

# Dynamics of human-induced lakes and their impact on land surface temperature in Toshka Depression, Western Desert, Egypt

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## Research Article

**Keywords:** Land surface temperature, Land use/land cover (LULC), Automated Water Extraction Index (AWEI), Toshka, GIS

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**Title:** Dynamics of human-induced lakes and their impact on land surface temperature in Toshka Depression, Western Desert, Egypt

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30 **Dynamics of human-induced lakes and their impact on land surface temperature in Toshka Depression,**  
31 **Western Desert, Egypt**

32 **Abstract**

33 Land surface temperature (LST) is a significant environmental variable that is appreciably influenced by land use  
34 /land cover changes. The main goal of this research was to quantify the impacts of land use/land cover change (LULC)  
35 from the drying of Toshka Lakes on LST by remote sensing and GIS techniques. Landsat series TM and OLI satellite  
36 images were used to estimate LST from 2001 to 2019. Automated Water Extraction Index (AWEI) was applied to  
37 extract water bodies from the research area. Optimized Soil-Adjusted Vegetation Index (OSAVI) was utilized to  
38 predict the reclaimed land in the Toshka region until 2019. The results indicated a decrease in the lakes by about  
39 1517.79 km<sup>2</sup> with an average increase in LST by about 25.02°C between 2001 and 2019. It was observed that the  
40 dried areas of the lakes were converted to bare soil and are covered by salt crusts. The results indicated that the land  
41 use change was a significant driver for the increased LST. The mean annual LST increased considerably by 0.6°C/y  
42 between 2001 and 2019. A strong negative correlation between LST and Toshka Lakes area (R-square = 0.98)  
43 estimated from regression analysis implied that Toshka Lakes drying considerably affected the microclimate of the  
44 study area. Severe drought conditions, soil degradation, and many environmental issues were predicted due to the rise  
45 of LST in the research area. There is an urgent need to develop favorable strategies for sustainable environmental  
46 management in the Toshka region.

47 **Keywords:** Land surface temperature; Land use/land cover (LULC); Automated Water Extraction Index (AWEI);  
48 Toshka; GIS

49 **Introduction**

50 Land surface temperature (LST) is a significant environmental variable, and its mapping and analysis with other  
51 environmental parameters assume greater importance, especially in desert ecosystems. LST is a crucial variable in the  
52 weather pattern that controls the heat of the surface and water transfer in the interface between the atmosphere and the  
53 land (Yu and Yu, 2020). Recognizing land temperature changes over a certain period is one of the key requirements  
54 for assessing climate change (Orhan et al. 2014). There is an increasing problem that human-related changes in  
55 Earth's surface can affect climatic conditions locally or globally by altering their physical characteristics (Jingyong  
56 et al. 2005). Land use changes act as a pivotal factor in environmental resource management and conservation. Land  
57 Use/Land Cover (LULC) patterns have a crucial role in conserving the environment from exposure to direct solar  
58 radiation and increasing the surface temperature of the environment. Nevertheless, the type of LULC changes over  
59 time has caused rapid alterations in the environment and enhanced the deterioration of the environment (Balew and  
60 Korme 2020). Most land cover changes related to human activities are caused by land cover transformation, land  
61 deterioration, or land use condensation (Lambin 1997).

62 Due to the diversity in Earth's surface properties, the LST changes significantly. Therefore, continuous measurements  
63 are required to distinguish temporal changes in the Earth's surface (Li et al. 2013). The statistics from meteorological  
64 stations are separate points that barely consider the spatial alteration of temperature created by various types of land

65 cover (Zhou and Wang 2011). Remote sensing images are the only practical application for estimating LST on very  
66 large spatial and temporal scales when meteorological stations cannot provide Earth's temperature in large areas  
67 (Owen et al. 1998; Feizizadeh and Blaschke 2012; Li et al. 2013). LST can be derived from the thermal bands of  
68 remotely sensed images by many methods, depending on the utilized bands (Pu et al. 2006). Sinha et al. (2015)  
69 suggested a heat generation index using spectral data from remote sensing images to enhance the validity of LULC  
70 classification. They showed that, for regional investigation of LST, Landsat ETM+ is a preferable option. Mendelsohn  
71 et al. (2007) tested the climatic data obtained from satellites compared to ground estimates. They concluded that  
72 temperature detection from satellite images partially exceeds the capabilities of ground stations because of the  
73 capability of satellites in large-scale temperature monitoring. Hooker et al. (2018) investigated that remotely sensed  
74 temperature detection provides preferable results compared to the data acquired from meteorological stations.

75 The relationship between changes in LULC and LST dynamics has been the subject of several studies. Hereher  
76 (2017) used MODIS LST data to show changes in LST due to changes in LULC from 2003 to 2015 in the Toshka  
77 region of Egypt. Abd El-Hamid (2020) used Landsat data and GIS technique to assess the impacts of LULC on LST  
78 over the last 19 years in the Nile Delta of Egypt. Jiang and Tian (2010) applied the temperature-vegetation index  
79 (HVX) to explore the action of land changes on LST. Hathway and Sharples (2012) examined the influence of water  
80 bodies on decreasing the impact of Urban Heat Island in the UK. Ogashawara and Bastos (2012) studied the  
81 relationship between urban areas, water bodies, and LST of Brazilian urban areas using Landsat TM images. Cai et  
82 al. (2016) used Landsat TM/ ETM+ images to quantify the impacts of LULC pattern changes on LST in Fuzhou City,  
83 China. Kumar et al. (2018) used Landsat TM and Landsat TIRS/OLI images to assess the impacts of land cover  
84 changes on LST between 1990 and 2015 in Spiti Valley, India. Mustafa et al. (2019) used remote sensing indices  
85 and Landsat images to estimate the effects of land use dynamics on Beijing's microclimate from 1997 to 2017.  
86 Hussain et al. (2019) applied remote sensing and GIS tools to investigate changes in LULC patterns and their impacts  
87 on LST over 40 years in Lodhran District, Pakistan.

88 Egypt's historical susceptibility to climate change alters remarkably with the accomplishment of the Aswan High Dam  
89 (Yates and Strzepek 1998). The Aswan High Dam was built in 1964 to afford the overflow of the Nile River. In  
90 1990, the dam approached its highest storage capacity (178 m above sea level), and the Egyptian government intended  
91 to redirect water to the Toshka Valley (Sparavigna 2012). At the end of 1990, water began to flow from the Nile  
92 River through Sadat Canal by a pumping station, supplying water to the depressions and forming four large lakes. In  
93 late 1998, the first giant lake grew far east, after which other successive lakes grew west from 2000 to 2001 (El  
94 Bastawesy et al. 2007). The Egyptian government confirmed to undertake a new project recognized as the "New  
95 Valley Project." This project aimed to move the population out of the Nile Valley and Delta into the desert by  
96 reclaiming new agricultural lands and enhancing infrastructure and industrial development. In 2006, evaporation  
97 began to reduce lakes as water supply from the Nile declined due to economic obstacles and the emergence of technical  
98 and environmental problems (El-Shabrawy and Dumont 2009). By 2012, the lowest portions of the main basins  
99 were filled with about 80% less water than in 2002 (Omran and Negm 2020). Salt crusts were then formed above the  
100 evaporated portions of the lakes (Sparavigna 2011). These crusts can affect microbial activity and soil chemical and

