

The Possibility of COVID-19 Pandemic in Eliminating Burning Activities: A Case Study at Ogan Komering Ilir Regency, South Sumatera

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Abstract

Forest and land fires occur almost every year in South Sumatera Province, including at Ogan Komering Ilir (OKI) Regency, mainly due to uncontrolled burning activities. This region has the largest peatland responsible for the adverse haze impacts. The advent of the coronavirus (COVID-19) pandemic in early 2020 has triggered massive consequences across the global communities, including Indonesia. Therefore, large-scale regulations on social restrictions were enacted. The purpose of this study is to analyze the variations in hotspots as forest and land fire indicators before and during the COVID-19 pandemic in OKI as well as to determine the rainfall effects. Daily Terra/Aqua MODIS satellite feeds and rainfall data between January 2018 and December 2020 served as the research materials. Subsequently, the paired t-test and correlation assessment were used to examine the hotspot variations in both datasets, respectively. The results showed significant statistical differences before and during the pandemic. Consequently, social restrictions were assumed to instigate the decline in burning activities. Furthermore, the rainfall demonstrated a vulnerable correlation to the hotspots, indicating that human factor was more pronounced as a fire trigger.

Keywords: haze, fire detection, Indonesia, burning activities

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Introduction

Forest and land fires are common occurrences in South Sumatera Province. This region is among the fire-prone locations in Indonesia (KLHK, 2018) with the highest hotspots and burned area (Ardiansyah et al., 2017) and widespread forest fire regions (Indratmoko & Rizqihandari, 2019), covering 336,798 ha of burned area in 2019 (KLHK, 2021). Based on Terra/AQUA MODIS satellite data, the greatest hotspot is situated in Ogan Komering Ilir (OKI) Regency, also known as the most affected zone (Indratmoko & Rizqihandari, 2019), with peak periods in dry seasons, particularly between August and September. However, the fires are known to have instigated haze alongside a significant impact on public health and are strongly related to the peatland incidence where huge carbon materials are released during the burning. These substances combine with water vapor producing thick smoke haze containing aerosols, particulate matter, and other trace gases which are hazardous to human health and the surrounding environments (Fujii et al., 2014; Heriyanto et al., 2015; Jayarathne et al., 2018; Kiely et al., 2019; 2020).

South Sumatera encompasses a peatland area of approximately 1.2 million ha or 14.5% of the total region. The province is a vital natural resource with potentials for community livelihood. Among the peatland area distribution, OKI demonstrated the maximum coverage of approximately 640,647 ha (50%) (BBSDL, 2011).

Currently, certain segments appear degraded, unproductive, and unsustainably managed with limitation to agriculture applications (Rosanti, 2014). However, the burnt peatland recorded severe impacts, compared to non-peatlands. The fires are triggered by land drying during land preparation and other uses as peat is a good fuel with irreversible drying characteristics (Usup, et al., 2004). Also, the peat dryness is expressed by the moisture content, due to rainfall effects (Putra, 2011). Furthermore, most of the burnt areas were outside the concession, while internal session showed particular typologies, including the fire activities in HTI (industrial forest plantation) on non-forested (26%) and forested land (24%), as well as in oil palm on non-forested (17%) and forested land (2%) (Ardiansyah et al. 2017).

Forest and land fires are monitored and detected using hotspot numbers and satellite data distributions. Previous studies show that hotspot data are employed as the key indicator of fire occurrence (Nurdiana & Risdiyanto, 2015; Usman et al., 2015), particularly clustered and sequential hotspots (Kirana et al., 2015; Wijaya et al., 2016; Syaufina & Sitanggang, 2018). The Ministry of Forestry and Environment based on the provisions of Number P.8/MENLHK/SETJEN/KUM.1/3/2018, defines a hotspot as a location with the temperature above a certain threshold from remote sensing data, used as a forest and land fire indicator.

Generally, fire occurrences in Indonesia are caused by

human factors both intentionally and unintentionally (Dennis et al., 2005; Syaufina, 2018). The intention is common for land preparation purposes, particularly in the approved cultivation areas (HGU). Fire in peatlands is mainly caused by burning activities (Pasaribu & Friyatno, 2012; Syaufina, 2018). The abandoned farmland typically has a 2–3 m peat depth and is located in the downstream watershed (Pasaribu & Friyatno, 2012). Human activity plays a very important role in causing a number close to 100% of forest and peatland fires. Apart from intentionally burning, climate change was also suspected as a significant wildfire source, with several indicators including increasing temperature, decreasing humidity, and low precipitation (Afifuddin et al., 2019).

