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Distribution of Land Surface Temperature In Yogyakarta Urban Area

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ABSTRACT

Yogyakarta urban area has grown throughout the year as the course of migration and its attraction to tourists and students resulting in the high demand of living space which leads to the increment of the built-up area such as hotels and other supporting-tourism-activity accommodation, so-called urban sprawl. The increase of paved-surface causes the increase of land surface temperature (LST), which may impact to micro-climate in the urban area with adverse consequences, for instance, erratic rainfall and rainstorm in the urban area. Consequently, it triggers new future problems. Therefore, it is necessary to study LST to perceive the state of temperature change in the study area as a preliminary assessment in studying urban climate.

This paper attempts to present the distribution of Land Surface Temperature (LST) in Yogyakarta urban area, extracted from remotely-sensed Landsat 8 image acquired from a two-year image. Before the extraction, several variables are incorporated such as Normalized Difference Vegetation Index (NDVI) to calculate emissivity, as well as atmospheric correction parameter transmissivity, upwelling and downwelling radiance. The reflectance values are also corrected to obtain NDVI. Land surface temperature extracted according to the procedures: conversion of the digital number of Landsat image to radiance, correction of radiance value, conversion of the corrected radiance value to brightness temperature, then brightness temperature to land surface temperature. The extracted temperature map then presented into 5°C interval.

Consecutively, the two-year of temperature maps are then analyzed to obtain the difference of its spatial distribution. The expected result is the expanding high-temperature distribution in the urban area. The result shows there is an increase in the average land surface temperature by 1,5°C from two different images, 2014 and 2018. The majority value of temperature is between 30 – 40°C, dominated with the built-up area. Two image shows that the respective area spread from 54% to 70%.

Keywords: Diurnal, LST, Landsat, Urban

1. INTRODUCTION

The urban area tends to expand in parts of the world as the increment of population migration to an urban area. The expansion of urban area tends to alter the urban landscape that the proportion of built-up area dominated the space as the demand for economic activities and space to live. Yogyakarta has attracted tourists and students recently, which contributes to the more built-up area to accommodates the activities, for instance, student housing, restaurants and hotels. As a consequence, Yogyakarta city expanded as the growth of the urban area as the implication of these demands.

The study of urban growth of Yogyakarta city had been studied by Giyarsih (2010). The growth of the urban area of Yogyakarta through a long process other than short one, the spatial distribution of the transformation influenced by street connections and growth areas. Moreover, the growth of Yogyakarta in terms of the expansion of urban area according to the finding of Divigalpitiya, P & Handayani, K. N (2015) shows that Yogyakarta city experienced outward expansion commenced with small urban patches then merged to form larger urban patches. The visible expansion mainly through the period from 2002 to 2013.

As the tendency of urban growth follows the increase of paved-surface/ built-up area, it affects the state of urban climate. Urban microclimate associated with effects of urban landscape design had studied in Yang, Lin, & Li (2018). One of the findings is the shade trees combined with grass can reduce air temperature by 0.75oC. Therefore, the monitor of temperature in the urban area becomes vital as the temperature is one of the elements of climate. The change of temperature

has indirect impacts on urban climate, for instance, erratic rainfall and rainstorm, and other hydro-meteorological hazards. Furthermore, the importance of temperature monitoring has been increasing as the concern in global warming.

The use of Remote sensing technique become possible to extract phenomena without direct contact with objects (Lillesand & Kieffer, 2002). Remote sensing imagery has been widely used as input to monitor the state of temperature. Remote sensing has the advantage of its efficacy and efficiency due to the less fieldwork and affordability comparing to field measurement. This image records the earth's surface every 16 days, making it possible to use it for monitoring an area regularly. Each pixel of Landsat 8 multispectral image represents 30 m x 30 m, with 12 bits of radiometric resolution. Landsat 8 consists of two instruments, namely Operational Land Imager (OLI) and Thermal Infrared Sensor (TIRS) which provides a wide range of spatial scope and decades of temporal data assuring to give beneficial inputs for temperature monitoring.

This paper attempts to demonstrate a sequence to extract land surface temperature (LST) information from Landsat 8 thermal band from two different years, on 2014 and 2018, in Yogyakarta urban area. This research emphasizes on the methodology which to provide a step-by-step guide for future monitoring of temperature involving longer time series data.

2. STUDY AREA

Special Region of Yogyakarta is located in the southern part of Java Island and has a direct border to the Hindia Ocean (Figures 1). The capital city of Yogyakarta has 32,1 km² of area and located in the centre of the province. It is famous for its culture and education and tourist destination. Although it is known as the smallest province in Java island, Yogyakarta is one of the most livable cities in Indonesia (IAP, 2014). Statistic Agency of Yogyakarta reported there had been growing in population. The average population density is 13,007 inhabitants/km². The urban area of Yogyakarta has been growing and encompasses 26 districts in total. The average rainfall in Yogyakarta is 2.012 mm/years. The average air humidity is quite high, ranges from 81-90%, where the lowest occurs in August and September, which is 81%. The average temperatures around 27°C. The dry season begins in April until October.