101 physical properties, thus reducing soil productivity and plant growth due to salt toxicity and intensified soil  
102 degradation (**Amini et al. 2016**). Most of the uncertainty about the Toshka project is due to the lack of itemized  
103 planning for all parts of establishment and development and the lack of information on the environmental impacts of  
104 this project. **Loneragan and Wolf (2001)** reported that the Toshka project had a tremendous impact on the Egyptian  
105 environment. They showed that upstream salinity and pollutants resulted in the loss of extra agricultural lands and  
106 significantly affected the Nile Delta environment. **El Bastawesy et al. (2007)** predicted that the rate of water loss  
107 from the Toshka Depression was approximately 2.5 m/y, and the water stored in the lake reduced to about 12.57 billion  
108 m<sup>3</sup> in 2006. **Baradei and Al Sadeq (2019)** utilized three models to estimate the evaporation rate in Toshka. They  
109 reported that the average evaporation rate in the study area ranged from 7.92 to 8.09 mm/day. **Mostafa et al. (2019)**  
110 have predicted that by 2090, Aswan's maximum temperature will rise by about  $5.6 \pm 0.5^{\circ}\text{C}$  compared to 2006–2015.  
111 **El-Marsafawy et al. (2018)** reported that Egypt is characterized by an arid climate with an evaporation rate ranged  
112 from 1500 to 2400 mm/y. In addition, the meteorological data managed by the Aswan Dam Authority (HADA) shows  
113 that the highest evaporation rate in this region was about 7.8 mm/day in 2010 (**Hassan et al. 2018**). **Vardavas and**  
114 **Fountoulakis (1996)** developed a model to estimate the rate of evaporation from lakes. They have shown that the rate  
115 of radiation absorbed by water is responsible for the rise in water temperature during the summer. However, the  
116 absorbed energy is ready to evaporate and lose water temperature in winter. **Bayoumi and Abu-Zeid (2001)**  
117 characterized the negative influences of the Toshka Lakes. They reported that the lakes were subject to higher  
118 evaporation rates and could ultimately turn into salt crusts. Previous studies have not assessed the degree of soil  
119 degradation caused by the formation of these crusts due to evaporation.

120  
121 Surface waters play a considerable role in mitigating the impacts of global warming and supporting climate adaptation.  
122 Surface water features create urban cool islands to mitigate the effects of urban heat islands (**Gupta et al. 2018**).  
123 Water bodies are recognized as high radiation absorbers and have a favorable influence on cooling the surrounding  
124 environment (**Khan et al. 2019**). Air temperature rise is considered the main factor in changing the surface water  
125 temperature of lakes and heat flux imbalance (**Czernecki and Ptak 2018**). There is no doubt that the drying of the  
126 Toshka Lakes due to evaporation increases intense warming in the surrounding area. The limitations of previous  
127 studies related to the impacts of changes in LULC on microclimate and environment were investigated, especially in  
128 the present study area. In this research, we used remote sensing and GIS to investigate the effects of LULC in the  
129 Toshka region from the drying of lakes on LST. Furthermore, the results are seen as the beginning of a detailed study  
130 on further reduction of environmental degradation due to anthropogenic changes.

## 131 132 **Study area**

133 The Toshka Depression is located almost 250 km south of the Aswan High Dam (179 m a.s.l.), between 22° 50' 0" N  
134 to 23° 30' 0" N and 30° 10' 0" E to 31° 20' 0" E (**Fig. 1**). It consists of four sub-depressions connected by natural  
135 intrusive sheets and covers an area of about 15,000 km<sup>2</sup>. The lowest point of the depression is Wadi Toshka, which  
136 is located at 150 m a.s.l. (**Hamdan et al. 2016**). The research area is mainly enveloped by sedimentary rocks. The  
137 Toshka region is part of the western desert of Egypt, which is considered one of the driest regions in Egypt. It has

138 undergone alternative climatic conditions that have affected the characteristics of the current terrains. The area is  
 139 characterized by arid climatic conditions with limited yearly rainfalls. Annual precipitation is about 1 mm/y. The  
 140 Toshka Depression is located in a hyper-arid zone, and the evaporation rate in this region proceeds toward 5.9 mm/day  
 141 (Chipman, 2019). The average temperature fluctuates from 16°C in the winter season to 35°C in the summer season.  
 142 The depression is characterized by a flat surface on its eastern side and sand dunes and low hills on its western side.  
 143 The Toshka region is distinguished by an enormous accumulation of sand dunes and is mostly influenced by the  
 144 intense winds that control the dunes in the Sahara Desert (Elbasiouny and Elbehiry 2019).The study area is  
 145 characterized by basement rocks surmounted by sedimentary successions. The sedimentary rocks are identified by the  
 146 Paleozoic and Cenozoic rocks, while the basement rocks are covered by granites, gneiss, and granodiorite (Abdel  
 147 Moneim et al. 2014). Many geomorphologic units are displayed in this area, including alluvial plains, depressions,  
 148 sandy and erosional plains, and hilly regions (Alfaran, 2013). The development of this plain is under the control of a  
 149 progression series of faults and folds (Hamdan et al., 2016).The soil temperature regime of the Toshka region is  
 150 classified as hyperthermic, and the soil moisture regime is considered aridic (torric) (USDA 2014).

## 151 **Materials and methods**

### 152 **Satellite images**

153 As a way to predict changes in LULC and LST due to the disappearance of Toshka Lakes, 12 images of Landsat-5  
 154 TM between July 2001 and July 2009, and 6 images of Landsat-8 (OLI/TIRS) from July 2013 to July 2019 (paths  
 155 175&176/rows 44) were obtained from Earth Explorer (<https://earthexplorer.usgs.gov/>). All images consist of thermal  
 156 bands (band 6 for ETM+ with a spatial resolution of 60 m and band 10 for OLI with a spatial resolution of 100 m). To  
 157 reduce the atmospheric effects, the FLAASH Module was utilized in ENVI 5.3 (Perkins et al. 2005), and the mosaic  
 158 tool was applied in ArcMap 10.5.

### 159 **Image processing**

160 The Automated Water Extraction Index (AWEI) introduced by Feyisa et al. (2013) was applied to the mosaicked  
 161 Landsat images from 2001 to 2019 to extract water bodies from the research area. The water targets are symbolized  
 162 by positive values, and other bodies are symbolized by negative values. A threshold value was adjusted to extract  
 163 water pixels from other land cover classes (-0.28). Pixels higher than the threshold value are considered water bodies.

