



## 1 Conference Proceedings Paper

# Urban Heat Island Analysis Using the Landsat 8 Satellite Data: A Case Study in Skopje, Macedonia

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11 Abstract: An urban heat island (UHI) is an urban area that is significantly warmer than its 12 surrounding rural areas due to antropogenic activities. The urban area of the city of Skopje has been 13 rising rapidly in the past decade. In this study, the effect of UHI is analyzed using Landsat 8 data in 14 the summer period of 2013 – 2017 as a case study in Skopje, Macedonia. An algorithm was applied 15 to retrieve the land surface temperature (LST) distribution from the Landsat 8 data. In addition, the 16 correlation between land surface temperature and the normalized difference vegetation index 17 (NDVI) and the normalized difference build-up index (NDBI) were analyzed to explore the impacts 18 of the green areas and the build-up land on the urban heat island. The results indicate that the effect 19 of urban heat island in Skopje is located in many sub-urban areas. The negative correlation between 20 LST and NDVI indicates that the green area can weaken the effect on urban heat island, while the 21 positive correlation between LST and NDBI means that the built-up land can strengthen the effect 22 of urban heat island in the study area.

- 23 Keywords: Remote Sensing; Urban Heat Island; Land Surface Temperature; Skopje.
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### 25 1. Introduction

26 Due to the global mean surface temperature has raised since the 20<sup>th</sup> century, global climate 27 change has acquired more attention. Urbanization is one of the most significant contributors to global 28 warming as more than 50% of the human population lives in cities. The growth of population and 29 urbanization has caused the phenomenon of Urban Heat Islands (UHI) which can be described as the 30 alteration of temperatures in urban areas compared to their rural surroundings [1]. In order to 31 observe UHI and their behavior, UHI studies are generally conducted in two ways: through the air 32 temperature measuring, or through measuring the surface temperature [2]. With the ability of remote 33 sensing sensors to measure the surface temperature, remote sensing has been successfully used in 34 UHI studies all over the world. Different instruments have been used in UHI studies as The Moderate 35 Resolution Imaging Spectroradiometer – MODIS [1, 3, 4], Landsat TM, ETM+ and OLI/TIRS [5, 6], 36 ASTER [7], as well as their combined use. As MODIS is a low-resolution sensor, it has been used for 37 big areas, while Landsat and ASTER with their medium resolution are more suitable for observing 38 single cities or smaller areas. Research on LST and UHI showed that surface temperature response is 39 a function of different land cover [8] which prompted research on the relationship between LST and 40 land cover, and especially vegetation abundance [5, 9-11].

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41 Using remote sensing techniques, various vegetation indices can be obtained and used in the

42 assessment of vegetation cover. The Normalized Difference Vegetation Index (NDVI) has been

- 43 widely used for vegetation extraction. Higher NDVI values indicate higher vegetation area in a pixel.
- 44 For extraction of urban areas, Normalized Difference Built-up Index has been widely used. Building
- 45 a connection between the land cover and LST can be valuable for urban climate studies.
- 46 In this study, the effects of UHIs, as well as the relationship between LST, NDVI and NDBI, has 47 been studied for the city of Skopje. For this purpose, two Landsat 8 satellite images from 2013 and
- 48 2017 summer period have been downloaded from the USGS webpage.

## 49 2. Study Area and Methods

50 Skopje is the capital of the Republic of Macedonia located in the center of the Balkan Peninsula. 51 Republic of Macedonia is a landlocked country with a population of approximately 2 million 52 citizents. The population of Skopje is estimated to be more than 700.000 [12]. In the last few decades, 53 the urban area of Skopje has been significantly expanding. Skopje has been built in the Skopje Valley 54 along the riverbed of Vardar. Skopje covers an area of 1818 km<sup>2</sup>, 23 km in longitude and 9 km in 55 latitude and is situated on a height of 245 meters above the sea level. Skopje has an average 56 temperature of 13.5 °C and its average rainfalls per year are 940 mm [13].

57 In this study, Landsat 8 satellite data have been used in order to investigate the UHI in 35 58 municipalities in the city of Skopje (Figure 1) and its behavior in the period of 2013 and 2017. For this 59 purpose, two satellite images (Path 185 and Row 31) from July 2013 and July 2017 have been 60 downloaded from the USGS webpage. Also, air temperature data from the meteorological station

61 have been obtained for the same day and time with the overpass of the satellite over the city of Skopje

62 (Table 1).

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Figure 1. Study Area; The city of Skopje.

Table 1. Air temperature data.