Indonesia was first hit by the coronavirus (COVID-19) pandemic in early 2020. The RNA virus possesses a particle size range of 120160 nm. This disease mainly infects animals, including bats and camels (Riedel et al., 2019). Consequently, the World Health Organization announced COVID-19 as a global pandemic on March 12, 2020, and in subsequent weeks, precisely on March 29, the confirmed cases had attained 634,835, with 33,106 deaths recorded worldwide. Meanwhile in Indonesia, 1,528 positive cases and 136 deaths were reported. Therefore, to minimize the rapid spread, the Indonesian government issued a regulation (Number 21) on large-scale social restriction (PSBB) in 2020 (Nurhalimah, 2020).

This study aims to determine the possibility of social restrictions in the burning prevention in OKI. These activities are approached by hotspot variations as indicators of forest and land fires before and during the COVID-19 outbreak. In addition, the PSBB enactment is assumed to decrease the triggers of forest and land fire. However, studies on the relationship between the COVID-19 pandemic and the fire appear very limited. Therefore, the purpose of this research is to analyze the hotspot differences prior and during the COVID-19 pandemic, in addition to determining the rainfall influence.

The present study is expected to provide valuable information on burning activities at the research area, before

and during the COVID-19 pandemic, with potential usefulness as a reference for forest and land fire control in the selected region.

Methods

Study area Figure 1 shows the position of OKI between the ranges of latitudes S2°30'–S4°15' and longitudes E104°20'–E106°00'. This location is among the regencies in South Sumatera Province, Indonesia, with an average elevation of 10 m asl. The name originated from two main rivers in the region: Ogan and Komerling, with relative influence on the dominant swamps. Furthermore, the regency is also located in both floodplains covering a total area of 19,023.47 km², where 7,690 km² (40.42%) forms the peatland area prone to the fire source.

OKI comprises 18 districts, including Air Sugihan, Cengal, Jejawi, Kayu Agung (City), Lempuing, Lempuing Jaya, Mesuji, Mesuji Raya, Mesuji Makmur, Pampangan, Pangkalan Lapan, Pedamaran, Pedamaran Timur, Sirah Pulau Padang, Sungai Menang, Tanjung Lubuk, Teluk Gelam, and Tulung Selapan. The climate of OKI Regency is classified as wet tropical with annual rainfall > 2,500 mm year⁻¹ and average rainy days of 116 year⁻¹. Furthermore, the dry season predominantly occurs between May and October, but the rainy period appears from November–April. The OKI population is estimated at 731,721 persons or with a population density of approximately 39 persons km⁻². Major cultivated crops include coffee, sugar cane, oil palm, rubber, cocoa bean, pineapple, tea, and fish. Furthermore, burning is one of the cultures in land preparation during farming and fishing activities.

Materials The research materials incorporated daily Terra/AQUA MODIS satellite hotspot data of OKI from January 2018–December 2020 and were also downloaded from the official website of the Centre for Remote Sensing Utilization, National Institute of Aeronautics and Space/LAPAN (<http://modis-catalog.lapan.go.id>). In



Figure 1 Study area of Ogan Komerling Ilir Regency, South Sumatera Province, Indonesia.

addition, daily rainfall data for similar intervals were obtainable from the National Agency of Meteorology, Climatology and Geophysics (BMKG), as well as the general map from the Google My Maps application. Furthermore, the JASP 0.14.1.0 software was employed for statistical analyses involving descriptive, t-test, and correlation test of hotspot and rainfall data.

Daily hotspot and rainfall data of OKI between January 2018 and December 2020 were downloaded from LAPAN and BMKG, respectively. The classification of the hotspot data was based on monthly distribution per year and on district distribution to define the fire season and the most prone district. Meanwhile, the daily rainfall distributions were used to determine the mean rainfall value for the three-year duration and to evaluate the rainfall-hotspot correlation. Finally, hotspot data for the period under review were analyzed, using the Google My Maps application to generate a spatial hotspot distribution during and before the COVID-19 pandemic (Figure 2).