164 For Landsat ETM+, the proposed mathematical index is calculated according to Equation 1:

$$165 \quad AWEI = \rho_{band1} + 2.5 \times \rho_{band2} - 1.5 \times (\rho_{band4} + \rho_{band5}) - 0.25 \times \rho_{band7} \quad (1)$$

166 where  $\rho$  represents the reflectance values of the Landsat ETM+ bands

167 For Landsat OLI, the proposed mathematical index is calculated according to Equation 2:

$$168 \quad AWEI = \rho_{band2} + 2.5 \times \rho_{band3} - 1.5 \times (\rho_{band5} + \rho_{band6}) - 0.25 \times \rho_{band7} \quad (2)$$

169 where  $\rho$  represent the reflectance values of Landsat OLI bands

170 After using the AWEI index, the target water area was estimated from 2001 to 2019. Landsat-8 OLI images obtained  
 171 in July 2019 were utilized to recognize the reclaimed area in the Toshka project. The Optimized Soil-Adjusted  
 172 Vegetation Index” (OSAVI), introduced by **Rondeaux et al. (1996)**, was utilized to detect the reclaimed zone in the  
 173 project by applying a threshold value of 0.2 and is estimated as by Equation 3:

$$174 \text{OSAVI} = (\rho_{nir} - \rho_r) / (\rho_{nir} + \rho_r + 0.16) \quad (3)$$

175 where  $\rho_{nir}$  and  $\rho_r$  are the reflectance values of near-infrared and red bands, respectively. Soil adjustment coefficient  
 176 (0.16) was chosen as a favorable value to diminish deviation with the soil background influence (**Rondeaux et al.**  
 177 **1996**). OSAVI outperforms NDVI as OSAVI and is formulated to diminish the sensitivity of NDVI to soil brightness  
 178 effect (**Steven 1998**). **Feyisa et al. (2013)** concluded that this index has an advantage in extracting water bodies in  
 179 areas with high environmental noise where other extraction indices fail to extract correctly. **Arshad et al. (2020)**  
 180 declared that the SAVI index is preferable to use in regions with relatively barren vegetation. In addition, **Mokarram**  
 181 **et al. (2015)** reported that OSAVI was derived as a modulation of the NDVI index to reduce the effect of soil brightness  
 182 in areas with sparse vegetation.

183 For Landsat-5 TM images, the LST of the research area was calculated using the formula established by the **Landsat**  
 184 **Project Science Office (2002)**. Digital number (DN) was converted to spectral radiance (L) using Equation 4:

$$185 L\lambda = (L_{MAX} - L_{MIN}) / 255 \times DN + L_{MIN}, \quad (4)$$

186 where  $L_{MAX}$ ,  $L_{MIN}$  are the spectral radiance of the DN value.

187 For Landsat-8 OLI images, the LST of the research area was estimated using NASA procedures on the thermal bands  
 188 of Landsat images obtained in 2019. Top of Atmosphere spectral radiance  $L\lambda$  is retrieved by Equation 5 (**Barsi et al.**  
 189 **2014**):

$$190 L\lambda = M_L \times Q_{cal} + A_L, \quad (5)$$

191 where  $M_L$  is a band-specific multiplicative rescaling factor from the metadata file  $Q_{cal}$  that corresponds to the thermal  
 192 band, and  $A_L$  is a band-specific additive rescaling factor from the metadata file.

193  
 194 For Landsat TM and Landsat OLI images, DNs were converted to reflection. The spectral radiance is converted to  
 195 brightness temperature (BT) in Celsius by Equation 6 (**Avdan and Jovanovska 2016**):

$$196 BT = \frac{K_2}{\ln[(K_1/L\lambda)+1]} - 273.15 \quad (6)$$

197  
 198 where  $K_1$  and  $K_2$  are band-specific thermal conversion constants from the metadata file. A regression model was used  
 199 to plot the correlation between LST and the area of Toshka Lakes in MS Excel.

## 200 Reference data

201 The high spatial resolution images provided by Google Earth™ and Landsat (NIR) bands were used as a reference  
 202 to distinguish between water pixels from non-water features. The selection of NIR band is because of its great  
 203 ability to differentiate water features from other land cover areas. The date of the Landsat imagery and the reference  
 204 data matched. A confusion matrix was created to quantify the accuracy of the AWEI index. The matrix contains values  
 205 for overall accuracy, user's accuracy, producer's accuracy, Kappa coefficient, Omission Error, and Commission Error.

206 **LST validation**

207 The estimated land surface temperature is compared with the actual air temperature data from meteorological stations  
208 to validate LST (Li et al. 2013; Mukherjee and Singh 2020). In this research, the LST values estimated from  
209 satellite images were validated by comparing them with the actual air temperature values measured simultaneously  
210 and derived from the Aswan Station 624140 (HESN). A flow chart of methodology is displayed in Fig. 2.

211 **Results and discussion**

212 By applying the AWEI to retrieve the surface area of Toshka Lakes, the results demonstrated that the total surface  
213 area of the four Toshka Lakes was about 1587.71 km<sup>2</sup> in 2001. By 2009, the surface area of the lakes reduced to about  
214 585.98 km<sup>2</sup> due to evaporation. In 2019, only one lake was left with an area of 69.92 km<sup>2</sup> (Table 1 and Fig. 3). The  
215 total shrinking of the lakes was 1517.79 km<sup>2</sup> between 2001 and 2019. Significant water loss in lakes can be related to  
216 evaporation, while limited amounts percolated into groundwater. Percolation into groundwater may be disregarded as  
217 the bedrock in the area is formed by a thick impermeable clayey layer (Fassieh and Zaki 2013). The drying parts of  
218 these human-induced lakes are converted to bare soil and covered with salt crusts. These crusts alter the  
219 physicochemical characteristics of the soil and pose tremendous environmental hazards. These effects include reduced  
220 plant water availability and soil organic matter content, plant dehydration, limited plant growth, and soil productivity,  
221 leading to soil degradation. Salt crusts can alter the environmental conditions of microorganisms and reduce the rate  
222 of organic matter decomposition in soil. These crusts are expected to reduce soil evaporation rate by covering the  
223 topsoil surface and increasing albedo, which resists moisture transfer from the soil surface. The accumulation of salts  
224 on the soil surface can also cause plant dehydration by reducing the osmotic potential and water availability to soil  
225 plants. In addition to plant dehydration, these crusts can damage the soil structure by disturbing soil aggregates. Salts  
226 affect crop production by reducing the nutrients uptake and restricting plant growth and reproduction. Compared to  
227 other studies, Chipman and Lillesand (2007) estimated that the total area of Toshka Lakes increased to 1740 km<sup>2</sup> by  
228 2001 but diminished to 900 km<sup>2</sup> by 2006.

229 In this research, AWEI index performs a good technique in the extraction of water features. The overall accuracy  
230 ranges from 93% to 99% during the study period. The Kappa coefficient ranges from 85.6% to 97.2%, which is in a  
231 high agreement with the reality (Table 2). The visual interpretation using high spatial resolution images supplied by  
232 Google Earth shows that the true boundaries of lakes coincide very closely the extracted boundaries using AWEI  
233 index. However, some omissions were found that led to errors in the results. The accuracy of this technique is  
234 influenced by many factors including sun angle, atmospheric condition, atmospheric correction technique, and water  
235 features properties (Feyisa et al. 2013).

