills are a major threat to the marine ecosystems, local communities and economy (Samiullah, 1985; Farrington, 2014; Jørgensen et al., 2019; Câmara et al., 2021; Sandifer et al., 2021). The research into the consequences of oil pollution has been long and extensive. The effects to wildlife are broad, ranging from exposure of birds (Jenssen, 1994; Stephenson, 1997; Fox et al., 2002; Ridoux et al., 2004) to oil to toxic, mutagenic and/or carcinogenic effects of polycyclic aromatic hydrocarbons (PAHs) present in crude oil and products based on fossil oil (Hylland, 2006; Abdel-Shafy and Mansour, 2016; Hayakawa, 2018; Honda and Suzuki, 2020). Besides harming the natural environment, oil spills can impair the economy in the affected region (Cohen, 1993; Taleghani and Tyagi, 2017; Ribeiro et al., 2021) and have adverse effects on human health and psychology (D’Andrea and Reddy, 2014; Shultz et al., 2015; Sandifer et al., 2021).
The Baltic Sea has always been an important route for maritime trade and transport, accounting for up to 15% of the world’s maritime traffic (HELCOM, 2003; WWF, 2010). Oil shipments in the Baltic Sea are projected to grow by 64% by 2030, from about 180 million tons to nearly 300 million tons (HELCOM, 2018a) and the overall volume of ship traffic has been estimated to double during the period 2010–2030 (Rytikönen et al., 2002). The immense volume of shipping in the Baltic Sea is accompanied by a high risk of accidents. According to the Helsinki Commission (HELCOM), 1,520 maritime accidents have occurred in the Baltic Sea area during the period 2011–2015, with a fairly stable rate of 300 accidents per year; 4% of these accidents led to loss of life, serious injuries or environmental damages (HELCOM, 2018a).

A considerable contributor to marine oil pollution is oil pollution by ships, which are concentrated to main shipping lanes and other areas of high maritime activity (Serra-Sogas et al., 2008; Ferraro et al., 2009; Liubartseva et al., 2015; Sankaran, 2019; Polinov et al., 2021). It is estimated that in the Baltic Sea 10% of the total amount of oil hydrocarbons comes from illegal discharges by vessels (HELCOM, 2003). These various smaller spills, referred to as operational oil spills, are not the result of ship accidents, but instead result from discharges of small amounts of oil, or more usually unfiltered oily water. Most of this pollution risk is concentrated along major ship routes (HELCOM, 2013). The number and size of these spills has been decreasing (HELCOM, 2018c). For example, in 2015 the number of observed spills was 80 and the total estimated annual volume of oil spills observed in 2009–2015 was in the order of 20 m³ (HELCOM, 2018a). However, even small amounts of oil can have a negative impact on the marine environment (Brussaard et al., 2016).

Current oil spill remote sensing methods have become more reliable but they still have many drawbacks and limitations (Fingas and Brown, 2018). Nowadays the most dominant and cost-effective means for remote spill detection is the combination of satellite-based synthetic aperture radar (SAR) images and aircraft surveillance flights for verification (Gade et al., 2000; Uiboupin et al., 2008; Solberg, 2012; Fingas and Brown, 2018). SAR sensors give a good coverage and are not limited by cloud cover or weather conditions. Furthermore, there are many studies on algorithms meant for identification of oil spills from SAR data (Marghany, 2016; Krestenitis et al., 2019; Al-Ruzouq et al., 2020; Zeng and Wang, 2020). Besides SAR, optical remote sensing techniques gathering information in different spectral range (ultraviolet, visible, and near-infrared) and can give useful information about oil pollution (Solberg, 2012; Fingas and Brown, 2018; Al-Ruzouq et al., 2020). Ships and aircraft equipped with radar or optical sensor are also widely used for detection and monitoring of oil pollution (Jensen et al., 2008; Xu et al., 2020). Also, sensors like microwave radiometers and laser fluorometers mounted on aircraft can provide additional information about the oil type and amount (Jha et al., 2008; Solberg, 2012).

Oil spills can be difficult to detect even with modern aerial surveillance equipment for numerous reasons. They can be small, and in rough sea the oil is mixed well below the surface, while a visible slick might also disappear because of evaporation. Optical methods of satellite sensing are also limited by resolution, cloud cover and the sea state (Brekke and Solberg, 2005; Jha et al., 2008; Fingas and Brown, 2018). Even the most studied and used remote sensing method of detecting oil from SAR images has problems, such as lookalikes—radar signatures similar to oil pollution but which can actually be for example floating algae, ship wakes, cold upwelling water or natural surface films produced by plankton or fish (Sipela and Uiboupin, 2007; Alpers et al., 2017; Al-Ruzouq et al., 2020). Even a comprehensive satellite coverage might therefore not be able to detect all oil spills. Nonetheless, PAHs will remain in the water after the spill is no longer visible on the surface but are still detectable by in-water fluorometers. Thus, the real-time detection of these spills would benefit from additional detection systems. In addition, in-situ sensors can be used for validation for remote sensing techniques and vice versa. One possible in-situ solution is to use fluorometric sensors installed on a suitable platform.