Data analyses The hotspot differences before COVID-19 (2018 and 2019) and the actual arrival (2020) were analyzed. Paired t-test was employed to successively compare the hotspots between 2018 and 2019, 2019 and 2020, as well as 2018 and 2020. Similar assessment and comparison were also applied to evaluate the rainfall. In addition, the correlation test was deployed to determine the relationship between both variables and to observe the extent of rainfall influence on the fire occurrence. The correlation equation is as shown in Equation [1].

$$r_{xy} = \frac{n\sum x_i y_i - (\sum x_i)(\sum y_i)}{\sqrt{\{n\sum x_i^2 - (\sum x_i)^2\}\{n\sum y_i^2 - (\sum y_i)^2\}}} \quad [1]$$

note: r_{xy} = correlation coefficient of linearity, x = rainfall data variable, y = hotspot data variable, and n = number of data. The r_{xy} correlation coefficient value ranges from -1 to 1. A positive symbol indicates a proportional relationship, while a negative one shows an inverse proportionality.

Results and Discussion

Hotspot distribution before and during the COVID-19 pandemic Fire and haze are common phenomena in OKI, with great influence on human health, economic activities, and the surrounding environment. The impacts are not only limited to the natives but further extend to neighboring provinces and countries. This study analyzed the hotspot data as forest and land fire indicators before (2018, 2019) and during COVID-19 (2020). Hotspot distribution varies in the three periods under review, where the maximum and minimum occurrences were recorded in 2019 (13,871) and 2020 (435), respectively. Descriptive analyses showed that the daily hotspots in 2018, 2019, and 2020 ranged from 0–77, 1–1,398, and 1–26, correspondingly, as represented in Table 1.

The temporal distribution shows that hotspots were discovered between January and December, although the fire season started in January and peaked in October 2018. However, the incidence commenced slightly late in 2019, as a large hotspot number was reported in July, with the maximum point attained in October. Based on Figure 3, no hotspot was recorded in June 2019. The extreme dry season in 2019 possibly triggered the large fire incidence as indicated by extensive hotspots, corresponding to El Nino southern oscillation and Indian Ocean dipole (Hendrawan et al., 2019; Novitasari et al., 2019). Similar conditions have been

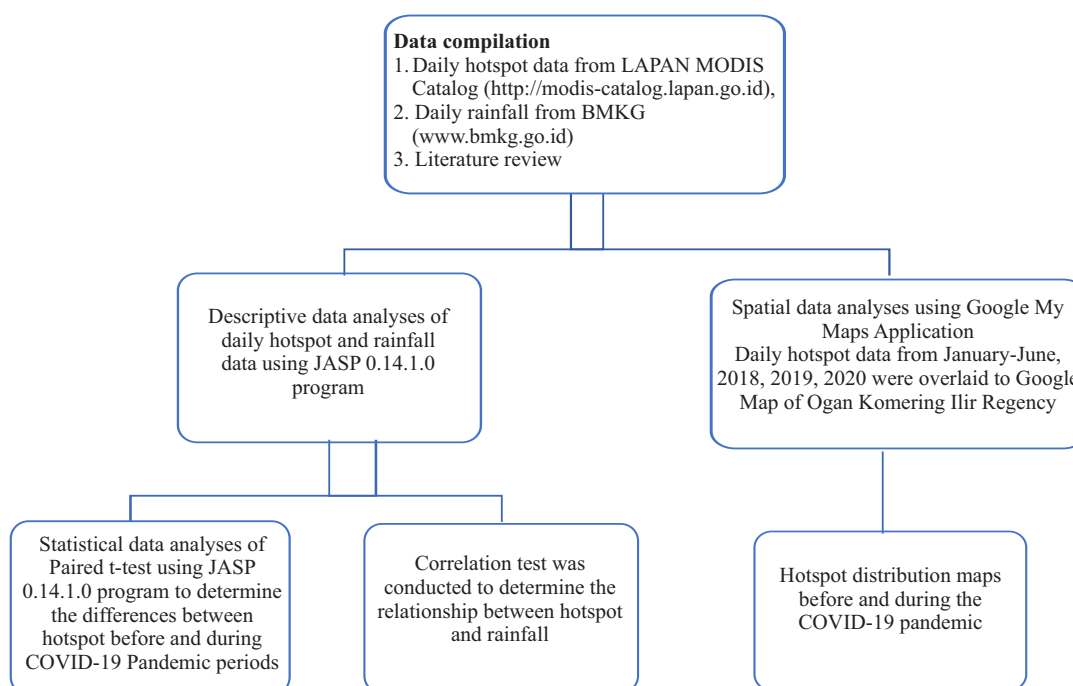


Figure 2 Flow chart of study implementation.

