

## Space-Time Cluster Analysis of Accidental Oil Spills in Rivers State, Nigeria, 2011–2019

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**Abstract:** Oil spills cause environmental pollution with a serious threat to local communities and sustainable development. Accidental oil spills can be modelled as a stochastic process where each oil spill event is described by its spatial locations and incidence-time and hence allow for space-time cluster analysis. Space-time cluster analysis can detect space-time pattern distribution of oil spills which can be useful for implementing preventive measures and evidence-based decision making. This study aims to detect the space-time clusters of accidental oil spills in Rivers state, Nigeria through the Space-time Scan Statistic. The Space-time Scan Statistic was applied under the permutation model to the oil spill data (each for sabotage and operational oil spills) collected at Local Government Area (LGA)-level during the period from 2011 to 2019. The results show that the sabotage oil spill clusters have covered most of the LGAs in the southern part of the state at the start of the study period and then in 2018–2019, it moved to the west covering a single LGA. The operational oil spill clusters covered two neighboring LGAs in the south. In addition, the temporal cluster of sabotage oil spills was seen in 2019 and operational oil spills in 2011–2012. The sabotage oil spills show an increasing trend with the maximum in 2019 while the operational oil spills show a decreasing trend with the minimum in 2019. These findings assist in more effective decision-making for combating the environmental problems and controlling the future spill incidence in the cluster-regions.

**Keywords:** Oil spills; space-time clusters; SaTScan<sup>TM</sup>; permutation model; Nigeria

### 1 Introduction

Oil spills are frequently occurred in oil exploration, refining, and processing units [1], and thus exposing the nearby societies to environmental pollution [2]. The oil spills are mainly caused by operational failure (pipeline erosion, production, poor infrastructure and maintenance, faults during oil processing phases),



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third-party interference (attempts of thievery or intentional sabotage) [3,4]. Oil spills have an impact on soil and water, and consequently on the health of resident population and agricultural productions [5]. Toxic contaminants (e.g., polycyclic aromatic hydrocarbons and heavy metals) contained in the oil, seep into the soil, reach the groundwater and hence damage the vulnerable ecosystems [6,7].

Africa and especially Nigeria has suffered unprecedented pipeline interdictions. The Shell Petroleum Development Company (SPDC), Nigeria has been recording an average of 200 oil spill events each year since 2005 [8]. About 75% of the oil spills events during 2007–2013 were attributed to sabotage/theft [8]. Indeed, the presence of gangs in some communities sabotages pipelines in order to impede operations of oil companies, while claiming compensation for the farmlands or rivers contaminated by oil spills [9].

Although several studies have focused on oil spills detection and mapping [10–14], environmental, health and socioeconomic impacts of oil spills [15–21], and geospatial analysis of oil spill distribution [22,23] in different parts of Nigeria. However, space-time cluster analysis of oil spills at LGA-level seems to be lacking. The availability of spatially and temporally stamped information on oil spills in Nigeria allows for space-time cluster analysis of oil spill occurrence in the country. Spatiotemporal cluster analysis can detect important patterns in oil spill occurrence which could be helpful for oil spill management and environmental protection authorities in evidence-based interventions. In addition, it assists the officials of the oil industry in proper allocation of limited resources for quick response, cost and benefits distribution, exposure mitigation, and foster public-level involvement for sustainable development.

In this context, the main objective of this study is to identify the significant space-time clusters of oil spill events in the Rivers state, Nigeria during the period from 2011 to 2019 using Space-time Scan Statistic [24,25]. The Space-time Scan Statistic is implemented in the freely available software called SaTScan™ [26]. It has been widely used in environmental studies such as clusters detection of forest fire events [27–29], and pollution [30]. It has the additional advantage of detecting the cluster's location and time-frame, while also evaluating its statistical significance as recommended for environmental studies by Tonini et al. [31] and Tuia et al. [32]. A useful functionality of SaTScan™ to assist decision making, is the integration of the model outputs and a geographic information system, for a map-visualization of the oil spill-prone regions. SaTScan™ is a group of methods, applied in many domains to search for spatial, temporal, and space-time clusters. It was first developed for health sciences but nowadays it is used in a variety of fields such as environment [27,28], road traffic [33,34], criminology [35], and public health [36].