Ships of Opportunity (SOOPs) fitted with oceanographic instrumentation and automated water sampling systems, so-called FerryBoxes, are commonly used for studies of near surface waters (Petersen et al., 2003, 2011; Hydes et al., 2010; Petersen, 2014). Also, the Baltic Sea is well covered with FerryBox lines (Petersen, 2014; Schneider et al., 2014; Karlson et al., 2016; Kikas and Lips, 2016). FerryBoxes have a great potential for gathering scientific data, especially when installed on ferries and cargo ships cruising the same route on a regular basis. While SOOPs give a good spatial coverage, monitoring buoys give an excellent temporal coverage. Such autonomous monitoring buoy systems are being developed and operated worldwide to measure the physical and biogeochemical properties of coastal surface waters (Mills et al., 2003; Nam et al., 2005). The technological progress has resulted in SmartBuoys with a stable power supply, two-way communication and real-time data acquisition for effective environmental monitoring (Chavez et al., 1997; Mills et al., 2003; Nam et al., 2005; Benson et al., 2008; Papoutsou et al., 2012). Moroni et al. (2016) have developed a sensorized buoy for detecting oil spills from the air-side. Fluorescence-based in-situ sensors and systems have been globally used for real-time monitoring of oil spills and determining the levels of the contamination (Lambert et al., 2001, 2003; He et al., 2003; Kim et al., 2010; Malkov and Sievert, 2010; Tedetti et al., 2010). Combining FerryBoxes and SmartBuoys with portable fluorometric sensors could provide an additional method for early notification of oil spills in the sea. Nonetheless, to our knowledge, such an approach has not previously been adopted.

Fluorometric detection (FLD) is essentially about measuring fluorescence at predetermined wavelengths, meaning that FLD can’t resolve between different sources of hydrocarbons and fluorescent compounds – for example oil and oil refined products, combustion processes (e.g., power plants, maritime and land-based traffic, forest fires). Compound-specific laboratory methods such as gas chromatography–mass spectrometry can make such distinctions, but field-usability (data continuity), speed and low running costs still make FLD-based field protocols appealing alternatives compared to laboratory-based techniques.

For purposes of the operational detection of oil spills, the accuracy of the sensor is a crucial, but not the sole part. We split
up the operational chain of oil-detection with fluorometers to five steps: 1) the sensors need to accurately and selectively detect oil, 2) the system where the sensors are operating need to function at least semi-autonomously, 3) the system needs to reliably transmit real time data, 4) an automated algorithm detects anomalic events and 5) the data is available to the user in a reliable interface on a short notice. In this study we focus on steps 1–3 and 5. We first present laboratory tests of sensors and then evaluate the real-time oil detection capability of two autonomous in-situ remote sensing platforms that are equipped with fluorometers. The two platforms, a FerryBox and a SmartBuoy, covered high-risk areas in the Baltic Sea.

2. MATERIALS AND METHODS

2.1. UviLux, EnviroFlu-HC 500 and Turner Design C3 Fluorometers

Three commercially available fluorometric instruments, designed for in-situ and real-time quantification of oil or oil compounds, were chosen for the experiments: UviLux and EnviroFlu-HC 500 for the FerryBox system, and C3 Submersible Fluorometer for the SmartBuoy. All three sensors were also tested in laboratory conditions.

The UviLux UV-fluorometer (Chelsea Technologies Ltd, UK) is an in-situ UV fluorometer. Oil detection is based on the measurement of PAH concentrations. The sensitivity of the sensor is 0.005 µg carbazole per liter, the calibrated range is 0.005–2,000 µg carbazole per liter, the excitation wavelength is 255 nm, and the emission wavelength is 360 nm.

The EnviroFlu-HC 500 submersible UV fluorometer (TriOS Optical Sensors, Germany) is an instrument designed for the measurement of PAHs in water. The fluorometer has a minimum detection limit of 0.3 µg phenanthrene per liter, a calibrated range of 0–500 µg phenanthrene per liter, an excitation wavelength 254 nm, and an emission wavelength of 360 nm.

The Turner Design C3 submersible fluorometer (Turner Design, USA) is manufactured according to users' requirements. C3 fluorometers come with a factory-installed temperature probe and can be configured with up to three optical sensors. Each optical sensor is designed with fixed excitation and emission filters. The SmartBuoy experiment the C3 fluorometer was equipped with three sensors: a hydrocarbon sensor for crude oil with sensitivity of 0.2 µg pyrene-tetralsulfonic acid (PTSA) per liter and linear range 0–1500 µg PTSA per liter, excitation light 325/120 nm and emission light 410–600 nm; a sensor for colored dissolved organic matter (CDOM) with minimum detection limit (MDL) 0.1 µg quinine sulfate per liter, linear range 0–1.5 µg quinine sulfate per liter, excitation wavelength 325–120 nm and emission wavelength 470–60 nm; a turbidity sensor with MDL 0.05 Nephelometric Turbidity Units (NTU) and range 0–1,500 NTU, excitation wavelength 850 nm, and emission wavelength 850 nm.

2.2. Laboratory Tests

In order to examine selectivity and sensitivity of fluorometric sensors to oil, two separate laboratory experiments were performed in 2017. All experiments were performed in a dark climatic room set to a temperature of 16°C.

The first experiment examined the selectivity of the sensors, i.e., interferences to oil detection caused by interfering substances. Baltic Sea water (S; salinity 6.2 PSU) was used as a blank and as a solution for extraction of humic substances (H), cyanobacterial algae (A) and production of clay suspension (C).

Stock solutions (one of each) of clay suspension (from Baltic Sea’s clayey sediment), seawater extract of humic substances (originating from decayed plant material) and a seawater extract of cyanobacterial algae (from dried Baltic Sea surface phytoplankton) were produced. Also, two batches of water accommodated fraction (WAF; W) of diesel oil were produced. Each extract was made by combining 100 ml of commercial winter-quality diesel oil with 1,000 ml of seawater under gentle stirring overnight. All solutions were prepared 1 week prior to measurements in the aquarium. Concentrations of PAHs were quantified using gas chromatography—mass spectrometry and aliphatic hydrocarbons (decane-tetracontane) using gas chromatography—flame ionization detection at SYKE laboratory center. No attempt was made to determine chemical composition of each interference solution in detail, instead the solutions mimicked natural constituents in Baltic Sea water.