Table 1 Descriptive statistics of daily hotspot distribution at OKI Regency in 2018, 2019, and 2020

	Hotspot 2018	Hotspot 2019	Hotspot 2020
Valid	183	166	130
Missing	913	930	966
Mean	5.213	83.560	3.346
Std. deviation	8.613	183.317	3.617
Minimum	0.0	1.0	1.0
Maximum	77.0	1398.0	26.0

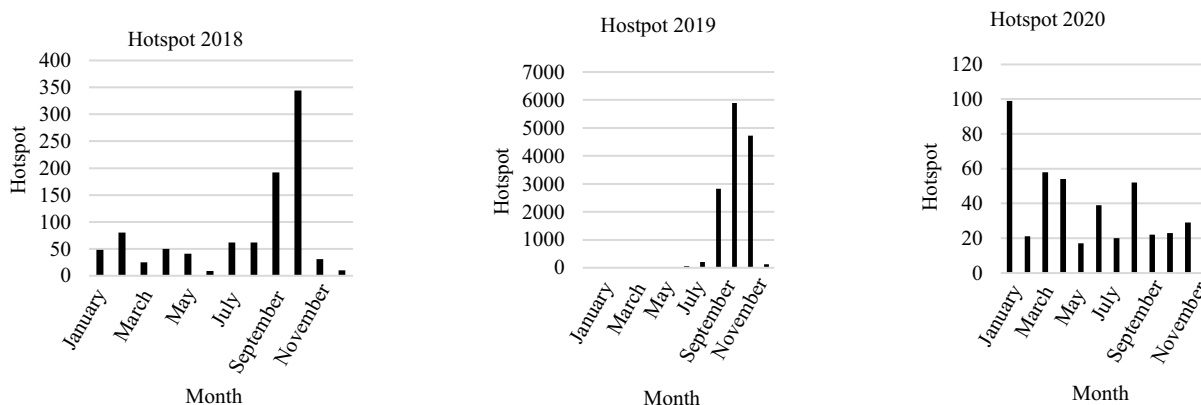


Figure 3 Monthly hotspot distribution at OKI Regency from 2018–2020.

experienced in 1998 and 2015 (Asteriniah & Sutina, 2017). During the introduction of social restrictions in 2020, the hotspot numbers, however, decreased by 54.4 and 96.9% of the overall estimations in 2018 and 2019. Figure 4 shows a minimal temporal distribution for 2020, compared to 2018 and 2019. Furthermore, it was assumed that the pandemic instigated a significant constraint in human activities, resulting to a decline in burning. This observation is in line with the statistical data analysis (Table 2), where a substantial variation was visible between 2018 and 2020 as well as 2019 and 2020 hotspots, indicating significant decreasing figures for both compared intervals. Apparently, burning activities were known to reduce drastically, mainly as a result of the social restrictions.

The Indonesian government issued a regulation on large-scale social restriction (PSBB) to curb the COVID-19 spread. This health quarantine measure is defined as a “restriction on certain human activities in selected locations possibly infected or contaminated by the virus” (UU 6/2018, Chapter 1 number 11). Also, PSBB is among the health isolation categories, apart from houses, hospitals, and quarantine areas (UU 6/2018, Chapter 49 verse (1)). The objective of the initiative is primarily to prevent the broader spread of public health emergency disease (KKM) among population in a certain area (UU 6/2018, Chapter 59 verse (2)). In addition, the restricted activities including no activities for schools and workplaces, religious programs, and/or public facilities (PP 21/2020, Chapter 4 verse (1)). Furthermore, PSBB was implemented by provincial and regency/city governments, after the approval of the Ministry of Health through the Ministerial decrees (PP 21/2020, Chapter 2 verse (1) and UU 6/2018, Chapter 49 verse (3)).

Figure 5 represents the spatial hotspot distribution in OKI from 2018–2020. The results of data mapping were achieved using Google My Map. Consequently, clusters were formed covering the research location as a strong indicator for the actual fire (Kirana et al., 2015; Wijaya et al., 2016; Syaufina & Sitanggang, 2018).