## **2 Materials and Methods**

### **2.1 Study Area**

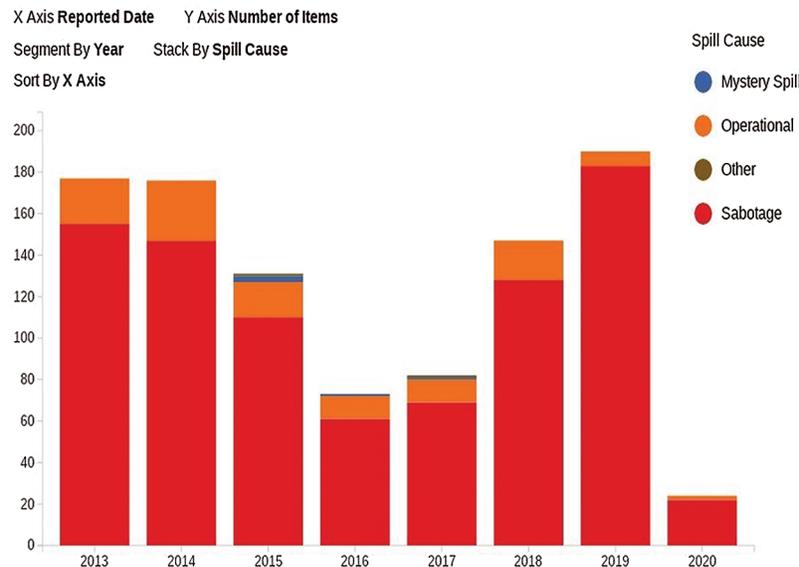
This study was carried out in the Rivers state located in the eastern part of the Niger Delta region of Nigeria. The total area of the Rivers state is 11,077 km<sup>2</sup>, making it the 26th largest state of the country [37]. The total land area of the state is divided into twenty-three Local Government Areas (LGAs). It is surrounded on the north by Imo, Abia and Anambra state, on the east by Akwa-Ibom and Bayelsa states, on the west by Delta state and on the south by the Atlantic Ocean. Its geography ranges from plains, with a network of big rivers to tributaries. The inland area of the state is covered by tropical rainforest and the coastal area by mangrove swamps. The oil exploration activities in this area have severely affected the economic, cultural, and daily living conditions of the local communities [38].

### **2.2 Oil Spill Data Collection**

The yearly data on oil spills in each LGA of the Rivers state, Nigeria were collected from the database of SPDC Nigeria [39]. SPDC Nigeria, in collaboration with government officials, quickly responds to spill incidents and cleans up spills from its facilities. SPDC Nigeria provides the monthly record of oil spills

publicly on its website under the sustainability report to enhance the transparency on spills in SPDC Joint Venture (SPDC-JV) facilities. The SPDC records the location, time, and cause of each oil spill incidence in the joint investigation report which is available publicly on its website [39].

This study analyzes the oil spills due to each sabotage and operation separately, because these are the major causes of oil spills in Nigeria as shown in Fig. 1. Two datasets each for sabotage and operational oil spills were collected.



**Figure 1:** Oil spills graph in Nigeria [39]

### 2.3 Space-Time Scan Statistic

The Space-time Scan Statistic is characterized by a cylindrical scanning window with a circular base. The circular base of the scanning window is related to the spatial size and the height to the temporal length of the cluster. The cylindrical window visits each location as well as each time point (day, month, or year). The size of the cylinder continuously increases from zero up to a fixed value in both radius (space) and height (time). Hence, a large number of overlapping cylinders of different sizes are obtained that mutually cover the total study-area. Each of the cylinders denotes a possible candidate cluster.