236 The land surface temperature (LST) was estimated from Landsat images between 2001 and 2019 to address the impacts  
237 of Toshka Lakes dryness on the environment. The results clarified that the average LST was about 23.39°C in July  
238 2001. By July 2009, the average LST of this region increased to 42.34°C, while in July 2019, the average LST rose to  
239 approximately 48.41°C (0.6°C / y) (Figs. 4 and 5). This rise in temperature is due to the conversion of evaporated  
240 areas of lakes to bare soil covered with salt crusts. The rareness of precipitation in this arid environment may illustrate  
241 this rising in LST values. In comparison with water features, bare soil reflects the greatest rate of radiance flux and  
242 this explains the elevated values of temperature in the images (Carrasco et al. 2020). The actual air temperature was

243 obtained from Aswan Station 624140 (HESN) to validate the LST. **Table 3** displays the average air temperatures for  
244 LST from Landsat data and Aswan Station. Based on Aswan Station data, the highest mean temperature in July 2019  
245 was 38.1°C, and the lowest mean temperature in July 2001 was 34.4°C. Validation of LST estimated from Landsat  
246 images against recorded at Aswan station is shown in **Table 4**. A negative variation indicates that the LST estimated  
247 from Landsat images is higher than the Aswan Station air temperature. In contrast, a positive variation indicates that  
248 the estimated LST is lower than the station temperature. The maximum deviation for all estimated LST datasets in  
249 July 2017 was approximately -11.7°C. The minimum deviation for all estimated LSTs was 1.82°C that occurred in  
250 July 2005. This may be due in part to cloud coverage. In addition, the recorded data is not pertinent to the entire region,  
251 but only to specific regions where the meteorological stations are available. Therefore, it is difficult to obtain “correct”  
252 temperature data for the entire region.

253 The regression analysis showed a strong negative correlation between LST and the area of Toshka Lakes (R-square  
254 was about 0.98) (**Fig. 6**). **Pal and Ziaul (2017)** showed a negative correlation between the ratio of water bodies and  
255 average LST. . In spite of the fact that a certain part of LST change can be related to external climate change conditions,  
256 the results indicated that human-induced changes in land cover create considerable involvements on LST variation.

257  
258 Although Toshka Lakes were formed in recent years by diverting water from Lake Nasser, the drying of these lakes  
259 had serious impacts on the microclimate of this region. This study estimates that the temperature has risen by about  
260 6.07°C over the last decade, signaling a growing concern about global warming rates in the region. It is predicted that  
261 the temperature has increased by about 0.6°C/y between 2009 and 2019. Our findings comply with **Hereher (2017)**,  
262 who estimated that LST increased by about 0.4°C/y from 2003 to 2015 in the same study area. The difference in LST  
263 estimation was associated with many factors, including seasonal variation, land cover change, different study periods,  
264 climate change, and the resolution of satellite images. This study revealed that the dried parts of the lakes had a higher  
265 LST than areas which are covered with water before evaporation with lower temperatures. Global warming has many  
266 adverse impacts on the environment.

267 High temperatures increase soil evapotranspiration, reduce the moisture content of the soil, and promote the rate of  
268 mechanical weathering. Temperature rise due to the disappearance of these lakes enhances salt accumulation in the  
269 soils in these arid regions, reduces organic matter, and stimulates soil erosion and degradation. Increasing temperature  
270 is expected to decrease soil carbon storage by enhancing the microbial decomposition of soil organic matter. Thus, if  
271 the decomposition rate of soil organic carbon is greater than the growth of plants in arid environments, the soil is a  
272 source of carbon dioxide in the atmosphere. High temperatures can reduce soil water content and decrease crop  
273 production. Biological activity is also influenced by the low moisture content of the soil. The frequency of wind-  
274 induced soil erosion in arid regions probably increases due to climate change. In warmer climates, it is expected that  
275 the risk of wind erosion increases in arid and semi-arid environments because of the loss of biomass. Global warming  
276 can have serious consequences, especially in dry environments where precipitation is reduced, and evaporation rates  
277 are likely to increase, contributing to more drought conditions. In addition, high temperatures can affect biological  
278 systems by altering the adaptive capacity of many animal and plant species, thus increasing their risk of extinction.

279 High temperatures are expected to influence the dispersal of infectious diseases and increase human health hazards.  
280 People in these dry areas can be affected by changes in the temperature regime in these areas. As surface temperatures  
281 continue to rise, people are more likely to migrate from hot regions where air-cooling systems are expensive. In such  
282 areas, the risk of diseases escalates and threatens their lives and work. **Lozano-Parra et al. (2018)** pointed out that  
283 soil vegetation cover and moisture content lower soil temperature in dry environments.

284 The Toshka project, which aimed to create new arable land in the western desert, ended due to failure to reach that  
285 goal and many environmental issues. The OSAVI index was used to estimate the total area of land reclaimed in Toshka  
286 by 2019. By applying this index (threshold 0.2), the total land reclaimed in 2019 was only 41,659 acres (**Fig. 7**).  
287 **Hereher (2014)** detected that the total land reclaimed in Toshka by 2013 was approximately 30,000 acres by applying  
288 the SAVI index. This study estimated that by 2019, the land reclaimed in this project was 13% of the land that the  
289 Egyptian government aimed to reclaim by the Toshka project in 2017 (560,000 acres). A low supply of water and  
290 elevated temperatures reduce the vegetation density in the study area. A documented environmental impact assessment  
291 had to be conducted on-site before the project was launched. Assessing the potential impacts of a planned project on  
292 the environment helps define the achievability of the project. Improper planning of the Toshka project has caused  
293 innumerable environmental damage. The Toshka project must be treated as an existing reality. The mistakes made in  
294 it must be corrected and put back on the right path to achieve its goals.

295

## 296 **Conclusion**

297 This study uses Landsat TM/OLI images to investigate the effects of LULC changes on LST in the Toshka region.  
298 Current research addresses important issues by concretely analyzing human-induced changes and LST changes in one  
299 of Egypt's aridest environments. According to this study, the total area of Toshka Lakes decreased by about 1517.79  
300 km<sup>2</sup> from 2001 to 2019. In this period, the LST increased consistently (0.6°C/y). The results revealed that LST rose  
301 significantly in the lake areas that were dried by evaporation. This represents a high correlation between the land use  
302 change caused by these artificial lakes and the LST. Due to the high surface temperature, the area is vulnerable to  
303 natural hazards, diseases, severe droughts, and reduced soil productivity. These areas have also been observed to be  
304 covered with salt crusts, which may increase the risk of desertification. More areas are expected to be susceptible to  
305 wind erosion due to LST. This can adversely affect the survival of various habitats for environmental balance. The  
306 author recommends expanding the vegetation cover in the Toshka region to reduce further increases in LST and control  
307 ecological damage. This paper focuses on the benefits of satellite data for estimating human-induced changes and their  
308 impacts on microclimate. This helps land management experts and policymakers manage newly reclaimed areas and  
309 make decisions that limit the impacts of human-induced changes on LST dynamics and the environment. Therefore,  
310 it is highly required to impose appropriate plans to manage environmental changes in the Toshka region.