Figure 6 shows the fire distribution virtually in every district of OKI before and during the COVID-19 crisis. The highest average hotspot (52) occurred in Tulung Selapan, between January and June, followed by Pedamaran (38), Pampangan (38) and Cengal (37). This outcome observed Tulung Selapan as the most fire-prone district in OKI (Meiriza et al., 2017; Nainggolan et al., 2020). However, Air Sugihan, Cengal, Bayung Lencir were on a medium scale, while the rest were reportedly low (Meiriza et al., 2017). Furthermore, fire-related human activities appear as the primary cause of forest and land fires in this region, including the 'sonor' techniques employed in rice planting, where the seeds are spread on the burned area (Zulfah et al., 2020), and land preparation for plantation and fishing (Nurhayati et al., 2020).

Moreover, the pandemic era showed additional numbers of hotspot distribution districts (10) compared to the other two periods of 2018 (7) and 2019 (3). Also, it appears that burning activities are currently implemented during COVID-19. However, the OKI had attempted to adapt zero burnings by responding with the prohibition on the use of fire in land clearing, despite it has not been fully implemented (Waluyo et al., 2020).

The effects of rainfall on hotspot In tropical regions, rainfall forms an important climatic consideration that

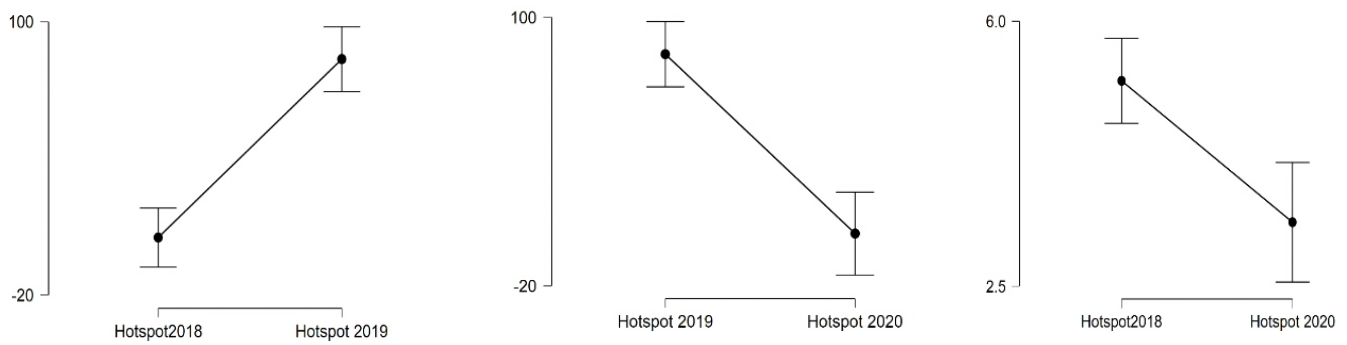


Figure 4 Hotspot distribution comparison at Ogan Komering Ilir District before (2018–2019 and during the COVID-19 pandemic (2020).

Table 2 Paired sample test of hotspot in 2018, 2019, and 2020

Measure 1	Measure 2	t	df	p-value
Hotspot 2018	Hotspot 2019	-4.262	94	< 0.001
Hotspot 2019	Hotspot 2020	3.273	63	0.002
Hotspot 2018	Hotspot 2020	2.224	69	0.029

Note: Student's t-test.