Events are supposed to be distributed with a Poisson model with a constant risk inside and outside all cylinders under the null hypothesis, and with different risk inside and outside at least one of the cylinders under the alternative hypothesis. For each possible cylinder, the observed and expected numbers of events inside and outside the cylinder are calculated. Based on these numbers, the likelihood is calculated for each cylinder and the one with the maximum Log-Likelihood Ratio (LLR) is considered to be the most likely cluster. The other cylinders with a higher LLR and with statistical significance are considered as the secondary clusters. The statistical significance of the detected clusters is evaluated by a Monte-Carlo simulation [40].

The Poisson model requires a control-population for estimating the expected number of events inside each cylinder. However, for scenarios like ours, where the control population is not relevant or unknown, this problem is solved by using the Space-time Scan Statistic Permutation (STSSP) model [24]. This model requires the input data on observed events only. Thus, the number of expected events is estimated on the basis of the observed events under the assumption of no space-time interaction in observed events.

For instance, for a cylinder  $A$ , the expected number of events  $E_A$  can be calculated as the sum of  $\mu_{ij}$  (the expected counts for  $i^{th}$  sub-region and  $j^{th}$  time-point) belonging to cylinder  $A$  as in Eq. (1).

$$\mu_A = \sum_{i,j \in A} \mu_{ij} \quad (1)$$

where  $\mu_{ij}$  can be calculated as in Eq. (2)

$$\mu_{ij} = \frac{\sum_i C_{ij} \times \sum_j C_{ij}}{C..} \quad (2)$$

where  $E_{ij}$  is the expected events for  $i^{th}$  sub-region over the  $j^{th}$  time-point;  $C_{ij}$  is the observed/reported events in the  $i^{th}$  sub-region at the  $j^{th}$  time-point;  $C..$  is the grand total of the observed/reported events.

Based on the observed and expected events, the Poisson Generalized Likelihood Ratio can be calculated according to Eq. (3).

$$GLR = \left(\frac{c_A}{\mu_A}\right)^{c_A} \left(\frac{C.. - c_A}{C.. - \mu_A}\right) \quad (3)$$

where  $c_A$  is the observed/reported count in a cylinder,  $A$  is considered to follow the Poisson-distribution with mean,  $\mu_A$  [31].

The GLR is computed and maximized for each possible cylinder and the statistical significance is tested by Monte Carlo simulations.

The STSSP model is a very useful statistical tool to analyze environmental data [29,31,32]. Its main advantage is that it requires input data on the observed events instead of both: observed events and the control-population. In the context of oil spill events, the control-population is not relevant, limiting the application of other models in SaTScan<sup>TM</sup>. Therefore, we used the STSSP model for detecting space-time clusters of oil spills in the Rivers state, Nigeria. The STSSP model was applied with the circular shape and with the default setting of maximum spatial and temporal window sizes [40]. The Monte-Carlo method was used to test the statistical significance of the clusters. The clusters having a  $P$ -value of less than 0.05 were considered to be the significant clusters.

### 3 Results and Discussion

#### 3.1 Sabotage Oil Spill Clusters

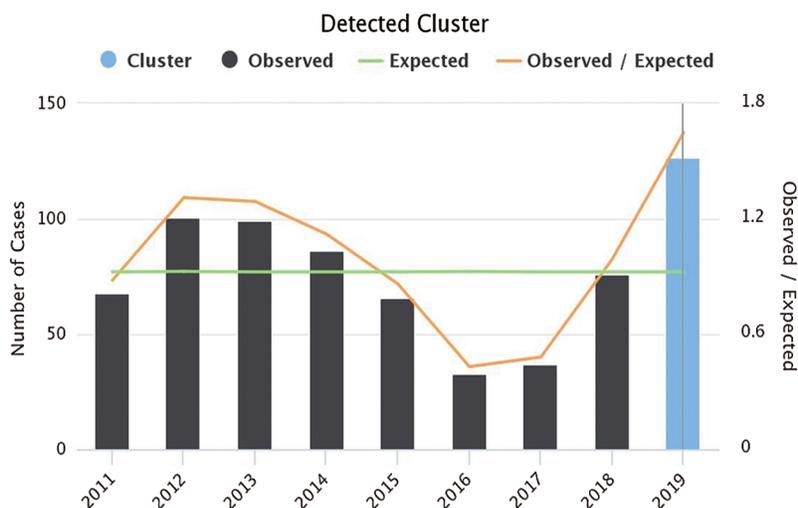
The Space-time Scan Statistic under permutation model detected three significant clusters of sabotage oil spills in the Rivers state, Nigeria during the year 2011 to 2019 as given in Tab. 1).