Figure 1. Map of Yogyakarta's Urban Area

3. DATA AND METHODOLOGY

Landsat 8 Thermal Infrared band (TIR) was adopted to extract land surface temperature (LST) information in Yogyakarta urban area. Landsat 8 TIR images were acquired from two different dates in the dry season, in August 2014 and May 2018. Due to the LST extraction process, atmospheric correction parameters obtained from the website (<https://atmcorr.gsfc.nasa.gov/>) such as transmissivity, upwelling, and downwelling radiance (Figure 3). Normalized Difference Vegetation Index (NDVI) value also extracted to determine emissivity value before LST extraction. Atmospheric correction was also applied to obtain reflectance value to extract NDVI. The required parameters to extract temperature were obtained from image metadata.

Figure 3. Atmospheric Correction Parameter for (a) 2014 and (b) 2018 Landsat 8 Images

Generally, land surface temperature extraction from Landsat 8 TIR band requires several following procedures:

1. Conversion pixel value/ digital number (DN) to radiance (Equation 1), correction of radiance value (atmospheric correction) using Equation 4
2. Conversion of the atmospheric-corrected radiance value to brightness temperature using Equation 2
3. Conversion of brightness temperature to land surface temperature using Equation 3

Reflectance value of NDVI was corrected using Equation 6, 7, and 8.

Emissivity was calculated using Equation 5

The equations below were applied to extract land surface temperature from Landsat 8 TIR image.

Digital Number (DN) to Spectral Radiance:

$$L_{\lambda} = M_L Q_{cal} + A_L \quad (1)$$

L_{λ} = Top of Atmosphere spectral radiance (Watts/(m² *srad* μ m))
 M_L = Multiplicative scale*
 A_L = Additive value*
 Q_{cal} = Digital Number (DN)
 *from metadata

Spectral radiance to brightness temperature was calculated using equation mentioned in McCarville, et. Al. (2011):

$$T_B = \frac{K_2}{\ln\left(\frac{K_1}{L_{\lambda}} + 1\right)} \quad (2)$$

T_B = Brightness temperature (K)
 L_{λ} = Spectral radiance (Watts/(m²*srad* μ m))
 K_1, K_2 = Constant value (from metadata)

Brightness temperature to land surface temperature was calculated as follows (Weng et al., 2003):

$$T_s = \frac{T_B}{1 + \left[\frac{\lambda T_B}{\sigma} \times (\ln \varepsilon)\right]} \quad (3)$$

T_s = land surface temperature (K)
 T_B = brightness temperature (K)
 λ = wavelength (11, 5 μ m)
 σ = hc/ σ (=1.438 x 10⁻² mK)
 h = Planck's Constant (6.26x10⁻³⁴ J sec)
 c = light's speed (2.998 x 10⁸ m.sec⁻¹)
 σ = Stefan-Boltzman Constant (1.38 x 10⁻²³ JK⁻¹)
 ε = emissivity

Atmospheric correction by applying equation mentioned in Coll et al. (2010):

$$L_{\lambda}'' = \left[\frac{L_{\lambda} - L_{\downarrow}}{\varepsilon \tau} \right] - \left[\left(\frac{1 - \varepsilon}{\varepsilon} \right) L_{\downarrow} \right] \quad (4)$$

L_{λ}'' = corrected L_{λ} (Watts/(m² *srad *μm))
 ε = emissivity
 τ = transmissivity
 $L_{\uparrow}/L_{\downarrow}$ = upwelling/ downwelling radiance (Watts/(m² *srad *μm))

Emissivity was calculated using the following equation (Urbanski, 2014):

$$\varepsilon = 1.0094 + [0.047 \times \ln(\text{NDVI})] \quad (5)$$

Reflectance value was calculated as follows (Congedo, 2016):

$$\text{ESUN} = (\pi * d^2) * [\text{Rad max/Ref max}] \quad (6)$$

$$L_p = (M_L * \text{DN}_{\min}) + A_L - 0.01 * \text{ESUN} * \cos \theta_s / (\pi * d^2) \quad (7)$$

$$\rho = \pi * d^2 [L_{\lambda} - L_p] / \text{ESUN} * \cos \theta_s \quad (8)$$

ρ = Reflectance
 ESUN = Solar exo-atmospheric irradiance
 L_p = Path radiance
 $\text{DN}_{\min}$ = Minimum Digital Number
 d = Earth-Sun Distance
 θ_s = Solar zenith angle
 Rad max/Ref max = radiance and reflectance maximum (from metadata)

The LST image is visualized in 5°C interval value. Two different images then analyzed in order to present how the difference of temperature between two image.