## 311 **Declarations**

## 312 **Ethics approval and consent to participate**

313 Not applicable

## 314 **Consent for publication**

315 Not applicable

316 **Data availability**

317 The datasets supporting the conclusions of this article are included within the article

318 **Competing interests**

319 The author declares that she has no competing interests

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322 **Authors' contributions**

323 Rasha M. Abou Samra contributed to study conception and design, data collection, analysis and interpretation of  
324 results, draft manuscript preparation and to the approving of the final version of the manuscript.

325

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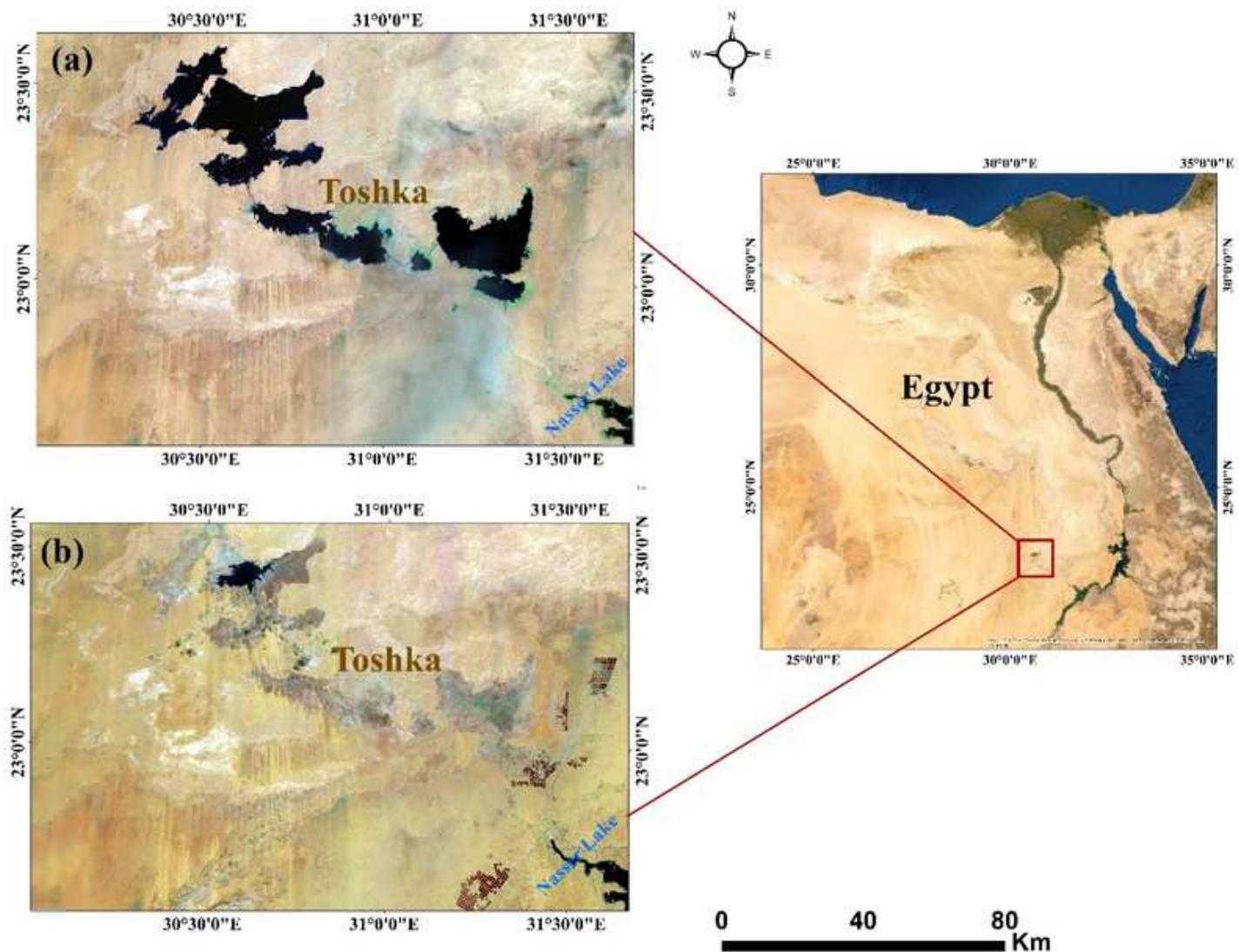
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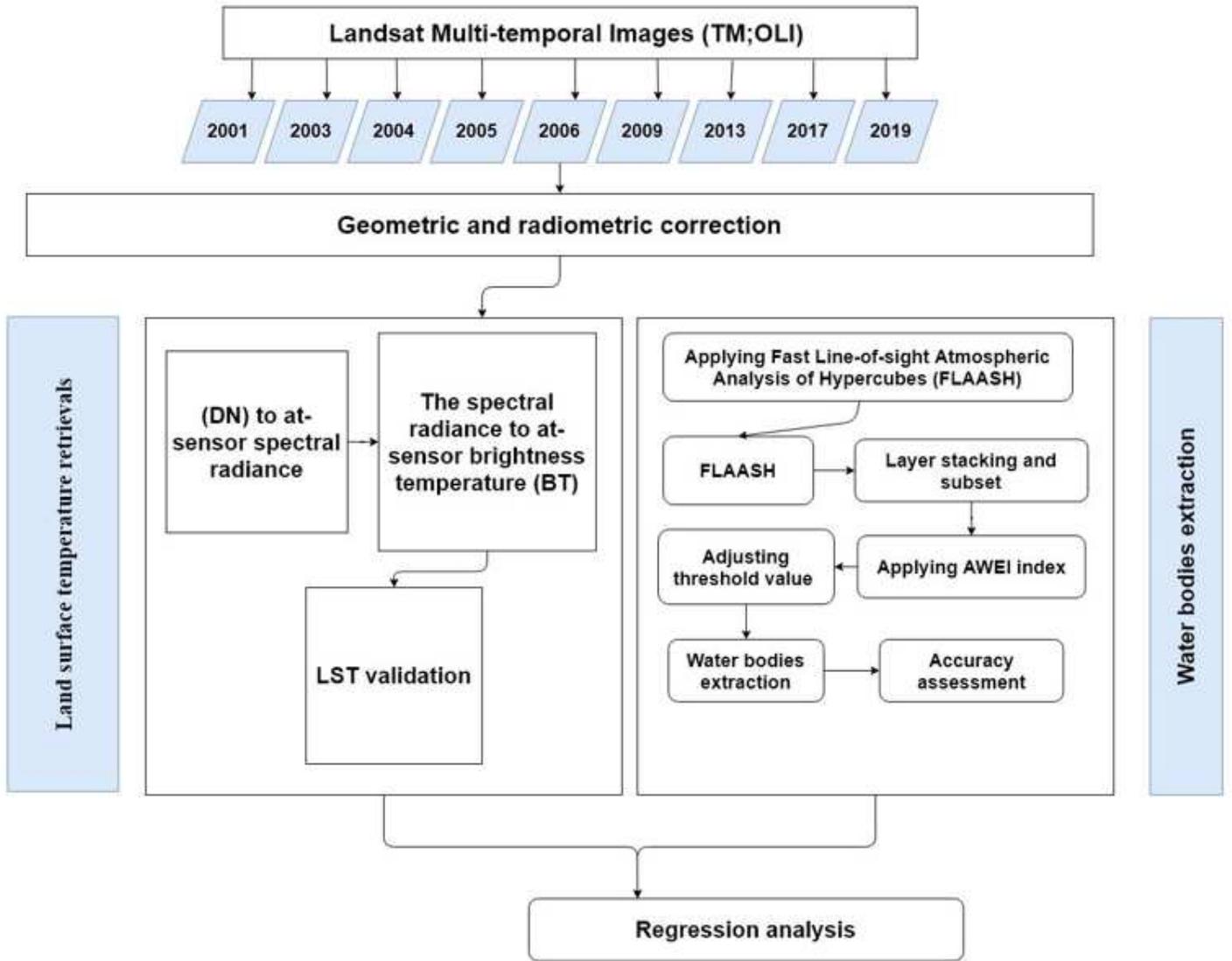
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# Figures