**Table 1:** Space-time clusters of sabotage oil spills in Rivers state, Nigeria, 2011–2019

Cluster type	LGAs	No. of LGAs	Time-frame	Observed counts	Expected counts	LLR	$P$ -value
Most Likely	Ahoada-East	1	2018–2019	24	7.03	12.71	<0.001
1 <sup>st</sup> secondary	Bonny, Andoni, Gokana, Ogu-Bolo, Opobo-Nkoro, Okrika	6	2013–2015	92	61.57	7.26	0.002
2 <sup>nd</sup> secondary	Asari-Toru, Degema	2	2011–2013	26	11.99	6.26	0.01
3 <sup>rd</sup> secondary	Etche, Ikwerre	2	2012–2012	26	14.14	4.083	0.176



temporal graph shows that the sabotage oil spills have increased from 2011 to 2012, declined from 2012 to 2016 to the minimum, and then increased from 2016 to 2019 to the maximum (Fig. 3). The increasing trend in the later year may be associated to the surveillance efforts that might be less consistent than in previous years.



**Figure 3:** Temporal cluster of Sabotage oil spills in Rivers state, Nigeria

### 3.2 Operational Oil Spill Clusters

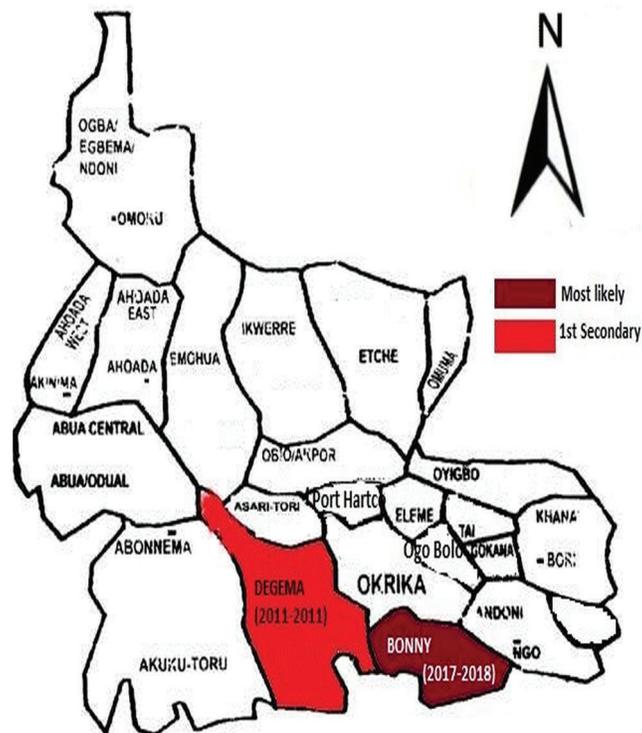
In the operational oil spill dataset, the Space-time Scan Statistic under the permutation model identified a total of two significant space-time clusters (Tab. 2). The most likely cluster was seen in Bonny LGA for two years 2017–2018, and the secondary cluster in Degema for one-year 2011. These two clusters are statistically significant ( $P$ -value  $< 0.05$ ). Both operational oil spill clusters occurred in the southern part of the Rivers state covering each a single LGA. These clusters have disappeared before the end of the study period and no present cluster of operational oil spills was seen at the end of the study period.

