

Tracking Environmental Changes in Manzala Lake Integrating Landsat Data Based on Interpolation Nonlinear Models

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Abstract

Change detection is one of the most significant indicators that serve the sustainable development of the ecosystem. The aim of study is to make spatial interpolation prediction models for environmental changes of Manzala Lake based on Landsat images, different indices and statistical models. This aim is to assess the environmental changes in the Lake from 1992 to 2019. Vegetation cover was increased tremendously from 19.36 to 35.18 %. Water bodies were decreased from 52.45 to 45.50 % and bare lands were decreased from 28.19 to 19.32 %. Results indicated the order of models efficiency and accuracy as follow; quadratic > logarithm > Linear > Exponential. Consequently, the quadratic model was selected for assessing the environmental changes in the study area. Quadratic model was used to expect changes from 2020 to 2042. There will not be change in the area of the lake, when surface water and urban areas are equally within the studied area. The obtained data from used statistical models at the end of 2042 give assumption that the vegetation cover will disappear. While in the following period, models failed to describe what will be occur. This is ensured from the value of absolute error that less than one. The analyzed data for some physiochemical parameters showed different water quality between 2016 and 2019, as it is more suitable in 2019. MODIS/Aqua data indicate lower chlorophyll concentrations at 2018 and higher in 2015. The study can aid decision makers within solutions to manage and keep area and biodiversity of Lake Manzala using new technologies.

Keywords: Change Detection; Manzala Lake; Remote Sensing; Statistical Model; MODIS-Aqua

Introduction

Studying land use/cover changes of Manzala Lake is very important for sustainable development of biodiversity. Anthropogenic activities are responsible for controlling change in the uses of land. Assessment and detection of environmental changes serve the ecosystem of Manzala Lake. Ecosystem services contribute to human well-being in various ways and can be subdivided into provisioning, regulating and cultural services. Adverse impacts on socio-economic, biological, climatic and hydrological aspects of the environment may be related to land use/cover changes. Where environmental changes may be increased due to numerous sources of pollution. Change detection in land use/land cover is one of the most fundamental key issues to monitor the environment change and natural resource management, and critical factor to properly understand the interaction and the relationship between human activities with the environment. Also, change detection across spatial and temporal scales is indispensable to achieve precisely sustainable environmental management. Manzala Lake typically occupies a significant role for fish consumption by a human. Fundamental changes may also occur by the government of Egypt that drained large areas of the lake to convert it to farmland. Manzala Lake may be exposed to environmental changes within different impacts by Human activities. The lake changed from a connected regenerated saline water body to semi-closed sub-basins as far as the narrow water exchange inlets are not sufficient. Also, the lake typically receives large amounts of untreated sewage and irrigation returns [1]. Domestic, agricultural and industrial effluents are brought from urban centers along the lengths of main drains such as Bahr El Baqar through which more than 30% of the inflow passes to the lake [2]. Manzala Lake was chiefly affected by the continuous discharge of wastewaters. Recently, using remote sensing and modeling of environmental changes has been applied universally. Remote Sensing is the science of obtaining information about surface features and their creative process from a distance without direct contact with the earth. Remote sensing applications are good tools for instantly improving and detecting environmental changes. It was recorded that human activities, represented by wastewater discharge, have dramatically affected the quality of water. Domestic and agricultural activities in the surrounding area to Burullus Lake have remarked a great impact on the nutrient concentrations [3]. Remote sensing characterized by a high variety of spatial and temporal scales. Remote Sensing techniques have been successful

in mapping and detecting land use and land cover changes. The spatial and temporal distribution of land use/cover using satellite images is critically important to recognize and explain the phenomenon of global environmental change. Depending of remote sensing images, spatial Interpolation was applied for resolving many problems of prediction. One of the problems facing environmental changes is the accuracy of area accurately calculating because the data are missing or the considerable difficulty of measurement or it is impossible to make measurements for the continuous area studied because this work is costly materially or morally. These mathematical models are called spatial interpolation [4]. Therefore, spatial interpolation models can be defined as a set of statistical methods used to predict the values of phenomena in sites where accurate measurements are not available based on a limited number of measured points. Many researchers have typically published scientific studies on Lake Manzala with linear regression lines that predict the surface area of the lake as a relationship with time sees [5,6]. However, after several years and marked changes in the conditions of the lake, it became clear that the linear relations do not accurately describe the surface area of the lake. Based on a set of statistical criteria, the best spatial interpolation model can be determined precisely to represent the reality of the environmental changes in the study area. The aim of the present study is to positively predict the environmental changes in Manzala Lake in the future using spatial interpolation based on data acquired from remote sensing and GIS techniques.

Materials and Methods

Study Area

Manzala Lake is the largest coastal lake in Egypt, a shallow brackish lake extending between the Damietta Nile River branch and the Suez Canal with a maximum length of 50 km along the vulnerable Mediterranean coast. Manzala Lake is located at the northeastern part of the Nile Delta, a sandy beach ridge separate the lake from the sea. While it connected to the Mediterranean Sea through Boughaz El-Gamil inlets 1 & 2 (Figure 1). It is typically located between longitudes 31° 45' and 32° 22' E and latitudes 31° 00' and 31° 35' N. Km². The lake is shallow, typically ranging from 0.7 to 1.5 m in considerable depth. It connected to the North by the Mediterranean Sea, East by Suez Canal, South by El-Salam Canal and West by Damietta Branch of River Nile.