Date	Time	Air Temperature	Humidity
01 July 2013	10:20	22°C	35%
12 July 2017	10:20	30°C	43%

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70 For extracting the UHIs, LST tool developed using ERDAS IMAGINE Model Maker was used

71 [14]. The tool is used for retrieving LST of a given LANDSAT 8 image with the input of the fourth

- 72 (red wavelength/micrometres, 0.64–0.67), fifth (near infrared (NIR) wavelength/micrometres, 085–
- 73 0.88), and tenth (thermal infrared sensor (TIRS) wavelength/micrometres, 10.60–11.19) bands. A
- flowchart of the used algorithm is shown in Figure 2.



75 76

Figure 2. Flowchart of the Land Surface Temperature algorithm.

77 The UHI has been extracted using the following method [15]:  
78 
$$UHI = \mu + \frac{\sigma}{2}$$
 (1)

where μ is the mean LST value of the study area, and  $\sigma$  is the standard deviation of the LST. NDVI and NDBI were calculated in order to determinate the correlation between the land cover and the LST results. NDVI is calculated using the 5<sup>th</sup> and 4<sup>th</sup> or Near Infra Red and Red Landsat 8 bands (Eq. 2).

83 
$$NDVI = \frac{NIR - Red}{NIR + Red}$$
(2)

The NDBI index [6] is calculated using the 6<sup>th</sup> and 5<sup>th</sup> or the Middle Infra Red and Near Infra-Red
 Landsat 8 bands respectively.

(3)

86 
$$\mathbf{NDBI} = \frac{\mathbf{MIR} - \mathbf{NIR}}{\mathbf{MIR} + \mathbf{NIR}}$$

#### 87 3. Results

A correlation coefficient between LST and NDVI, and LST and NDBI has been calculated in
 order to explore the impacts of the green land and the build-up land on the UHI. The results from the
 performed statistical analysis are shown in Table 2.

92	Year	LST-NDVI	LST-NDBI
	2013	-0.63	0.67
93	2017	-0.59	0.64

94 The results from the land cover analysis showed that the vegetation area has slightly decreased 95 from 2013 to 2017 and expansion of some urban areas has been noticed. The surface temperature The 2nd International Electronic Conference on Remote Sensing (ECRS 2018), 22 March–5 April 2018; Sciforum Electronic Conference Series, Vol. 2, 2018

- 96 analysis showed that the minimum and maximum temperature in the study area were 15 °C and 37
- 97 °C for 2013, and 24 °C and 49 °C for 2017, which matches the air temperature differences from the
- 98 meteorological data. The results from the LST, NDVI and NDVI analyses are shown in Figure 3 and
- 99 Figure 4.



## 100 101 102

Figure 3. Results for 01 July 2013.



103 104 105

Figure 4. Results for 12 July 2017.

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106 The comparison between the UHI from 2013 and 2017 indicates an increase of the UHI area for

107 approximately 4 km<sup>2</sup> (Table 3).

108	Table	<b>Table 3.</b> 2013 – 2017 UHI comparison.		
109	Year	Study Area (km²)	UHI area (km²)	
110	2013	130	51	
110	2017	130	55	

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100

#### 112 4. Discussion

113 In this paper, the relationship between land cover and UHI of the city of Skopje has been studied. 114 Correlation of LST with both NDVI and NDBI indices were calculated and the results showed that 115 the effect of UHI in Skopje is located in many sub-urban areas. The negative correlation between LST 116 and NDVI indicates that the green areas can weaken the effect on urban heat island, while the positive 117 correlation between LST and NDBI means that the built-up land can strengthen the effect of urban 118 heat island in the study area.

#### 119 5. Conclusions

120 Global climate change is expected to raise the occurrence of urban heat island effects. Air 121 temperatures in cities increase disproportionately to urban areas and concluded locally acute adverse 122 human health, economic, social and environmental impacts. Land use/cover has relationship with 123 LST that can help in land use planning contributed to global climate change and changed UHI 124 intensity mainly through the process of the study area. By the latent heat flux from the surface to 125 atmosphere via evapotranspiration, the amount of green areas determines LST. Lower LSTs usually 126 are found in areas with high NDVI. This negative correlation between NDVI and LST and positive 127 correlation between NDBI and LST is valuable for urban climate studies. As an outcome of this study, 128 it can be concluded that in order to reduce UHI effects, local governments should increase the green 129 areas in the intensive populated urban areas.

130 Conflicts of Interest: "The authors declare no conflict of interest."

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