Altogether six glass beakers (each 2,000 ml) were wrapped with black plastic except for the top section and filled with 1,200 ml of seawater. Aliquots of the clay suspension (50.0 ml), humic extract (50.0 ml), cyanobacteria algae extract (10.0 ml) and diesel WAF (50 ml) were sequentially added so that all combinations of the aforementioned solutions in seawater were achieved. The solutions were kept in slow stirring motion using a magnetic stirrer. In first three sets of measurements no WAF was added. WAF was applied in three last measurement sets. After addition of each solution the responses were measured using one sensor at a time at exactly 50 mm below the upper level of the beakers. The top part of the beakers and the sensor housing were wrapped with black plastic. Fluorescence responses were collected for 3–5 min with each sensor and the average and standard deviation of fluorescence for each period of data collection were calculated.

The logging rates used with fluorometer sensors during the experiment were 1 min (average of values collected at 0.5 Hz), 30 and 1 s for the UviLux, EnviroFlu-HC and Turner C3 sensors, respectively. Data from EnviroFlu-HC were logged into a laptop PC using the TriOS MSDA_XE software and the data from Turner C3 into a PC using Turner C-soft software. Data from UviLux were logged into a portable logger connected to automatic GSM network-based modem and sent via GSM network to Tallinn University of Technology once a minute. Altogether 1.32 million, 79,000 and 30,000 time points of data from Turner C3, Enviroflu-HC and UviLux were generated, respectively. The small dilution effect in WAF solutions, caused by the volume increase by the added interference solutions, was compensated by a corresponding multiplier factor in final calculations thus yielding oil-related fluorescence (ORF) presented. Normalized responses from each fluorometer were calculated by dividing observed responses from different solutions with responses obtained...
from seawater. No compensation was calculated for CDOM or turbidity values.

The second experiment for diesel WAF persistence and for method comparison lasted for 3 weeks. A 25-liter glass aquarium was used and placed over a black plastic sheet in a steel tank and filled with 21.8 liters of filtered seawater (salinity 6.2 PSU). Thermostatted (11°C) tap water was circulated in the exterior tank. The FLD sensors were fixed into laboratory stands so that the optical window of each sensor was at 10 cm below the surface of liquid (Figure 1). Gentle mixing was applied using an electricity-powered laboratory stirrer. The aquarium was covered by a black plastic wrapping except during water sampling and maintenance. The UviLux sensor needed to be moved further away from the two other sensors due to interference that was unfortunately not noticed before 7 days into the experiment. The data transmission, storage and sensors’ acquisition rates were as described for the first experiment.

The persistence of diesel oil fractions in seawater and comparability between oil detection methods were measured in the second test. The methods were: the three fluorometric sensors, total oil monitoring method used by Finland, gas chromatography flame ionization detection (GC–FID; for aliphatic hydrocarbons) and gas chromatography—mass spectrometry (GC–MS; for PAHs including 1- and 2- methylnaphthalene). At S1–S9, water samples of 100 ml were collected for total oil HELCOM protocol and another 100 ml for the GC–MS and GC–FID analyses.

The Finnish HELCOM monitoring method for total oil analysis can be described briefly as follows: seawater is extracted with 10 ml of n–hexane under stirring. Fluorescence is measured using 310 and 360 nm excitation and emission wavelengths, respectively. Calibration is performed using solutions of Norwegian crude oil (Ekofisk) dissolved in n–hexane. Limit of quantification is 0.05 µg oil/l and analytical error is ±15%.

Aliphatic hydrocarbons (decane-tetracontane) were extracted using n-hexane, cleaned up using adsorbent, separated using gas chromatography and quantified by flame ionization detection. The method has a limit of quantification of 100 µg total aliphatics/l and analytical error of 30 %. PAHs were extracted using n-hexane under stirring or solid phase extraction. The n-hexane solution was then concentrated and analyzed using GC–MS. The quantification limits of the method were between 0.005 and 0.01 µg/l, while the analytical errors varied between 20 and 40% depending on compound.

Normalized responses for each detection method were calculated by dividing observed responses with period S1 responses at given water sampling time points (S1–S9).

2.3. FerryBox Operation and Oil Detection Capability

A FerryBox system developed by Marine Systems Institute at Tallinn University of Technology is used on board of the M/S Baltic Queen (Tallink Group, Estonia) (Figure 2), which commutes between Tallinn (Estonia), Mariehamn (Åland) and
Stockholm (Sweden) (Figure 3). The ship route covers the western Gulf of Finland, Northern Baltic Proper, Southern Archipelago Sea, and Southern Åland Sea (Figure 3). One crossing, including a stop in Mariehamn harbor, is about 425 km long and takes approximately 17 h. The ship then returns after a 7 h stay in the destination harbor.

All FerryBoxes are flow-through systems where the water is taken in from an inlet in a vessel hull and then pumped into a measuring circuit containing sensors (Petersen, 2014). On M/S Baltic Queen the water was taken from the ship’s sea chest and pumped continuously through the FerryBox system at a rate of 9 l/min. The opening of the chest is situated about 4 m below the waterline. Parameters were measured with 1-min intervals, giving an approximately 0.5 km spatial resolution along the fairway, depending on the ship’s speed. A typical transect contained roughly 1,000 records for each variable.