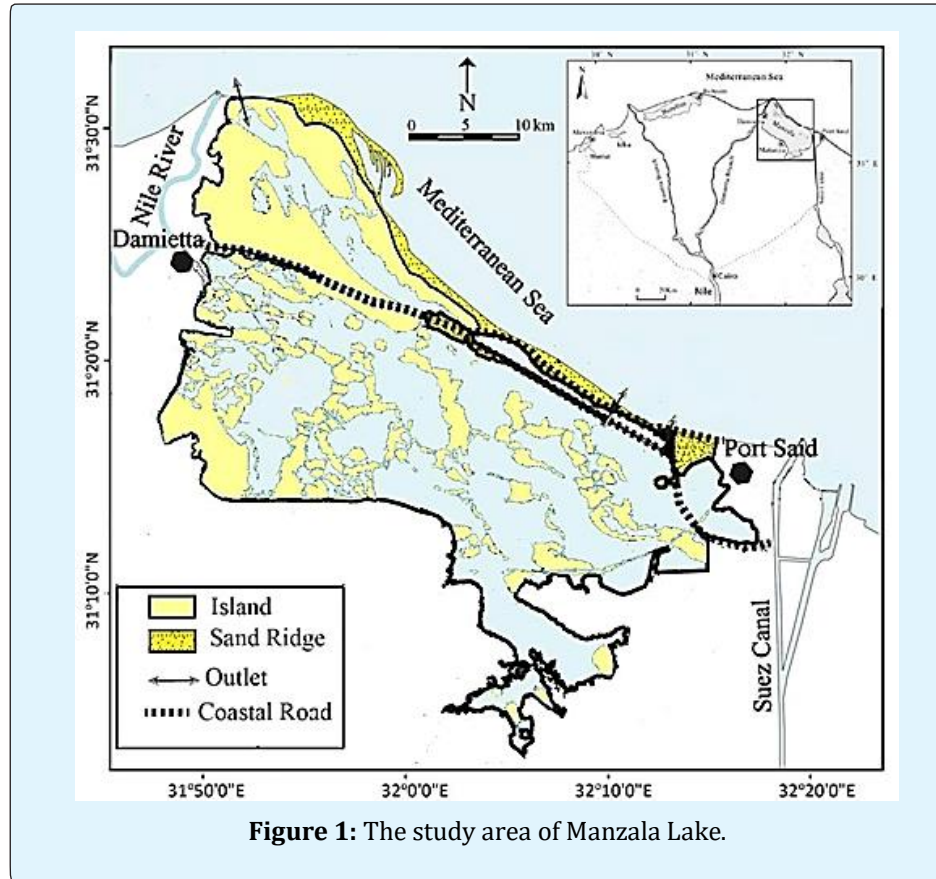


Figure 1: The study area of Manzala Lake.

Water Samples Analyses

Water samples collected at two periods (2016 & 2019) and are analyzed according to Table 1.

Parameter	Method & Reference
Depth and Transparency	APHA [7]
Temperature and Dissolved Oxygen	Lutron YK-22 DO meter
pH	Electrical-pH meter (Model Lutron YK-2001pH meter)
Conductivity (EC), Total dissolved salts (TDS) and salinity	Conductivity meter (Model Corning, NY 14831 USA)
Nutrients (NH_4 , PO_4 , NO_2 , NO_3 and SiO_4)	For ammonium determination, samples were fixed in the field. All are measured according the methods described by Grasshoff, et al. [8]

Table 1: Methods of water parameters analysis.

Remote Sensing Applications

Data Collection: Three Landsat images were acquired using <http://glovis.usgs.gov/>, Path: 176 and Row: 38. Landsat Thematic Mapper (TM), Enhanced Thematic Mapper Plus (ETM+) and Operational Land Imager and Thermal Infrared Sensor (OLI/TIRS) images were

acquired on 1992, 1994, 1997, 2000, 2003, 2006, 2009, 2014, 2017 and 2019, respectively have been selected for mapping and detecting environmental changes of Manzala Lake (Table 2). A field investigation was applied

universally and instantly confirmed typically using a topographic map with scale 1: 50,000.

Sensor	Year	Resolution (m)
TM	1992	30
TM	1994	30
ETM	1997	30
ETM	2000	30
ETM	2003	60
ETM	2006	30
ETM	2009	30
OLI	2014	30
OLI	2017	30
OLI	2019	60

Table 2: Characteristics of Landsat images from 1992 to 2019.

Image pre-processing and enhancement:

Preprocessing applications as; atmospheric correction, radiometric calibration, geometric correction was typically applied to all images for further interpretation and enhancement using ERDAS Imagine. The geographical coordinates are unified by the latitude and longitude coordinate system WGS_1984_UTM_Zone_36N, and the projection mode is the horizontal axis Mercator projection. After images correction, Principal component analysis (PCA) was applied to reduce interferences for sustainable development of change detection.

Image Classification: Maximum likelihood supervised classification was applied as the most accurate method in this present study to instantly detect different features of Manzala Lake. Supervised classification technique can studiously avoid digital problems resulting from the fundamental difference in the sensor and the atmosphere.

Land Use and Land Cover: Land use and land cover are one of the most important natural resources. Remote sensing is one of the most important mechanisms of monitoring changes at different time periods. Vegetation cover was applied using Normalized different vegetation index (NDVI) as equation 1. While water surface area was calculated from normalized different water index NDWI (equation 2) according to the Modeler Function in ERDAS. The values of these indexes varies from -1 to 1, based on the vegetation density and water cover areas, and resembles high when the value close to one. NDWI was applied using the ERDAS molder to sufficiently reduce any noise pixels [9]. It's difficult to carefully distinguish between built-up areas and bare lands in any area; therefore, an index to map bare lands in the selected

study area. The bare area can be detected using normalized built up area index (NDBI) (equation 3). The appropriate methodology comprised three arithmetic computations [10].

$$NDVI = \frac{Near\ infrared - Red}{Near\ infrared + Red} \quad (1)$$

$$NDWI = \frac{Green - Near\ infrared}{Green + Near\ infrared} \quad (2)$$

$$NDBI = \frac{SWIR - NIR}{SWIR + NIR} \quad (3)$$

Change detection analysis: Change detection determines the place and the considerable amount of fundamental change in a certain land cover at different times. Most of the change detection techniques rely on images classification, whether supervised classification or unsupervised classification. In this study, drilling and drying were applied to averaged time images between 1992 and 2019, to identify the occurred changes in Manzala Lake within different time.