**Table 2:** Space-time clusters of operational oil spills in Rivers state, Nigeria, 2011–2019

Cluster type	LGAs	No. of LGAs	Time-frame	Observed counts	Expected counts	LLR	$P$ -value
Most likely	Bonny	1	2017–2018	8	2.24	4.59	0.032
1 <sup>st</sup> secondary	Degema	1	2011–2011	15	6.53	4.41	0.041
2 <sup>nd</sup> secondary	Ikwerre, Etche, Emohua	3	2016–2016	5	1.00	4.13	0.072
3 <sup>rd</sup> secondary	Eleme, Tai, Okrika, Ogu-Bolo, Port Harcourt, Obio-Akpor, Oyigbo, Gokana	8	2012–2013	12	5.51	3.08	0.313

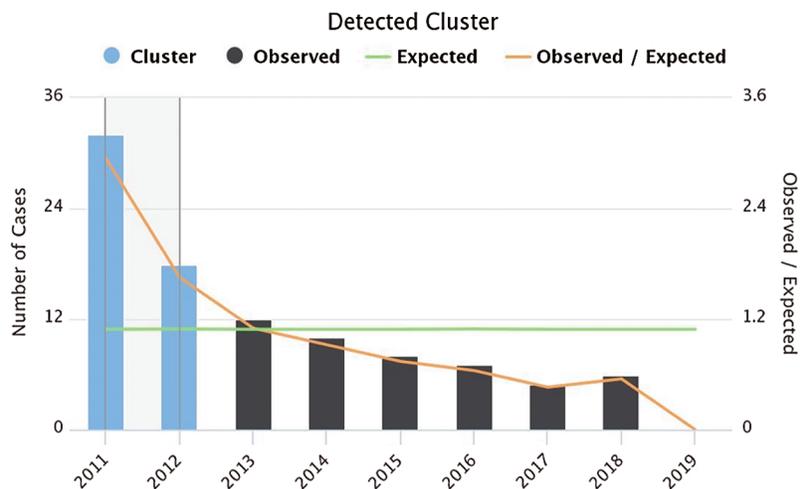
The geographic sites of the significant clusters of operational oil spills in the Rivers state are shown in Fig. 4. The geographic sites in Fig. 4 show that during the first year of the study period (2011), the cluster of operational oil spill occurred in one LGA in the south, and then in 2017–2018, it moved toward the

neighboring LGA in the East. Similar to the sabotage oil spill clusters, the operational oil spill clusters are not geographically persistent.



**Figure 4:** Locations of operational oil spills clusters in Rivers state, Nigeria

The resulting temporal graph showing the temporal cluster of operational oil spills is given in Fig. 5. The period (2011–2012) was detected as a potential temporal cluster of operational oil spills in the study area. The temporal graph shows a decreasing trend in operational oil spills from 2011 to 2019 with a maximum in the year 2011 and minimum in 2019. This substantial decrease in the operational oil spill might be due to the efficient control strategies for operational oil spills in the study area.



**Figure 5:** Temporal cluster of operational oil spills in Rivers state, Nigeria

Our data analysis does differentiate the temporal and space-time clusters where oil spill accidents were frequent. The space-time cluster analysis shows that in both datasets (sabotage and Operational oil spills) most of the potential clusters occurred in the southern LGAs. However, the temporal graphs show opposite trends in both datasets (Figs. 3 and 5). The operational oil spills seem to be controlled while controlling the sabotage oil spills in the study area is still a challenge.

#### 4 Conclusion

The Space-time Scan Statistic approach detected temporaneous hotspots of oil spills in the Rivers state, Nigeria, thus provides clues for policymakers for better designing and targeting oil spill pre-suppression and control facilities at the local level. The results of the space-time cluster analyses provide useful information regarding the characteristics and occurrence-structure of the oil spill clusters in the study area. Most of the space-time clusters (sabotage and operational) have occurred in the southern part of the state. However, one potential cluster of sabotage oil spill occurred in the western LGA (Ahoada West) which is still present at end of the study period highlighting a current oil spill pollution hotspot for an immediate intervention of the oil spill control and environmental protection authorities. Similarly, the temporal graph of the sabotage oil spill also shows a tremendous increase at the end of the study period alerting the oil spill control and environmental protection authorities for taking preventive measures. Moreover, the detection of oil spill clusters could be a starting point for the deeper analysis of observing the social and environmental factors of oil spills in the study area.

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**Conflicts of Interest:** The authors declare no conflicts of interest.

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