**Figure 1**

Location of the study area (a) using Landsat TM image (742) acquired on July 2001 (b) using Landsat OLI image (742) acquired on July 2019, brown dots in the south eastern corner are represented the reclaimed lands in the study area until 2019



**Figure 2**

Flow chart of methodology showing the steps used in this work

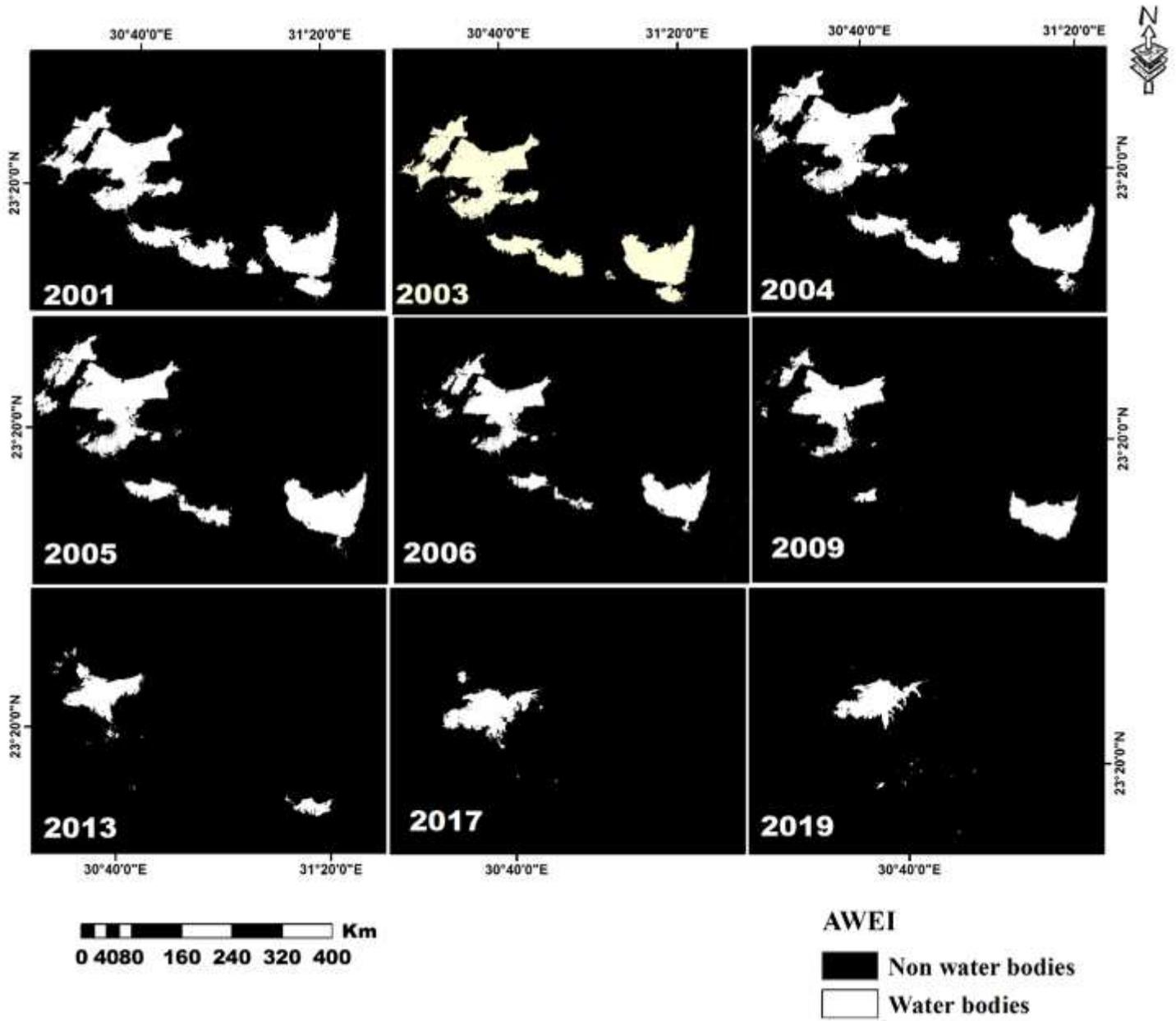
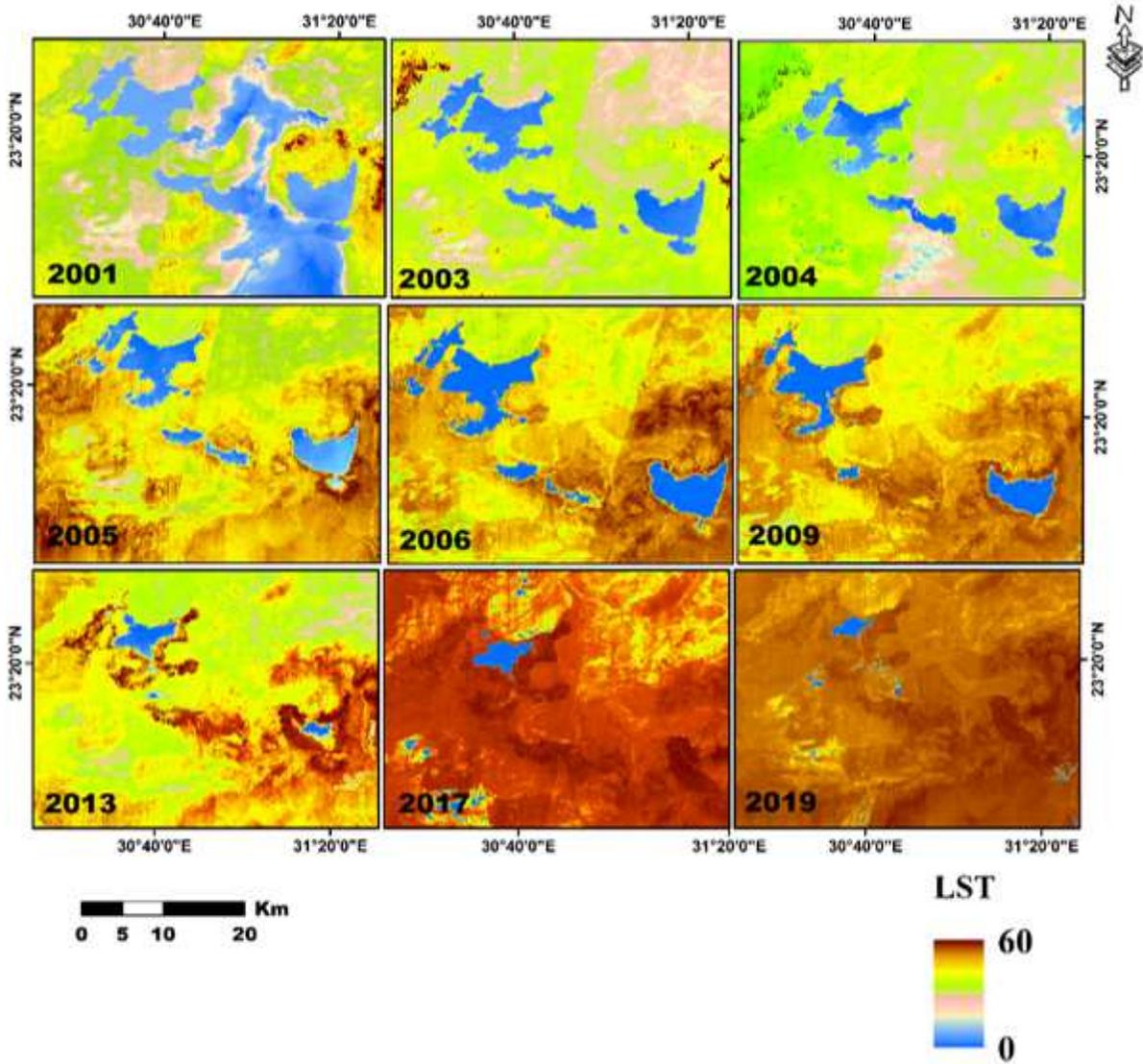