Chlorophyll a concentration according to MODIS-Aqua:

The data of chlorophyll from 2002 to 2019 is available as the site of Giovanni v. 3.41. Acquired available data were selected in different years of 2005, 2008, 2012, 2015 and 2018 in 12 months. The selected region involved Manzala Lake with coastal area nearby the lake with coordinates of (31.7725 E, 31.0666 N and 32.3822 E, 31.5417 N). This data acquired from <https://giovanni.gsfc.nasa.gov/giovanni/>.

Statistical Models

Regression lines are helpful to know what interstitial values we could not practically know. It is useful to predict future values theoretically and can be trusted in statistical ways. The main advantage of these models is that they allow the reproduction of large-scale systems and long timescales. Statistical models vary from simple linear to complexity nonlinear depending on the nature of the phenomenon to be modeled. Four models were tested in this study for prediction of environmental changes: Linear, quadratic, Exponential and logarithm using Origin software 8.5.

Linear model: Linear model is used in different ways according to the context. The most common occurrence is in connection with regression models, and the term is often taken as synonymous with linear regression model. They can help you understand and predict the behavior of complex systems or analyze experimental, financial, and biological data. Linear regression is a statistical method used to create linear model.

A quadratic regression: It is the process of finding the equation of the parabola that best fits a set of data.

Exponential model: It is exhibited when the change the change per instant or unit of time of the value of a mathematical function of time is proportional to the function's current value, resulting in its value at any time being an exponential function of time. Exponential decay occurs in the same way when the growth rate is negative.

A logarithmic Model: It is a model that measures the magnitude of the thing it's measuring. It can also be seen as the inverse of an exponential model.

Assessment Methods: The accuracy of spatial interpolation models was assessed depending on three Statistical criteria: Root Mean Square Error (RMSE), coefficient of determination (R^2) and residual sum of squares (RSS). The model can be validated with low RSS, low RMSE and with R^2 value near to one.

RMSE: It is used as an essential indicator that assesses the accuracy of spatial analysis in geographic information system and remote sensing (Lower (RMSE) is mainly to give most reliable values in the areas with no data. The minimum (RMSE) calculated by Cross Validation can be used to find the best spatial interpolation model control parameters [11]. The root mean square error (RMSE) was calculated for each model prediction using the formula:

$$RMSE = \sqrt{\frac{\sum_{i=1}^n (f_i - O_i)^2}{n}}$$

R^2 : It is expressed by the ratio of total squares of regression divided by total squares. The value ranges between the correct one and zero and is calculated by the following equation

$$R^2 = \frac{[\sum_{i=1}^n (O_i - \bar{O})(f_i - \bar{f})]^2}{\sum_{i=1}^n (O_i - \bar{O})^2 \sum_{i=1}^n (f_i - \bar{f})^2}$$

RSS: it is also known as the sum of squared residuals (SSR) or the sum of squared errors of prediction (SSE) is the sum of the squares of residuals (deviations predicted from actual empirical values of data). It is a measure of the discrepancy between the data and an estimation model. A small RSS indicates a tight fit of the model to the data. It is used as an optimality criterion in parameter selection and model selection.

$$RSS = \sum_{i=1}^n (f_i - O_i)^2$$

Where n is the number of predicted value and f_i denoted to our estimation of observed value O_i . And so, \bar{O} and \bar{f} are mean values of observed and estimation respectively.

Results and Discussion

Lab Analysis

Tables 3 & 4 indicate the results of water samples in 2016 and 2019. While Table 5 showed comparison with other results in last periods. The data represent the periods at 2016 and 2019. It showed lower values in 2019 than in 2016 may attributed to more management in the lake area. As it agree with the statistical model as water body percentage is 45% and low percentage of vegetation than the other periods of last years. From the results of study in 2019, it's obvious that average values of pH is within the permissible limit of WHO [12] of 6.5-8.5. Phosphate average value in 2019 is within the limit of IPCB [13], which put a limit of 0.05 mg/l for reservoirs and lakes. Before applying the spatial interpolation models, the analysis of the data used should be performed using geostatistical techniques supported by GIS and remote sensing programs. Spatial models give more representative results if the data are distributed naturally and may result in distorted results if the data are abnormal.

No	Site	Depth	Trans	DO	Temp	pH	EC	Sal	TDS	Chl	NO ₂	NO ₃	PO ₄	SiO ₄	NH ₄
1	Elboughaz_1	140	20	11.9	18.9	9.33	57.3	37.3	34400	452.75	100.72	152.75	226.65	870	215
2	Elboughaz_2	70	20	8.8	19.7	8.94	69	46.4	42500	102.76	107.29	254.81	191.42	660	132
3	near Port Said	90	30	7.7	19.9	8.38	11.6	6.6	6465	90.15	305.92	1367	243.49	2095	108
4	Hadous	100	20	8	20.7	8.51	8.55	4.7	4640	865.95	103.54	247.7	200.61	5885	91
5	Bahr Elbaqar	70	20	2.6	21.2	8.22	9.25	5.1	5050	368.03	116.67	68	298.62	7545	396
6	Elmatrria	110	20	9.6	20.6	8.53	6.24	3.4	3300	646.61	131.69	476.65	209.8	5600	111
7	Elnasima	150	30	6.5	20.4	8.17	5.58	3	2930	18.27	80.08	368.93	183.77	4665	80
8	Temsah	70	20	11.7	20.8	8.82	20.6	12.2	11700	228.45	17.83	344.28	90.35	6115	55

Table 3: Physiochemical Parameters in water samples (2016) from different sites in Manzala Lake.

No	Site	Temp	pH	EC	Sal.	TDS	Chl	NO ₂	PO ₄	SiO ₄
1	Bahr Elbaqar	30.5	8.05	4.59	1.8	1710	101.8	52.238	27.26	1111
2	Elboughaz	30.4	7.87	37.1	15.9	14600	83.43	25.337	4.59	648
3	Elbashteel	30.5	8.19	4.87	1.9	1790	152.15	24.398	5.51	770
4	Eltemsah	30.5	7.92	46.6	20.6	18400	316	24.086	14.4	859
5	Legan	30.5	8.38	4.04	1.5	1430	241.87	27.214	28.18	865
6	Deshdy	30.5	8.31	7.95	3.2	3070	87.79	26.901	16.54	988
7	Elhmra	30.6	8.05	45.5	20	18000	61.13	25.337	5.51	808
8	Aboat	30.7	8.17	29.3	12.3	11600	51.75	26.588	8.27	1054
9	Elzerka	30.8	8.17	20.1	2	1970	71.46	24.086	4.59	1464
10	Elgenka	30.7	8.36	5.36	8.3	8040	234.34	-	20.52	774

Table 4: Physiochemical Parameters in water samples (2019) from different sites in Manzala Lake.