The UviLux and EnviroFlu-HC fluorometers, were installed in parallel to the FerryBox system (Figure 2B), which recorded additional real-time seawater properties. Conductivity (salinity) and temperature was measured with a High-Precision Pressure Level Transmitter Series 36XiW (KELLER AG, Switzerland), turbidity was measured with a Seapoint Turbidity Meter (Seapoint Sensors Inc., USA) and pCO2 was measured with an OceanPack pCO2 analyzer (SubCtech, Germany).

We made four maintenance visits to M/S Baltic Queen in order to clean the FerryBox system and the sensors. These visits were made roughly every 2 weeks (March 3, March 20, April 3, April 15) during our 2-month experiment (February 21 to April 21, 2018). The maintenance consisted of washing the system with a solution of oxalic acid and a manual cleaning of the optical sensors. The FerryBox also has a automatic cleaning system, which was not functional during the experiment. Similar automatic acid-washing cleaning method is applied to prevent biofouling on a FerryBox traveling between Tallinn and Helsinki (Lips et al., 2016).

Measurements from the sensors in the FerryBox were collected by a datalogger (RTCU-MX2i pro by Logic IO) on board M/S Baltic Queen. The datalogger includes a modem and a GPS, and adds the position and a time stamp to the measurement before sending the data to an on-shore FTP server of the Marine Systems Institute using the GSM/GPRS protocol. The data were sent in real-time (every minute).

A publicly available, web-based, user interface was built to visualize the data online, where the ship’s track, real-time position and gathered FerryBox data was available in real-time (Figure 4) (TalTech, 2017). The web-based user interface is also equipped with different options to view historical data: the user can select parameters or time periods, and construct a map view and 2D graphs of multiple parameters. Data are also available in a tabulated form and parameters can be viewed in color-coded view along the ship’s track. The data that were provided in real-time during the experiment is still available on the web-page.
2.4. SmartBuoy Setup and Experiment for Oil Spill Detection

The SmartBuoy, developed by Meritaito Ltd (Finland), is a combination of a polyethylene spar buoy with mobile and/or satellite communication technology and a versatile selection of monitoring sensors (Figure 5). The Turner Design C3 fluorometer was installed inside of the buoy in a vertical monitoring well with an open flow through a pipe, enabling continuous sea water exchange. The monitoring depth was 2–3 m depending on the sea level. Concentration of CDOM and turbidity values were measured simultaneously with the hydrocarbon measurements. To ensure high data quality, the C3 sensor was equipped with a mechanical wiper cleaning all three optical lenses before taking the measurements. In addition to the water quality monitoring, the SmartBuoy also collected data about current speed and direction at the depth of 7 m to identify spreading direction of potential oil spills. Furthermore, the significant wave height was calculated from roughly 8.5-min time series, measured by a pressure sensor (Aanderaa Wave and Tide Sensor 5218).
FIGURE 4 | Web-based user interface for viewing the real-time data from the FerryBox system (TalTech, 2017). The gaps in the data were caused by loss of the GSM-network.

FIGURE 5 | Schematic figure of the SmartBuoy system (A) and the SmartBuoy deployed to the off-shore area in the Gulf of Finland for oil spill detection (B). Photograph: Joose Mykkänen.
Table 1 | Relative responses (fold compared to average fluorescence in seawater) of sensors to interference solutions, sensitivity to water accommodated fraction (WAF) of diesel oil.

|                          | UviLux     | Turner C3                      | TriOS Enviroflu-HC |
|--------------------------|------------|--------------------------------|--------------------|
| Maximum attenuation      | none       | 0.95 (clay suspension)         | 0.97 (clay suspension) |
| without diesel WAF       |            |                                |                    |
| Maximum false positive   | 7.61 (algae extract) | 3.96 (humic extract + algae extract) | 2.64 (algae extract) |
| without diesel WAF       |            |                                |                    |
| Sensitivity to WAF       | 4.48       | 1.1                            | 2.62               |
| (change from S to SW)     |            |                                |                    |
| Maximum attenuation      | 4.41 (humic extract) | none                           | 2.25 (clay suspension + humic extract) |
| during presence of diesel |            |                                |                    |
| WAF                      |            |                                |                    |
| Maximum false enhancement| 10.6 (algae extract) | 3.89 (humic extract + algae extract) | 4.96 (algae extract) |
| during presence of diesel |            |                                |                    |
| WAF                      |            |                                |                    |

Water quality and sea state sensors were connected to a datalogger (Aanderaa SmartGuard 5300) that was programmed to operate the sensors with an 1-h measurement and data transmission interval. Real-time data from the SmartBuoy were transmitted over satellite modem to a server, where it was subject to an automatic data quality control before being visualized online.

The SmartBuoy buoy was moored on a junction of the main merchant shipping lanes in the Gulf of Finland (Figure 3) for a 2-month measurement period (October 25 to December 24, 2018). Vessels navigating on the north–south direction shipping lane between Helsinki and Tallinn were passing close to the buoy, and vessels on the east–west direction shipping lane passed the buoy further south. The SmartBuoy was visited once during the monitoring period for datalogger reprogramming and manual maintenance of the sensors.

3. RESULTS OF THE LABORATORY EXPERIMENTS

The produced WAF contained the following main PAH constituents (all µg/l): naphthalene (65), 1-methylnaphtalene (89), 2-methylnaphtalene (31), anthracene (1.6), acenaphthene (2.7), acenaphthylene (1.2), phenanthrene (1.2), fluoranthene (0.20), fluorene (4.4) and pyrene (0.2). Other PAHs fell below the 0.005 µg/l limit of quantification. The concentration of aliphatic compounds was 1,200 µg/l.