Figure 3

Comparison of area of Toshka lakes extracted from AWEI index between 2001 and 2019



**Figure 4**

Comparison of LST of Toshka lakes between 2001 and 2019. Note that higher LST are detected in the drying areas of lakes which are converted to bare soil by evaporation

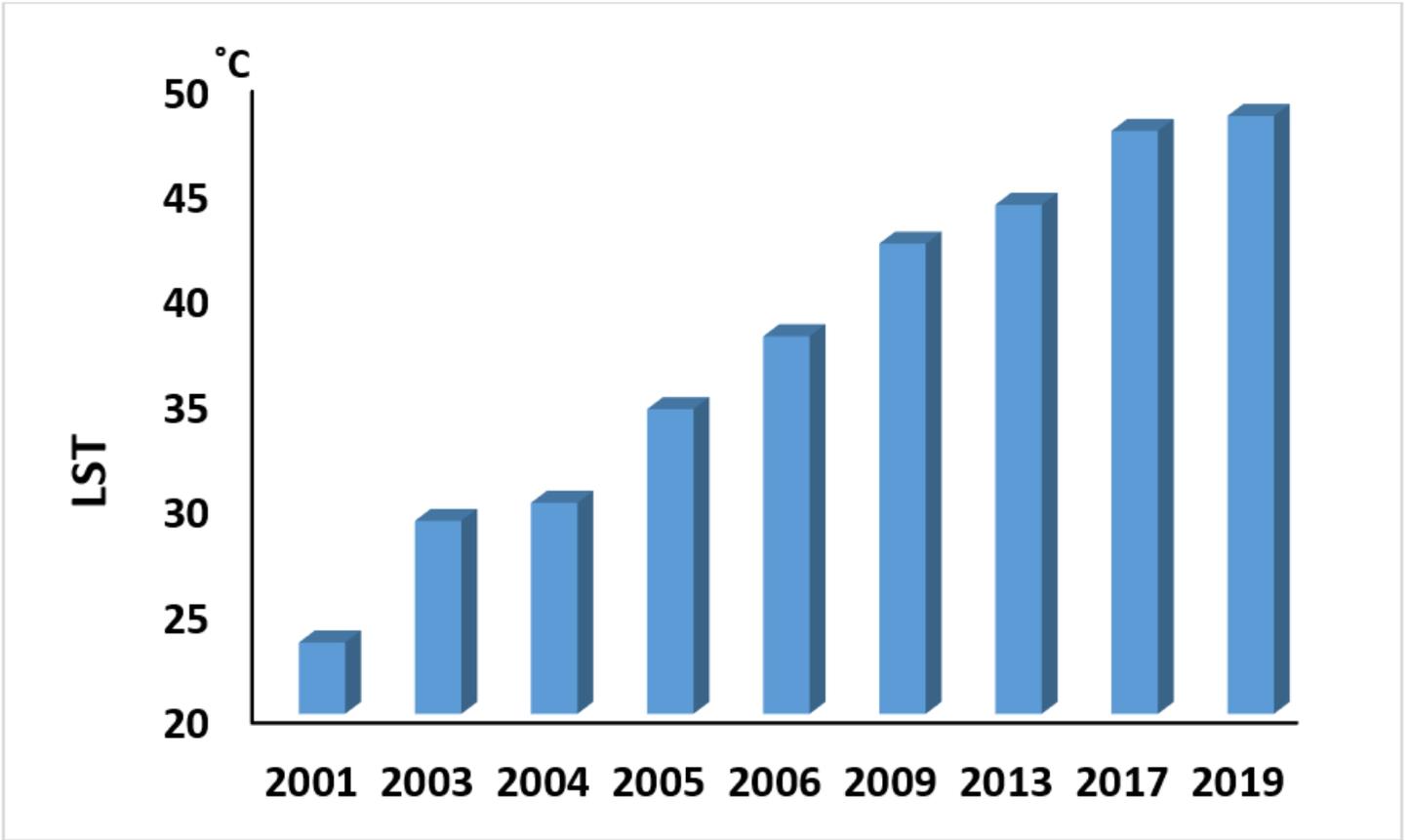


Figure 5

LST change of Toshka region between 2001 and 2019