Parameter	Unit	Average of study (2016)	El-Sonbati et al. [14]	Elnaggar & El-Alfy [15]	EL-Shafei [16]	Elshemy [17]	Abd El-Hamid, et al. [18]	El-Mezayen [19]	Average of study (2019)
Depth	cm	100	150	130	-	-	-	-	-
Trans		22.5	-	32	-	-	25	-	-
DO	mgL ⁻¹	8.35	7.5	10.57	10.65	5.36	9.2	5.01	
Temp	°C	20.28	-	19.5	23.75	23.43	15.3	21.9	30.6
pH	-	8.61	8.03	8.3	8.18	8.3	8.7	8.42	8.15
EC	ms/cm	23.52	7.2	10.9	-	7.35	-	40.6	20.5
Sal	‰	14.84	4.7	-	32.25	-	15.6	25.4	8.75
TDS	mgL ⁻¹	13873.13	4520	-	-	-	-	23790	8061
Chl	µg L ⁻¹	346.62	-	38.9	-	41.9	163	-	140
NO ₂		120.47	126.1	82.03	305	-	180	-	28.5
NO ₃		410.02	136.75	232.27	1905	190	370	1.38	-
PO ₄		205.59	-	49.2	-	220	200	8.6	13.5
SiO ₄		4179.38	-	2241.7	-	-	560	-	934
NH ₄		149	457.83	753.67	215	-	50	4.52	-

Table 5: Comparison of study results (2016 and 2019) with other different studies in other years.

Trans: Transparency of Secchi Disk, DO: Dissolved Oxygen, Temp: Temperature, EC: electrical conductivity, Sal: salinity, TDS: Total Dissolved Solids, Chl: Chlorophyll