3.1. Effect of Interference Solutions

In the first laboratory experiment we investigated how the sensors reacted to the addition of interference solutions. In the absence of interference solutions, Turner C3 showed the largest absolute fluorescence values (primary data), followed by Enviroflu-HC and UviLux. The data from Turner C3’s PAH and CDOM channels had occasional spikes, but overall the data stability from each channel of the detectors were good. When WAF was added to the seawater, the relative change in signal strength (normalized response) was largest for UviLux (4.48-fold), followed by EnviroFlu-HC (2.62-fold) and Turner C3 (1.1-fold) (Table 1).

All sensors reacted when interference solutions were added to seawater. The addition of suspended clay to seawater slightly attenuated the PAH channel’s fluorescence signal in Turner C3 and Enviroflu-HC, while slightly enhancing it in UviLux (Figure 6, left half, S vs. SC). When suspended clay was added to the seawater-WAF solution, the signal was slightly attenuated in Enviroflu-HC and UviLux, and slightly enhanced in Turner C3 (Figure 6, right half, SW vs. SWC). In all combinations, the changes were at most 12% (0.88–1.04).

The sensors were more sensitive to the presence of humic substance extracts, with the PAH-signals increasing between 1.05- and 1.8-fold in all sensors (compared to seawater). The enhancement was strongest in Turner C3 and weakest in Enviroflu-HC (Figure 6, left half, S vs. SH). In the seawater-WAF solution the added humic extracts enhanced the PAH signal in Turner C3 (1.67-fold), while slightly attenuating the signal in the other sensors (Figure 6, right half, SW vs. SWH).

Clearly the largest effect on the PAH fluorescence signal was observed after adding algae extract. The signal was enhanced in all combinations of sensors and solutions. Importantly, the enhancement caused by the algae extract was stronger than the PAH signal of the WAF-solution without any interference solutions, which means that the presence of algae can interfere strongly with the sensors ability to detect oil compounds in the water. The enhancements when adding the algae extract to the seawater were roughly between 3- and 8-fold, with UviLux showing the largest change and Enviroflu-HC showing the smallest change (Figure 6, right half, SW vs. SWA). When the algae extract was added to the seawater-WAF solution, the PAH signal was enhanced roughly 2- to 3-fold in all sensors, with Turner C3 showing the strongest change in signal strength (Figure 6, right half, SW vs. SWA).

3.2. Diesel WAF Persistence and Method Comparison

In the second laboratory experiment we studied how long the sensors can detect oil compounds in seawater, while also comparing detection methods. Lagging software caused several gaps in the Turner C3 data and one gap in the Enviroflu-HC data. UviLux was relocated after 165 h when we noted that it did not react clearly to the third addition of WAF (WAF3; Figure 7), having large subsequent oscillations in the signal. This behavior was caused by optical interference by the two other sensors. Despite that there was substantial suppression in UviLux responses during the suppressed period, the sensor overall did show slight responses even during this period. These small responses occurred simultaneously when adding WAF 1 and WAF 3 (but not when adding WAF 2), and were also simultaneous with HELCOM and Enviroflu-HC responses (Figure 7). After UviLux was relocated, it provided data with
FIGURE 6 | Fluorescence responses from EnviroFlu-HC (TriOS), Turner C3 (Turner) and UviLux sensors in seawater, seawater containing interferences of clay, humic substances, algae substances, and seawater containing diesel WAF with and without the interferences. S = seawater, C = clay suspension, H = humic substance extract, A = cyanobacterial algae extract, and W = water accommodated fraction (diesel fuel extracted in seawater). PTSA = 1,3,6,8-pyrenetetrasulphonic acid tetrasodium salt. The change in FLD response respective to the baseline (S) value in each series (e.g., S–SC–SCH–SCHA) is indicated on top of each treatment.

a considerably higher response and less noise. For response calculations we therefore only used the UviLux data gathered after the relocation.

The responses from all sensors and the standard HELCOM oil monitoring method indicated that fluorescent compounds originating from diesel oil were detectable for at least 20 days since the start of the experiment (Figure 7). The temporal evolution in PAH-related responses measured by the EnviroFlu-HC and UviLux sensors were in line with the HELCOM method, especially after the readjustment of the sensors. In contrast, Turner C3 showed slightly increasing fluorescence between 100 and 505 h while the signal from the other two sensors decayed and leveled off.

Table 2 summarizes the responses obtained from sensor-based fluorescence right before water samples (GC methods) were collected at events S5 and S7. Event before sample S7 should serve as the primary point for response comparison since it was taken after the readjustment of sensor locations (R). The relative responses compared to initial state S1 of the fluorescence-based methods, during 165–505 h were (from strongest to weakest): HELCOM, UviLux, EnviroFlu-HC and Turner C3 (Figure 7 including responses at each S event).

Several PAH molecules were good indicators for diesel-originating contamination (Table 2). All napthtales provided pronounced responses also at S1 (strongest to weakest): 1-methylnaphtalene (1-MN), napthalene (NAF) and 2-methylnaphtalene (2-MN) (data not shown). Also, anthracene, acenaphthene and phenanthrene were sensitive and showed a similar temporal evolution as the FLD-based responses and the methylated naphttales (1/2-MN).

Concentration of fluoranthene did not change over the experimental period, and Dibenz[a,h]anthracene showed a declining trend compared to the initial seawater (Figure 8). These substances (in addition to acenaphthylene, fluoranthene, dibenz[a,h]anthracene and pyrene) were poor proxies for WAF contamination in the seawater (Figure 8).

The concentration of aliphatic hydrocarbons followed a similar pattern as those of FLD, methylated naphtalene and
the more sensitive PAHs. The PAH and aliphatic hydrocarbon concentrations also had a similar evolution pattern as the values derived using the HELCOM fluorescence method. Interestingly, the increase in aliphatic hydrocarbon concentrations between S8 and S9 showed a similar increase as the HELCOM method; this change was not visible in any other parameter (Figures 7, 8).