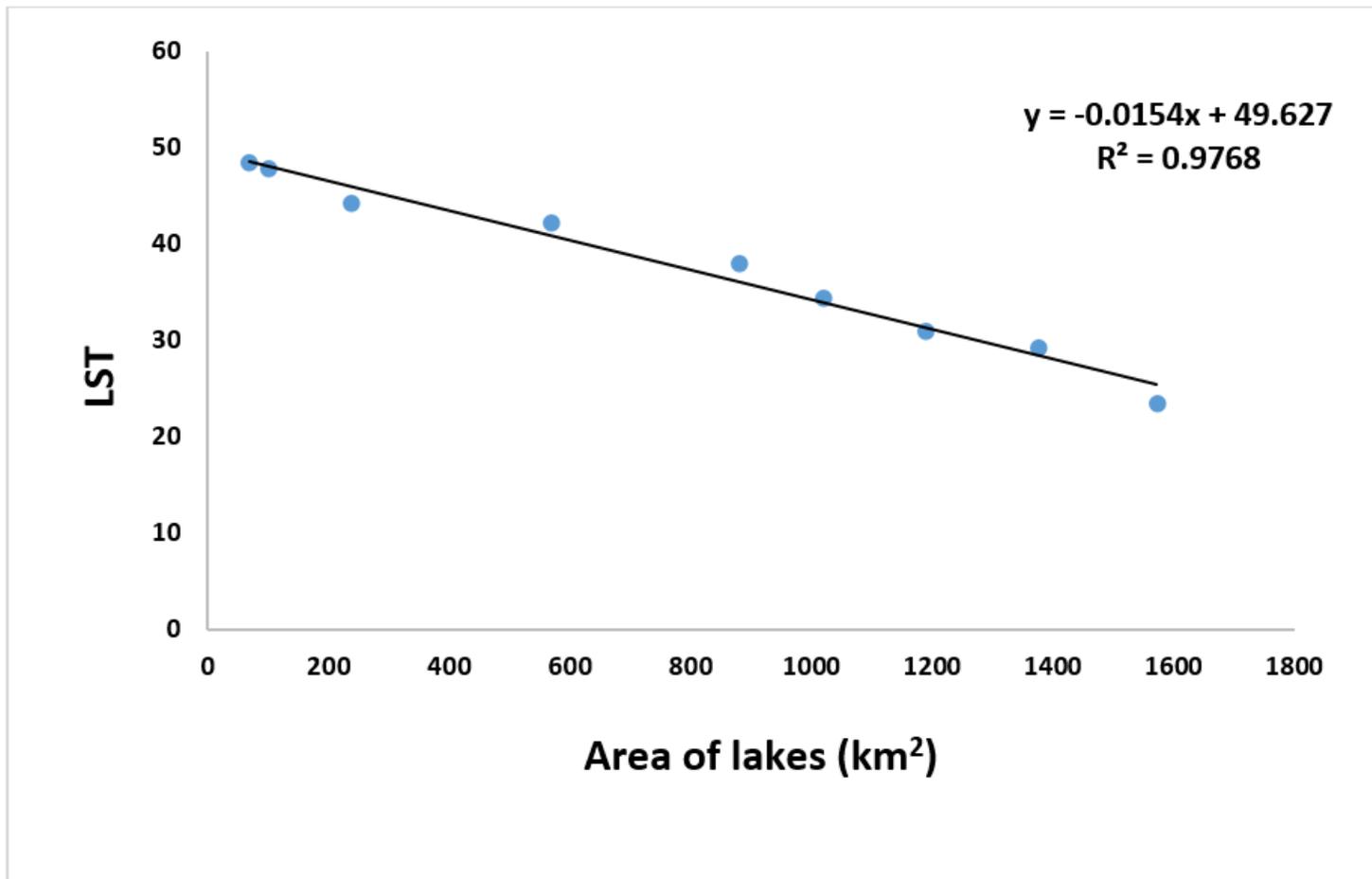
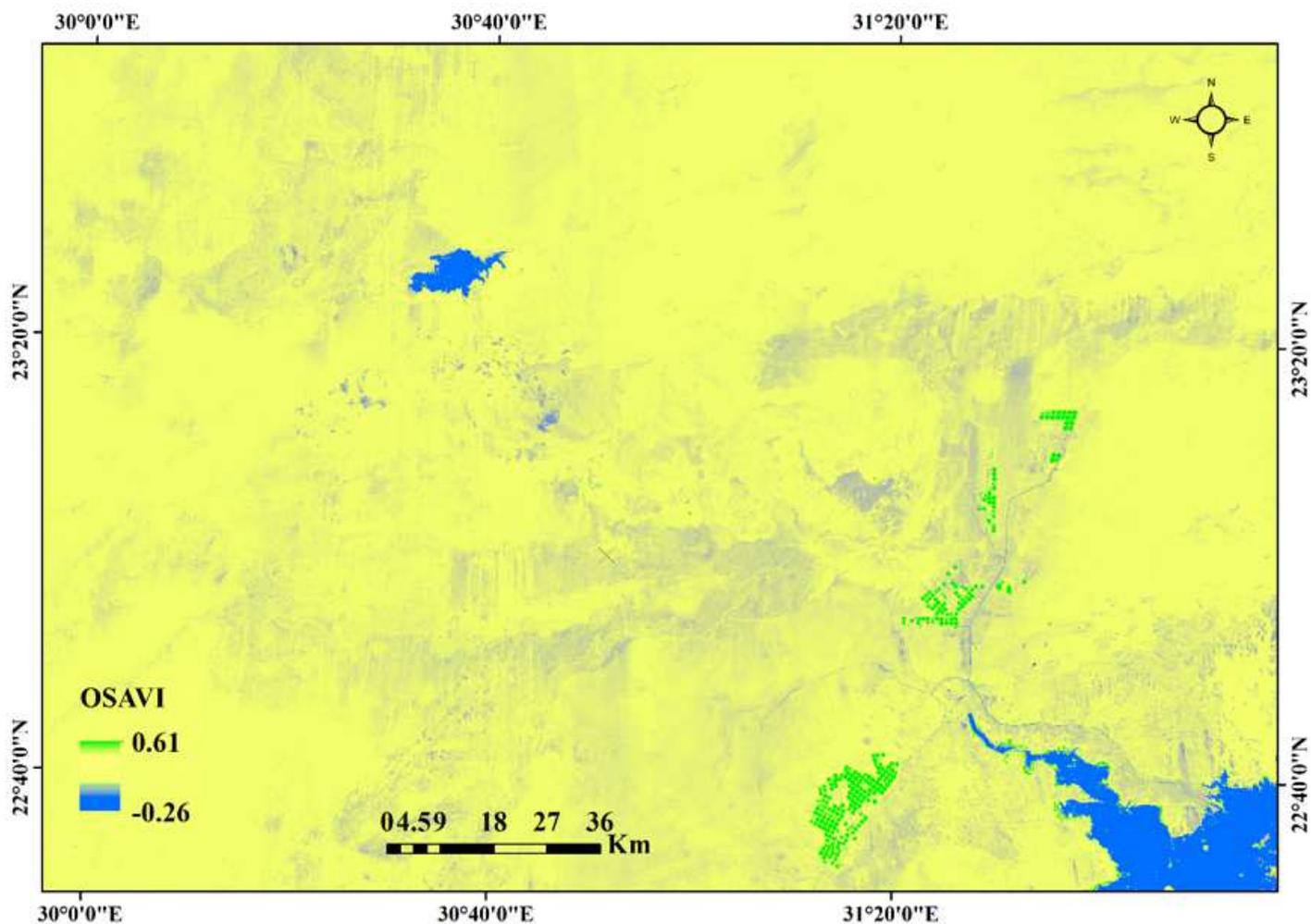


Figure 6

Regression plot between area of Toshka lakes and LST from 2001 to 2019



**Figure 7**

OSAVI image of Landsat OLI 2019 for Toshka region. Note that irrigated vegetation is represented by green color

## Supplementary Files

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