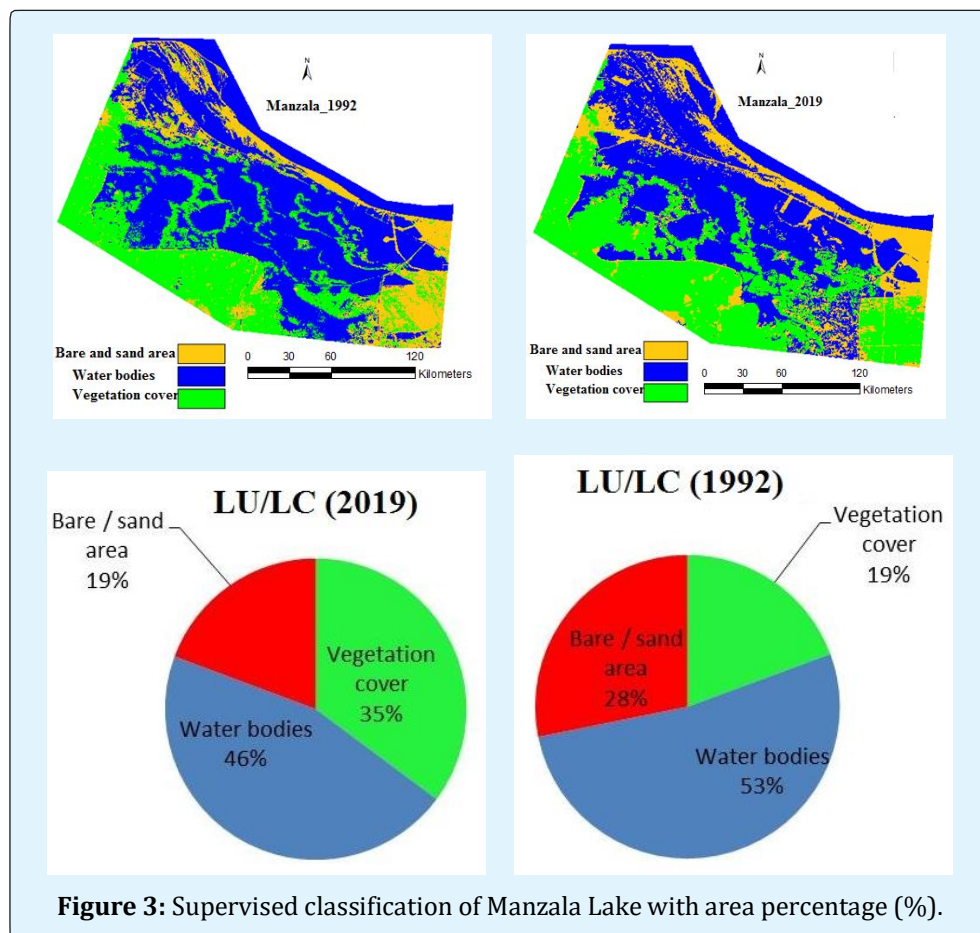
Remote Sensing Applications

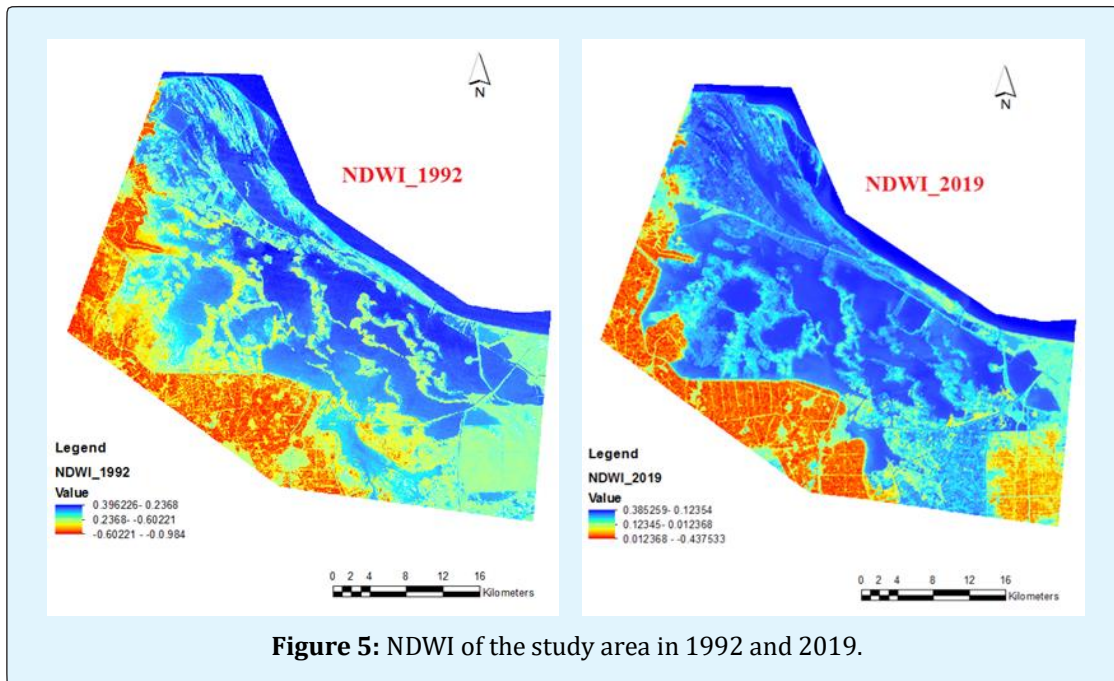
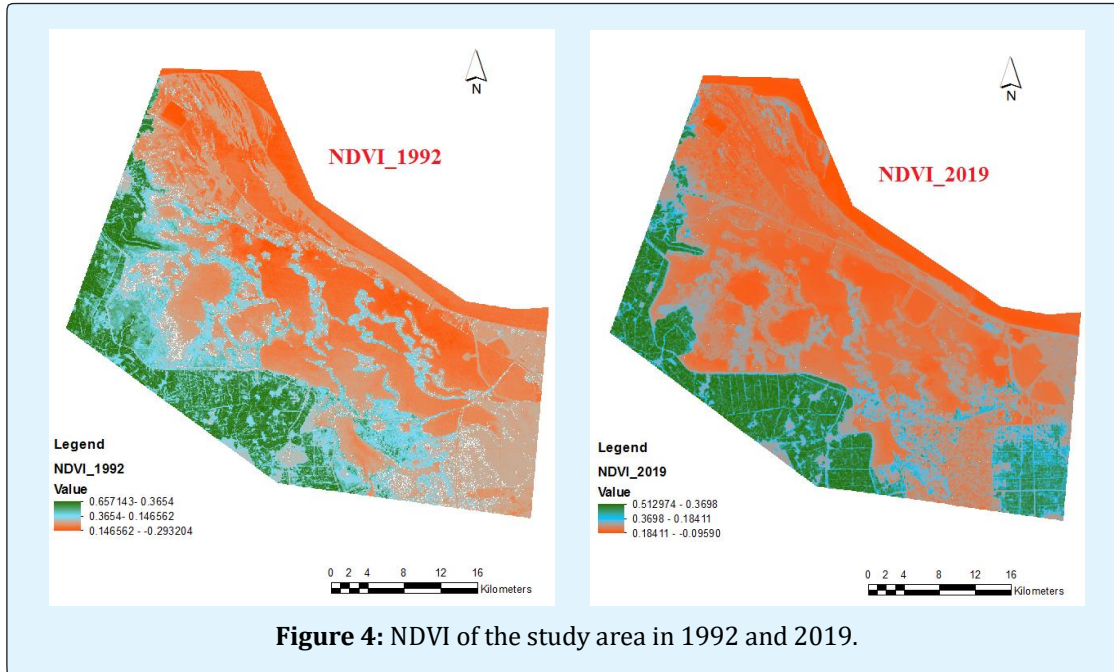
According to the topographic map and Google earth investigation, supervised classification was typically applied to show the unique features of the study area. In the present study, three classes were gently applied in 1992 and 2019. It was correctly classified into vegetation, water and bare/sand areas (Figure 3). The purpose of supervised classification is to assess the spatial distribution of each land cover class. Lake Manzala served as a significant source of biodiversity, but changes due to anthropogenic impacts have reduced the biodiversity of the lake. Remote sensing applications were applied for studying and detecting environmental changes. Table 6 shows the percentage of water, vegetation and bare/sand areas from 1992 to 2019. For vegetation cover, NDVI was

applied to all images from 1992 to 2019 (Figure 4). It was shown that the area of vegetation cover was increased from 19.36 to 35.18 %. This increase is due to the draining/drying processes at the south of Port Said. It is evident an intensive eutrophication process has been typically taken place in Lake Manzala through this period 1992 to 2019. The relative percentages increase the area of natural vegetation. Vegetation cover may be typically decreased. According to organic matter community which is related to change in temperature and precipitation, long-term climatic changes, human and animal activities. According to NDWI, water bodies decreased from 52.45 to 45.50 %. This can be mainly attributed to the reclamation process of the fish ponds inside the lake (Figure 5). Gradual deterioration in the water bodies of the lake due to the increased area of the floating vegetation. On the

other hand, bare lands have been decreased from 28.19 to 19.32 % (Figure 6). The marked decrease in bare land is a good indicator for the biodiversity of the lake and bird migration. This decrease in bare lands is probably due to increasing the area of vegetation cover. This decrease is due to continuous land reclamation projects along the southern and eastern shores of the lake and fish farming processes. Bare lands were like a result of the clearing of the lands for agricultural and building purposes. The gradual decrease in bare land is a good indicator for biodiversity of the lake and bird migration. This decrease in bare lands is probably due to increase tremendously the area of vegetation cover [20]. The government of Egypt drained substantial portions of the lake in an effort to convert its rich Nile deposits to farmland. The lake which was the biggest coastal wetland along the Mediterranean coast is moving toward its disappearance by two opposite forces; one of them is the shrinking of the water body by siltation of sediments typically coming from agricultural lands and the abundance of weeds and swamp vegetation as well as the drying practices for