4. RESULTS OF THE FIELD EXPERIMENTS

4.1. FerryBox Experiment

The real-time transmission and online presentation of the FerryBox data worked well, but data gaps occurred systematically in the Baltic Proper due to loss of the GSM network. These gaps typically lasted a maximum of a few hours. An example is presented in Figure 4. The complete data was recovered from the datalogger after the fact. Altogether 58 ship voyages were analyzed. The PAH-transects were visually checked for peaks that would have indicated an oil pollution. No such anomalies were found during the 2-month measurement period.

The detected PAH values stayed between 1 and 2.6 µg carbazole/l for the UviLux and 12.4–25.5 µg phenanthrene/l for the EnviroFlu-HC (Figures 9A,B). The detected hydrocarbon values were lowest in the sea areas near Åland and highest in the Stockholm archipelago, where the effect of the natural background of organic carbon from river waters is greater (Figures 9A,B). From 20.5°, longitude eastward, the PAH values were quite homogeneous. Some variation can be seen e.g., between 22°, and 23° longitude (trips 19–23), showing up more clearly in the UviLux data (Figures 9A,B). We surmise that the detected values do not reflect concentration of carbazole, phenanthrene or oil in seawater. This interpretation is supported by the chemical monitoring (HELCOM/EU MSFD) and analysis of the Baltic Sea, which has found concentrations below 1 µg oil hydrocarbons/l rather constantly (Pikkarainen and Lemponen, 2005; FIMR and Olsonen, 2007). These low values are considered to be typical for seawater without significant oil pollution (Bícego et al., 1996; Zanardi et al., 1999).
TABLE 2 | Overview on the response levels generated by the fluorescence-based methods and parameters obtained from GC-FID and GC-MS analyses.

| Parameter | Background level $\mu g \; l^{-1}$ | Response at S5 $\mu g \; l^{-1}$ | Response relative to background level at S5 (-fold) | Response at S7 $\mu g \; l^{-1}$ | Response relative to background level at S7 (-fold) |
|-----------|----------------------------------|----------------------------------|---------------------------------------------|----------------------------------|---------------------------------------------|
| Turner C3 fluorescence (PTSA) | 634.7 ± 3.74 | 1071.7 ± 2.23 | 1.68 | 1084.14 ± 2.85 | 1.71 |
| TriOS Envirolu-HC fluorescence (phenanthrene) | 7.22 ± 0.41 | 82.63 ± 0.56 | 11.45 | 63.4 ± 1.27 | 8.79 |
| UviLux fluorescence (carbazole) | 0.123 ± 0.003a | 0.819 ± 0.003a | 6.65a | 2.351 ± 0.002 | 19.10 |
| HELCOM fluorescence | 0.42 | 61 ± 9.15b | (±< 0.1) | 145.2 | 27.27 ± 4.09 | 64.92 |
| Aliphatics C10–C40 | 11 | 225 ± 6.65 | 2.351 | 145.2 | 27.27 ± 4.09 | 64.92 |
| Aliphatics C22–C40 | 120 | 1700 ± 510 | 4.33 | 410 ± 123 | 1.025 |
| Total PAHs | 0.988 ± 0.2964a | 9.39 ± 2.82 | 9.50 | 1.62 ± 0.63 | 1.63 |
| Napthalenef | 0.11 ± 0.02 | 1.7 ± 0.34 | 15.45 | 0.2 ± 0.04 | 1.82 |
| 1-methynapthalene | 0.07 ± 0.02 | 2.9 ± 0.87 | 41.4 | 0.25 ± 0.08 | 3.57 |
| 2-methynaphthalene | < 0.005 | 3.5 ± 1.05 | 700a | 0.17 ± 0.05 | 34a |
| Anthracene | < 0.01 | 0.10 ± 0.03 | 9.7 | 0.11 ± 0.04 | 11 |
| Acenaphthylene | 0.084 ± 0.017 | 0.21 ± 0.04 | 2.5 | 0.098 ± 0.02 | 1.17 |
| Acenaphthylene | < 0.005 | < 0.005 | NA | 0.087 ± 0.026 | 5.22 |
| Fluoranthene | 0.16 ± 0.05 | 0.16 ± 0.05 | 1 | 0.16 ± 0.05 | 1 |
| Fluorene | 0.14 ± 0.028 | 0.38 ± 0.08 | 2.71 | 0.18 ± 0.04 | 1.29 |
| Dibenzo[a,h]anthracene | 0.094 ± 0.028 | < 0.01 | 0.11a | < 0.01 | 0.11a |
| Phenanthrene | 0.22 ± 0.066 | 0.32 ± 0.066 | 1.45 | 0.24 ± 0.072 | 1.09 |
| Pyrene | 0.11 ± 0.03 | 0.12 ± 0.04 | 1.09 | 0.12 ± 0.04 | 1.09 |

a Suppressed signal. b ± 15% error attributed for the method. c Actual repeatability of parallel samples at S5 and S7. d Based on ±30% error attributed for the method. e Based on ±50% error attributed for the method. f Based on ±20% error attributed for the method. g Calculation of relative difference is based on the limit of quantification. h Based on ±35% error attributed for the method. NA = not available.

The FerryBox and the accompanying sensors were maintained and cleaned frequently. Nevertheless, biofouling still impacted the quality of the measured data as can be seen from the decreasing values between the maintenance visits (Figures 9A,B). The field data suggests that Enviroflu-HC is affected more strongly by the fouling than UviLux.