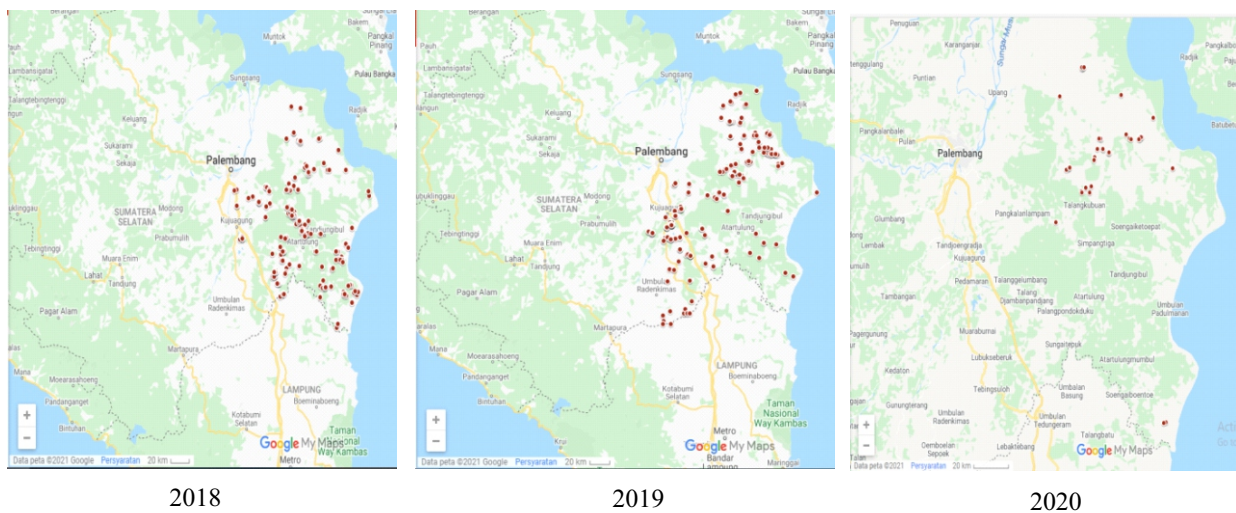


Figure 5 Spatial hotspot distribution at Ogan Komering Ilir Regency from 2018–2020.

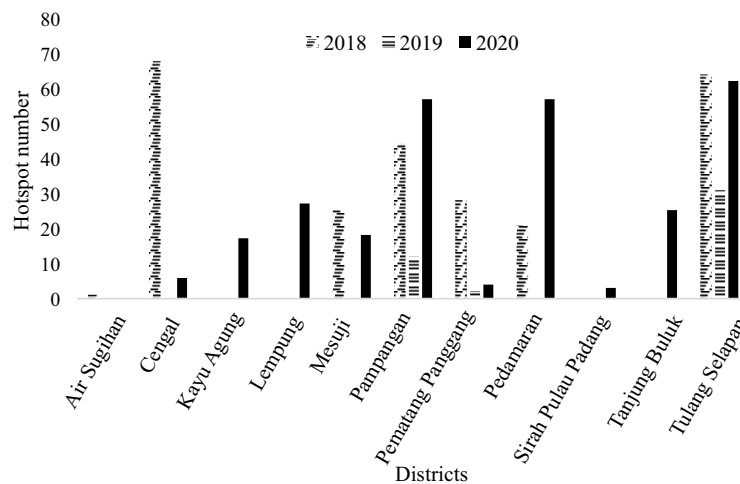


Figure 6 Hotspot distribution at Districts of Ogan Komering Ilir Regency from January–June 2018, 2019, 2020.

mostly influences forest and land fires and also determines the quantity and quality of forest fuels (Syaufina, 2008). Indonesia shows a wide rainfall variation, based on regions and the individual effect on the fire seasons. This weather element has the potential to serve as a parameter for fire risk prediction in a particular duration. In peatlands, the seasons show a huge impact on fuel moisture, bulk density, and water level (Syaufina et al., 2004).

There were fluctuations in rainfall distributions from 2018–2020. Table 3 shows the statistical analyses of the minimum mean daily occurrence in 2019 (7.1 mm) and the maximum in 2020 (10.6 mm). In addition, significant variations were reported from 2018–2019 and 2019–2020, with no change between 2018 and 2020 (Table 4). Furthermore, the OKI Regency in A region, obtained a single peak in both the rainy and dry seasons (July–August) (Aldrian & Susanto, 2003).

The dry season is determined by minimal monthly rainfall that influences fire occurrences and hotspot number as potential indicators (Syaufina, 2008). The hotspot number is increased as the location becomes drier. Also, the rainfall significantly decreased from 2018 to 2019, where El Nino occurred in the region. Therefore, forest and land fires were known to peak in 2019 (Figure 3), particularly from September, October to November, known as the common periods for most hotspots (Meiriza et al., 2017). One of the primary triggers of forest and land fires is lesser rainfall and extreme anomalies (Nurdiati et al., 2019), with the potentials to decrease peat moisture content and increase fire risk

(Armanto et al., 2018). This was followed by a significant rainfall improvement from 2019–2020, as illustrated in Figure 7.