agriculture, whereas the other force incorporates the removal of the coastal sand bar carefully separating the lake from the Mediterranean Sea by erosion, which should eventually lead to the conversion of the lake into a coastal embayment instead of being a closed coastal lagoon. Land reclamation has also reduced the lake surface by half, and, despite the declining quality of life and standards of living near the lake, human populations are increasing, exacerbating the lake's problems. According to change detection from 1992 to 2019, there are some gradual changes in this modern period (Figure 7). 70% of the total area not changed. Some drilling and drying processes applied in this period as a percentage of 12% and 14 %, respectively. According to vegetation cover increase, vegetation appears prominently in this period as a percentage 4%. According to pixel count and size two models for water and bare/sand areas were applied using ArcGIS 10.5. These models were assessed to predict the change detection from 1992 to 2019 (Figure 8). It showed that the coefficient of determination was approximately 0.92 which is close to 1.





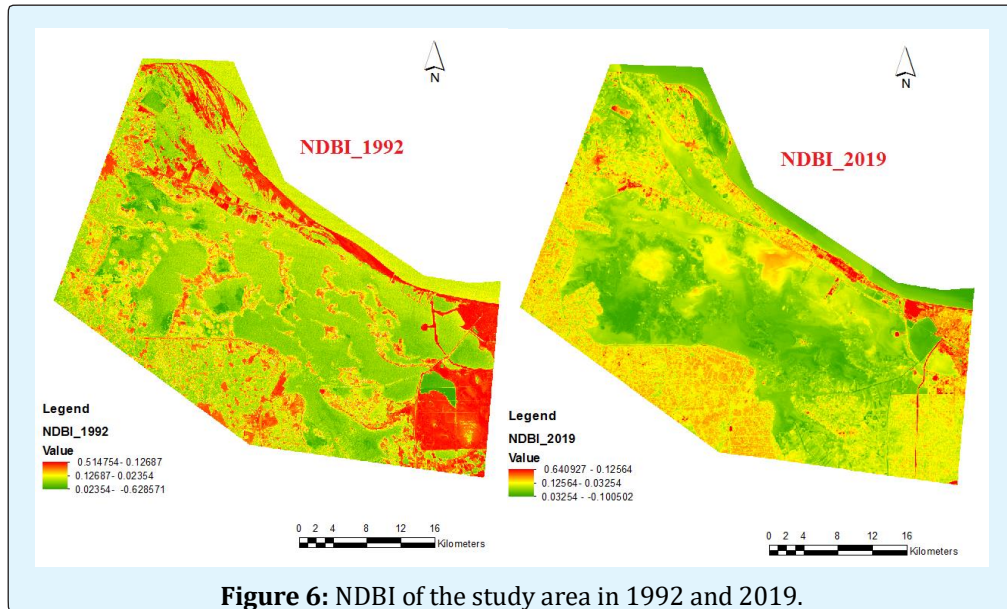


Figure 6: NDBI of the study area in 1992 and 2019.

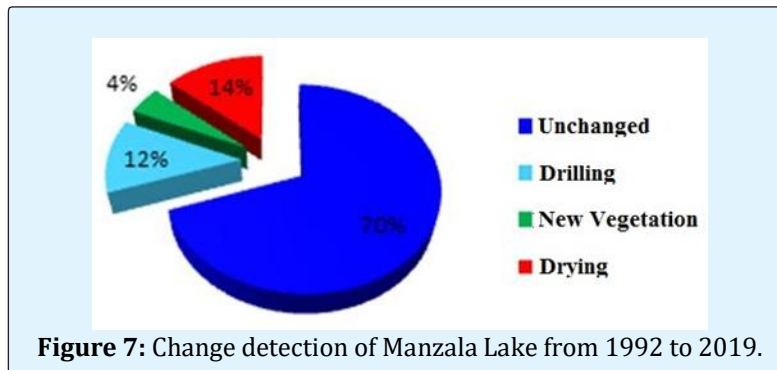


Figure 7: Change detection of Manzala Lake from 1992 to 2019.

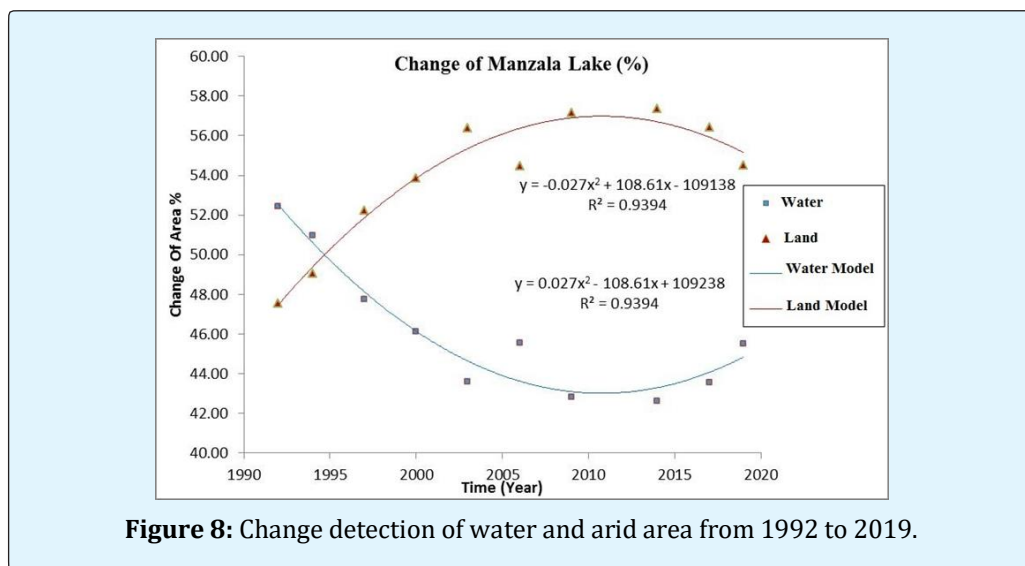


Figure 8: Change detection of water and arid area from 1992 to 2019.