4.2. SmartBuoy Experiment

The obtained values ranged between 790 and 1,250 during the monitoring period. The variation pattern was nearly identical with the variation of collected colored organic carbon (CDOM) values, varying between 600 and 890 (Figure 9). The measured data had systematic gaps. On the 12th of December a maintenance visit to the boy was made to reprogram the datalogger and manually clean the sensors.

The correlation between the collected hydrocarbon dataset and the CDOM data set collected by the C3 sensor on SmartBuoy was very high (Figure 9D). No anomalous spikes indicative of oil-contamination were present during the test period. The C3 sensor on SmartBuoy was equipped with a mechanical wiper, which cleaned all three optical lenses prior to measuring. No significant reduction in data quality due to biofouling was detected compared to the FerryBox where the detected values gradually decreased after the cleaning (Figure 9). The linear relationship between the two variables changed after the maintenance on 12 December. This change was probably caused by the readjustment of the mechanical wipers (Joose Mykkänen, Personal communication). Nonetheless, the correlation between the variables remained high (Figure 9D).

During the experimental period, the buoy tolerated several events with significant wave heights over 2 m. The maximum significant wave height reached almost 3 m, which is estimated to be exceeded at this location roughly 1% of the times (Tuomi et al., 2011; Björkqvist et al., 2018). During the monitoring period the current speed reached 39 cm/s, with the mean value being 12 cm/s. The mean yearly current speed (for depth 0–7.5 m) in the area is expected to be under 10 cm/s (Westerlund et al., 2018).

5. DISCUSSION

Laboratory tests showed that optical interferences strongly affected optical PAH detection. The effect of these interferences was corroborated by the strong correlation between oil-related fluorescence and CDOM fluorescence (Figure 9D). CDOM causes interference and false positives whenever it is present sufficiently and not accounted for properly. Similar results arose from the presence of cyanobacteria-derived material. The Baltic Sea contains substantial concentrations of CDOM at any time of year, most prominently close to river outlets. The quenching of
Extensive spring and summer blooms of phytoplankton occur regularly in the Baltic sea (Kahru et al., 2007, 2016; Groetsch et al., 2014). We can also suspect greater interference during summer blooms of cyanobacteria and minor contribution throughout the year as there are always small quantities of chlorophylls and phycocyanin of phytoplankton origin in the surface waters. The spring bloom duration in the Baltic Sea is about one and a half months (Groetsch et al., 2016) and the peak of the bloom in 2018 coincided with the end of the FerryBox experiment (Almén and Tamelander, 2020). Nevertheless, we did not see the signal enhancing effect of the phytoplankton during the FerryBox experiment, as the PAH signals should have increased toward the end of the testing period together with the bloom intensity.

There also exist several overall issues when using fluorometers that are related to the information about the presence and relative concentration of oil compounds in water. Fluorometers are generally calibrated using a specific oil or specific compounds; thus, the relevant concentration results are relative to the specific oil or compound and the procedure used to calibrate the instrument (Lambert et al., 2003). The response of the fluorometer to oil depends significantly on the oil composition and its weathering state, complicating the quantification of oil concentration further (Henry et al., 1999). Also other studies have noted that the other fluorescent substances in seawater can significantly interfere with direct fluorescence measurements of petroleum hydrocarbons (Henry et al., 1999; Bugden et al., 2008; Tedetti et al., 2010; Cyr et al., 2019). Lastly, a reliable comparison of different sensors would also require a consensus over the units used to report FLD-based results.

Taking account of the aforementioned issues and the knowledge that most of the oil pollution in the Baltic Sea comes from small operational oil spills, and that most cases requiring
FIGURE 9 | Heatmap plots showing spatiotemporal distributions of PAH concentrations in seawater along the ship track between Stockholm and Tallinn (February 21 to April 21, 2018) measured with the UvL Lux (A) and EnviroFlu-HC (B) fluorometric sensors. A temporal heatmap of the SmartBuoy data (October 25 to December 24, 2018) (C) and correlation between CDOM and hydrocarbon data before (black) and after (red) the maintenance (D). Red lines on the heatmaps indicate system maintenance visits.

criminal prosecution have involved diesel oil, the latter was selected to mimic the oil pollution in our laboratory experiments. Moreover, based on this background and the results of the laboratory tests, the field trials concentrated only on the detection of oil pollution, which should have appeared as a clear peak, deviating from the background fluorescence in the measured response patterns. The question of capabilities to detect different types of oil in the field is not only a question of the detection capabilities of the actual sensors, since also the weathering process for e.g., crude oil and diesel oil differ. In addition, the calibration of the sensors can reflect the a priori expectations of what type of oil might be encountered. It is therefore impossible to give any general recommendations on this subject, and follow-up studies are needed to test the suitability of the sensors used in this study in case they are planned to be used outside the Baltic Sea.

Due to the interfering substances it is difficult to distinguish water quality variations from the responses from oil spills. To combat that problem, suitable mathematical protocols should be explored. Some newer sensors also include built-in correction methods to discriminate oil from the natural interferences. Examples of sensors with such specification are Vlux OilPro (Chelsea Technologies Ltd), SeaOWL UV-A™ (Sea-Bird Scientific) and HM-900 (Pyxis).

Another aspect that needs careful consideration is the measurement or water intake depth of the oil detection systems. After a spill, wind, waves and currents can break the oil spill into droplets and may propel the oil deeper into the water column; this process is called natural dispersion. Dispersion is a complicated physio–chemical process affected by the characteristics of the oil spill (oil type, density, viscosity, thickness) and other seawater properties such as temperature and salinity (Xiankun et al., 1993; Papadimitrakis et al., 2011). Various other physical, chemical, and biological processes begin to alter the oil as well, altogether referred to as weathering process (Mishra and Kumar, 2015; Tarr et al., 2016).