Figure 8 represents the statistical analysis of a very inefficient correlation ($r = 0.103$) between daily hotspots and rainfall. This means that natural factors demonstrated a more vulnerable influence on burning activities, compared to human elements. A negative sign depicts the tendency for the hotspot to decline due to the increasing rainfall. Similar studies also reported a weak correlation ($r = 0.307$) between the hotspots and rainfalls in Sumatera and Kalimantan (Prayoga & Yananto, 2017). This outcome indicates that both variables do not linearly correlate as rainfall is not a single influencing factor of forest and land fires (Prasasti et al., 2012; Sitanggang et al., 2018). The alteration of climate parameters shows the slight climate change present in southern Sumatera, particularly in OKI, but without any effect on land and forest fire occurrences (Afifuddin et al., 2019). Furthermore, the fire incidence in the study region is strongly influenced primarily by human activities. Economic variables, including the proportion of plantation landholdings and the reported use of fires for agricultural land clearing, were important in comprehending the fire count at the regency level, although rainfall, slope, and population density, have been the most important considerations in predicting fires (Sze et al., 2019). Therefore, it appears OKI forest and land fires are sufficiently influenced by human activities, compared to the natural factors, including the rainfalls.

Table 3 Descriptive statistics of daily rainfall distribution at OKI Regency in 2018, 2019, and 2020

	Rainfall 2018	Rainfall 2019	Rainfall 2020
Valid	262	286	242
Missing	834	810	854
Mean	9.664	7.103	10.561
Std. deviation	16.553	14.364	16.859
Minimum	0.0	0.0	0.0
Maximum	97.0	80.5	90.6

Table 4 Paired sample test of rainfall in 2018, 2019, and 2020

Measure 1	Measure 2	t	df	p-value
Rainfall 2018	Rainfall 2019	1.420	204	0.157
Rainfall 2019	Rainfall 2020	-2.321	193	0.021
Rainfall 2018	Rainfall 2020	-0.478	171	0.633

Note: Student's t-test.

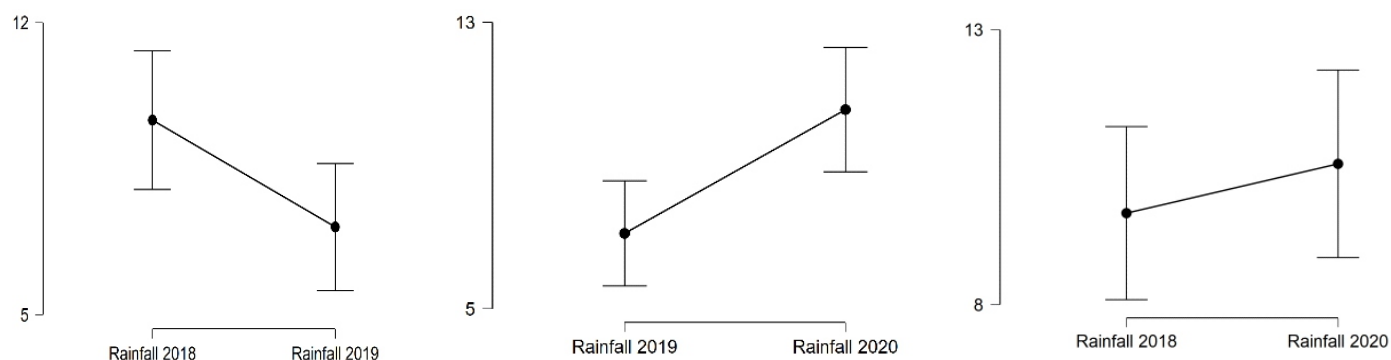


Figure 7 Comparison of mean daily rainfall at Ogan Komering Ilir Regency from 2018–2020.

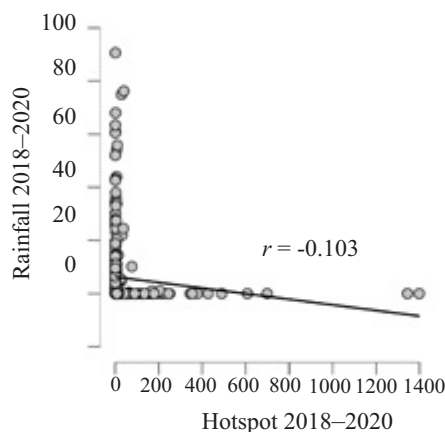


Figure 8 Correlation between hotspot and rainfall from 2018–2020 in OKI Regency.

Conclusion

This study indicates a significant decrease in the number of hotspots between before and during the COVID-19 period at the study site. We believe that the large-scale social restrictions have suppressed land burning activities. This study also verified that daily rainfall has a weak correlation with the number of hotspots. Therefore, this study shows that rainfall contributes weakly to the OKI Regency's land fires compared to human factors.

Recommendation

Field observation and hotspot ground checking are necessary for strengthening the research results.

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