Year	Area in Percentage (%)			Total
	Vegetation cover	Water bodies	Bare / sand area	
1992	19.36	52.45	28.19	100
1994	19.53	50.95	29.51	100
1997	27.19	47.77	25.04	100
2000	31.31	46.14	22.54	100
2003	32.71	43.6	23.69	100
2006	34.08	45.55	20.37	100
2009	40.73	42.85	16.42	100
2014	40.86	42.63	16.52	100
2017	40.02	43.57	16.41	100
2019	35.18	45.5	19.32	100

Table 6: Area percentage (%) for water, vegetation and bare area.

Chlorophyll a Concentration according to MODIS-Aqua

The data of chlorophyll recorded by MODIS-Aqua aid in the identification of area-averaged of Chlorophyll a concentration monthly. Also it aid in the knowledge of high and lower pollution impacts within different years. It is noticeable that the lowest concentration in selected years was at September, 2018 with concentration of 2.411 mg m⁻³, while the highest concentration was at November, 2012 with concentration of 8.675 mg m⁻³. Also the lowest mean annual concentration was 3.725 mg m⁻³ in 2018 and the highest mean annual concentration was 4.372 mg m⁻³

in 2015. The mean annual concentration of chlorophyll a within different months take this sequence 2015 > 2008 > 2005 > 2012 > 2018 as shown in Figure 9. More management aid in the decrease of pollutants and renew the water sources in this region may aid in the lower concentration obtained by MODIS-aqua Landsat observations in the year of 2018. The frequency of the increase and decrease in chlorophyll concentration may attributed to the different activities in the lake and monitoring systems and concern of the removal of pollutants, unwanted plants, renew water and keeping the water systems of the lake.

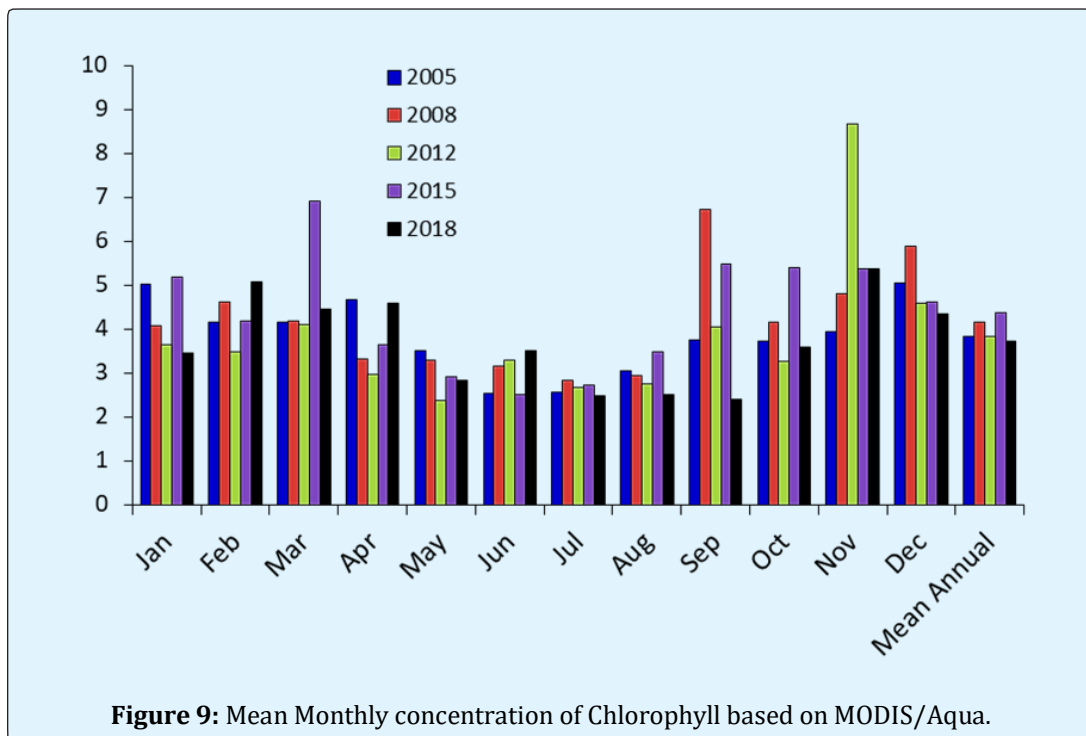
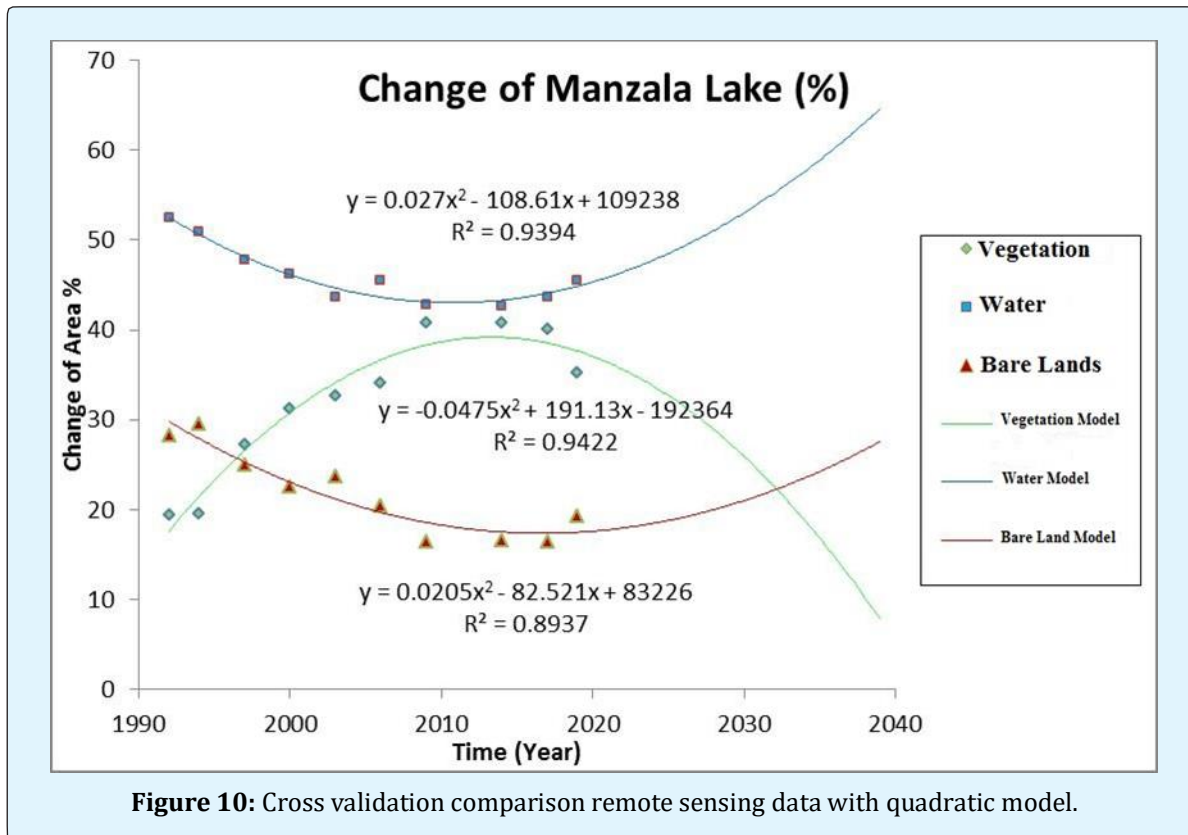