4. RESULT AND DISCUSSION

The Geometric correction was not applied to the images because it was conducted systematically from the data provider. However, the two-image acquisition was processed through the whole steps of atmospheric correction before temperature extraction to reduce atmospheric disturbance and improve temperature extraction value in terms of its accuracy.

Normalized Difference Vegetation Index (NDVI) calculated for obtaining emissivity value which used as the input for determining Land Surface Temperature (LST). The NDVI images of 2014 and 2018 classified into the same range. According to the range, it is estimated that the built-up area made up about 49% and 50% of the total urban area, respectively, according to those two images. It infers that the built-up area relatively remains unchanged. In other words, urban expansion does not occur. The NDVI results are shown in Figure 3 (a) and (b). According to those images, low NDVI values (approximate to and below 0 (zero) which indicates likely as paved-area) concentrated in the city centre. It is apparent that low NDVI values (in red-orange) depicted along the network line (asphalt road) in both images.

In addition to NDVI, the average emissivity value decreased, from 0.93 in 2014 to 0.88 in 2018. This result implies that more pixels in 2018 that reflecting the incoming energy compared one in 2014 (Figure 3 (c) and (d)). The emissivity of two images also shown in the histogram in Figure 6 (a) 2014 (b) 2018. According to the histogram's shape. The latter shows shaper bell on the left compared to 2014 image, in order word, flatter and gradually increase steadily compared to one in 2018. This pattern indicates the emissivity value distributes likely to the lower one. In this paper, emissivity value

is not further discussed as emissivity value depends on the function of wavelength together with temperature. Therefore, emissivity alone is complex to obtain. Therefore, further research is needed, for instance, using an instrument to estimates emissivity value and different time of image acquisition. Emissivity value can be more sensitive. The statistics of emissivity value, as shown in Table 1, with maximum and minimum in 2014 and 2018 value are 0.98 & 0.65, and 0.97 & 0.37, respectively.

Figure 4. NDVI Value at (a) 2014 and (b) 2018 also Emissivity at (c) 2014 and (d) 2018

The extracted temperature is assumed as “kinetic temperature” since it involved emissivity. The result of temperature extraction shown in Figure 4 (a) and (b). The LST was classified into 5°C interval. According to the map, the high LST distributes along the main road along with the concentration of paved-surface. The total study area is 19,662 Ha. Figure (5) shows the statistics of temperature distribution for the two-analyzed images in the classified interval. In 2014, the area with LST around 20 – 30°C was under 1,000 Ha. The area with 30 – 35 °C is around 5,000 Ha. About 7,000 Ha is covered with the surface at temperature ranges from 35 to 40 °C, which makes the most significant proportion of the total area of

Yogyakarta Urban Area. The area with 40 - 45 °C and 45 – 50 °C interval occupied around less than 6,000 Ha and 2,000 Ha, consecutively.

According to the 2018 image, the area with 25 to 30 °C occupies a small area of Yogyakarta Urban Area. The four highest areas occupy the urban area at range 35 - 40 °C, 40 – 45 °C, 30 - 35 °C, and 45 – 50 °C, respectively; being 35 – 40 °C dominates the surface, reach more than 7,000 Ha. Both images of 2014 and 2018 show that the highest proportion of LST in the study area is around 35 - 40 °C. Such high-temperature area, more than 50 °C, some reach 95 – 100 °C. The probable cause of such high temperature might find from asphalt road or the industrial chimneys. Both images also show a similar pattern. It illustrates that the centre of the city has the highest average temperature. The further from the city centre, the higher the temperature. According to the graph, it simply shows that the distribution of pixels with lower temperature tends to shift into higher temperature, comparing 2014 and 2018 image. The histograms, as shown in Figure 7 (a) 2014 (b) 2018, concede the shifted values aforementioned.

Table 2 represents the statistical value of temperature extraction of both image 2014 and 2018. The average temperature in 2014 was 38°C, it rose to 39,5°C in 2018. The minimum and maximum value rose by approximately 4 °C and 18 °C, respectively; while the standard deviation is not considerably changed.

Figure 5 Land Surface Temperature Map in (a) 2014 and (b) 2018

Figure 6. Graph of Changes in Temperature Class Area

Figure 7 Histogram of Emissivity (a) 2014 (b) 2018

Table 1. Statistics of Emissivity in 2014 and 2018

YEAR	MIN	MAX	MEAN	STDEV
2014	0.65	0.98	0.93	0.01
2018	0.37	0.97	0.88	0.05

(a)

(b)

Figure 8 Histogram of temperature value (a) 2014 (b) 2018

Table 2. Statistics of Land Surface Temperature in 2014 and 2018

Years	Min	Max	Mean	StDev
2014	24.75	81.1	38.04	4.62
2018	28.83	98.72	39,55	4.68

5. CONCLUSION

This result shows that the urban area in terms of the built-up area relatively remains unchanged. The average emissivity of both two images of 2014 and 2018 decreases which infers more proportion of reflected energy by the surface in the study area. The temperature tends to increase according to the average land surface temperature (LST).

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