Using sampling and further laboratory analysis the maximum detection depths of naturally dispersed oils has been between 2.5 and 15 m (Cormack and Nichols, 1977; Lichtenthaler and Daling, 1983, 1985). An experiment by Cormack and Nichols (1977), part of which was monitoring naturally dispersed Ekofisk crude oil in the water column, showed comparatively rapid decrease in oil concentration down to 5 m depth, followed by a slower decrease to background levels at 20 m. A similar experiment in Norwegian waters with 10 m³ of topped Statfjord crude oil showed that small concentrations of oil were present to at least a depth of 3 m in the water column under the slick (Lichtenthaler and Daling, 1985). In another experiment done with 2,000 liters of Statfjord
crude oil and of topped Statfjord crude oil the maximum detection depths were 2.5 m (Lichtenthaler and Daling, 1983).

At the Baffin Island Oil Spill experimental site in the Canadian Arctic, flow-through fluorometry was successfully used to monitor a subsurface release of chemically dispersed crude oil cloud over several days, providing real-time and continuous data on oil concentrations in the water column (Green et al., 1983). Nonetheless, during a surface release experiment performed in the same study, petroleum hydrocarbons from the untreated oil were not detected in the water column deeper than 1 m (Humphrey et al., 1987). In a contained oil spill experiment, part of which was 3 liters of crude oil spilled in moored 66 m³ containers in the ocean, a continuous flow-through fluorescence system was capable of detecting oil concentrations down to 8 m, over a period of 7 days (Green et al., 1982). Lunel (1995), using continuous-flow fluorometry (calibrated using discrete samples) found traces of oil down to 5 m after 20 metric ton of a mixture of medium fuel oil and gas oil (in a 50–50 ratio) were experimentally released in United Kingdom waters. Nevertheless, these experiments made in U.K., the Canadian Arctic and inside the British Columbian archipelago might not be directly transferable to the Baltic Sea because of differences in hydrography, and wind and wave conditions.

The probability of detecting oil pollution clearly depends on the weather conditions, and the size, oil type, source and weathering state of the spill. Most of the FerryBox systems are limited to a fixed depth set by the ship’s design, as is the case in this study. For the FerryBox system on M/S Baltic Queen, the water is taken in at a depth of approximately 4 m and the SmartBuoy sensors situate 2–3 m below the surface, depending on the sea level. In light of the results of the aforementioned studies, these depths should be suitable for detecting anomalies in fluorescence. Also, for the FerryBox the water will also be mixed by the moving ship itself. Nonetheless, when devising such systems in the future, measurement depths as close to the surface as possible are recommended to ensure higher chance of spill detection.

Biofouling (marine growth) is likewise a major factor limiting the reliability of optical sensors in aquatic studies, especially during long term measurements. Biofouling is the net result of various physical, chemical and biological factors such as water temperature, conductivity, season, and location—to name a few (Delauney et al., 2010). Several antifoulant approaches for optical systems on autonomous platforms have been suggested (Manov et al., 2004; Delauney et al., 2010). In our experiments the automatic mechanical cleaning of the sensor on the SmartBoy was satisfactory but the fouling of the FerryBox sensors after the maintenance visits is clearly there. It is not evident how much biofouling affects the oil detection capabilities of the sensors, but in our experiment is seemed to be of secondary importance compared to the presence of the other interfering substances.

6. CONCLUSIONS

Laboratory experiments indicated that the sensitivity of the sensors to diesel oil is good and they provide useful data on oil in seawater. Fluorescent compounds from the oil were detectable by the fluorometric sensors for at least 20 days. Our laboratory tests were conducted with diesel oil, Baltic Sea water, and local interfering substances and therefore may not be representative of different types of oils or marine areas. The main issue with the sensors we used was their specificity, since the presence of humic substances (CDOM), phytoplankton (phycocyanin and chlorophylls) or high turbidity (suspended/colloidal clay) can cause false positives or signal quenching when detecting oil. In our tests the impact of algal material was clearly the most significant. We suggest exploring mathematical protocols or use of more sophisticated sensors for distinguishing actual oil pollution from co-occurring optical interference.

The three tested portable fluorometers were successfully integrated to the FerryBox and SmartBuoy systems. The systems functioned semi-autonomously well during the 2-month testing periods, as did the real-time data transmissions and user interfaces/data visualization. The signal quenching effect of the biofouling could be seen on the FerryBox fluorometers, but it was a secondary problem compared to the interfering substances in the seawater. Nonetheless, automatic cleaning methods of the systems and sensors are recommended.

In this paper we did not touch upon the automatic detection of the anomalic events in the detected values that would indicate an oil spill, but this kind of an algorithm should be developed if the detection system is to be used operationally. When these systems can reliably detect oil, a comprehensive network of SmartBuoy and FerryBoxes covering the major fairways can greatly complement the aerial surveillance, ship-based and satellite monitoring that are presently used to, among other things, detect oil in seawater.

DATA AVAILABILITY STATEMENT

The raw data supporting the conclusions of this article will be made available by the authors, without undue reservation. The data received from the real-time transmission of the FerryBox system can be viewed on a public web-site (TalTech, 2017).

AUTHOR CONTRIBUTIONS

SP: investigation, validation, formal analysis, writing—original draft, and visualization. HK: methodology, investigation, formal analysis, writing—original draft, and visualization. J-VB: validation, writing—review and editing, and supervision. RU: writing—review and editing and supervision. All authors contributed to the article and approved the submitted version.

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