Figure 9: Mean Monthly concentration of Chlorophyll based on MODIS/Aqua.

Statistical Models

To compare the spatial interpolation models and the extensive selection of the best, the cross-validation technique provided by Origin software 8.5 was used to judge statistically the accuracy and statistical equations for different models were represented for accurate prediction of environmental changes in the study area as shown in Tables 7 to 10. The highest R^2 value and the lowest RMSE and RSS values of the quadratic model occurred. It was found that quadratic model is the best method to accurately predict the status of environmental changes in Manzala Lake. Quadratic model is the best model to obtain high quality in the proportional representation of changes data and typically produce the best equations for water, vegetation and urban in the study area. The results also indicated that the Exponential model is the lowest statistical model in the accuracy the prediction of environmental changes, and the models in

terms of preference in the following order: quadratic > logarithm > Linear > Exponential. According to the sufficient accuracy of these models, the quadratic model was selected for accurately assessing the environmental changes in the study area. By applying the best equations for water, vegetation and bare areas, the lake can be expected for 20 coming years as show in Table 10. The change in the area of the lake will be stopped when the sum of the surface area of the water lake with the area of the urban area equal to the total area of the study area. This naturally occurs according to these statistical models at the end of 2042, and the Green areas are almost completely disappeared in the study area. After that, the models fail to describe the change in the area of the Lake. The results in the table 10 show that the absolute error in these reasonable expectations is less than 0.1% (Figure 10).



Model	Equation	R ²	RSS	RMSE
Linear	$y = -0.2779x + 603.26$	0.6177	39.2264	1.9805
Quadratic	$y = 0.02701x^2 - 108.6205x + 109247.2$	0.9394	6.21398	0.7882
Exponential	$y = 6E6e^{(-0.006x)}$	0.6168	40.835	2.0207
Logarithm	$y = -558\ln(x) + 4289.1$	0.6194	39.0571	1.9762

Table 7: Statistical models for the surface area of water and its performance criteria.

Model	Equation	R ²	RSS	RMSE
Linear	$y = 0.73007x - 1431.77173$	0.7346	135.0881	3.6754
Quadratic	$y = -0.04746x^2 + 191.124x - 192357.17$	0.9422	33.11147	1.8196
Exponential	$y = 8.27936E(-6)e^{(0.00757x)}$	0.34678	332.4915	5.7662
Logarithm	$y = 1465.53113\ln(x) - 11110.97992$	0.73614	134.3075	3.6648

Table 8: Statistical models for the area of vegetable and its performance criteria.

Model	Equation	R ²	RSS	RMSE
Linear	$y = -0.4522x + 928.51012$	0.77845	41.16345	2.0288
Quadratic	$y = 0.02045559x^2 - 82.50423x + 83209.33$	0.8973	22.21736	1.4991
Exponential	$y = 2.85698E8e^{(0.00566x)}$	0.38346	114.5537	3.3845
Logarithm	$y = -907.49025\ln(x) + 6921.84973$	0.77957	40.95543	2.0237

Table 9: Statistical models for the area of bare and its performance criteria.

Class	Year				
	2020	2025	2030	2035	2040
Water bodies	45.39%	48.57%	53.09%	58.97%	66.20%
Bare/sand	17.60%	18.79%	21.01%	24.25%	28.51%
Vegetation cover	37.06%	32.70%	25.96%	16.85%	5.37%
Total area	100.05%	100.06%	100.06%	100.07%	100.08%
The absolute error	0.05%	0.06%	0.06%	0.07%	0.08%

Table 10: The predicted areas for the three classes during 20 years.

Conclusion

This study was applied to predict the environmental changes in Manzala Lake from 1992 to 2019 using remote sensing and GIS techniques. Vegetation cover was increased from 19.36 to 35.18 %. Water bodies were decreased from 52.45 to 45.50 %. Bare lands were decreased from 28.19 to 19.32 %. Based on Landsat images, models were analyzed: Linear, Quadratic, Exponential and Logarithm for the prediction of environmental changes. Quadratic model is the best method to predict the status of environmental changes in Manzala because this model has the lowest root mean square error (RMSE), the lowest residual sum of squares (RSS), and the highest coefficient of determination (R²) value. Chlorophyll a concentration according to data from MODIS/Aqua gives low mean annual concentration in 2018 and high in 2015. Water quality is more suitable in the period of 2019 than 2016.

Conflicts of Interest

The authors declare no conflicts of interest regarding the publication of this paper.

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