# **Chapter 4 Characterization and Classification of Soil Map Units by Using Remote Sensing and GIS in Bahar Al-Najaf, Iraq**



**Abdulameer S. Al-Hamdani and Hussein M. Al-Shimmary**

**Abstract** The Bahar Al-Najaf Lands is located in the west of the Al-Najaf Governorate. It is extended between  $31^{\circ}$   $39'$   $16''$   $-32^{\circ}$   $08'$   $08''$  N and  $43^{\circ}$   $47'$   $11''$   $-44^{\circ}$  $30'$  15" E with a total area of 2,000.2 Km<sup>2</sup>. This chapter aims to find out the possibility of using Remote sensing (RS) and Geographic Information Systems (GIS) techniques in contributing to soil surveys by selecting soil map units drawing and calculating spectral reflectance by satellite image of Landsat 8 provided with two Sensors; Operational Land Imager (OLI) and Thermal Infrared (TIRS). The image was acquired on 07/13/2014 from the USGS site and shows the nature and interference effect of the relationship between some of the soil physical and chemical properties on its spectral reflectance values by choosing false-color composite RGB (753), which is a specialist for studying soils and minerals and deducting study area using ERDAS imagine 2013. Some enhancements have been made (radiation, spectral and spatial) add to unsupervised classification, as well as using earth indicators (EIs). The used EIs include soil color, texture, natural plants and topography. These EIs are used in determining the researcher movement paths to select 16 Pedon's sites with 21 auger holes to a depth of 75 cm for surveying and isolating expected soil individuals within the study area which are identified using GPS. The unsupervised classification approach has shown considerable potential in the distribution of soil classes are close to supervised classification. The pedons revealed and described, according to the (Soil Survey Staff [1999\)](#page-31-0). Disturbed soil samples were selected from each horizon, air-dried, crushed and passed through sieve openings of 2 mm diameter for execution physical and chemical analysis. The results of particle size analysis show that there is no specific pattern distribution of soil particles whether within the pedon itself or between the pedons. There was no observed variations in textures classes for the vertical direction within the same pedon as in (P9, P4A, P10) or got little changes as in P4B. Horizontally variations observed among sites which returns to the nature of sedimentary additives and helps to occur the Litho-

A. S. Al-Hamdani  $(\boxtimes)$ 

© Springer Nature Switzerland AG 2020

A. M. F. Al-Quraishi and A. M. Negm (eds.),

*Environmental Remote Sensing and GIS in Iraq*, Springer Water, [https://doi.org/10.1007/978-3-030-21344-2\\_4](https://doi.org/10.1007/978-3-030-21344-2_4)

Kufa Technical Institute/Al-Furat Al-Awsat Technical University, Najaf, Iraq e-mail: [abdulameer.1957@yahoo.com](mailto:abdulameer.1957@yahoo.com)

H. M. Al-Shimmary Remote Sensing Center/University of Kufa, Najaf, Iraq

logical discontinuities in soils. All the pedon's horizons were strongly calcareous, while they varied in gypsum content in surface and subsurface horizons. Depending on the US modern classification, All the studied soils are within the Entisol order, which classified into two of Suborders. The first is Fluvents including great soil group Torrifluvents and subgroup Typic Torrifluvents which includes 9 Soil series ME1, DW56, TW964, DM44, MW3, TE354, DW124, DE47, DE126. The second is Psamments including two great soil groups, The first Torripsamments containing subgroup Typic Torripsamments including DE33, DE34, DE74, TE334, TW446 soils series. The other quartzipsamments include subgroup Typic quartzipsamments containing two soil series ME1, DM14. The soil series are classified according to the proposed (Al-Agidi [1976\)](#page-29-0). for alluvial Iraqi soils classification, It was diagnosed nine families and 15 soil series in the study area.

**Keywords** Remote sensing · Landsate · Soil image · Classification · GIS

### **4.1 Introduction**

Remote sensing (RS) is concerned with the use of devices to collect information about different objects and targets without direct physical contact with them by exploiting electromagnetic waves with their different wavelengths along the electromagnetic spectrum from short to high wavelengths and low frequencies (Elachi and Van [2006\)](#page-30-0). There is an urgent demand for rapid information regarding the classification of soils and their properties. Since there are no adequate studies in this information, thus RS is considered as one of the rapid techniques that can provide these requires using digital classification according to their spectral characteristics producing thematic map used in various fields of sciences especially agriculture (Yousif [2004\)](#page-31-1).

The Iraqi soils are different because of the variation in soil formation factors. Hence, they show different characteristics in both vertical and horizontal directions. Therefore, surveying and classification of these soils and isolation of their units are essential for the preparation of geomorphological and pedological maps. Soil mapping units, in fact, are modified units containing 85% of the characteristics of the nearest taxonomic unit. The spatial pattern and complexity of soil based on the interrelated effect of the soil formation factors at a given location and the extent of their variability within a given distance. Thus, the complexity becomes thornier as these variations become more substantial and on the basis of which the predictability of the pattern is determined at a particular classification level (Mutter [2008\)](#page-30-1). Given the urgent need to increase agricultural production to avoid food shortages in the world, especially Iraq, as well as the severe climate variability facing humanity. There has been a need to use RS and GIS applications due to their economic feasibility in the field of soil survey and classification by speeding up the work, minimizing effort, cost as well as giving them continuous geospatial values of earth characteristics, natural and temporal improvements. The process of linking the soil's physiochemical and morphological characteristics with its spectral properties helps predict a number of factors affecting the soil conditions and its components. Therefore, it is necessary to conduct the spectral properties measurements of the soil units in each region, which are useful aids in soil surveying and classification (Abbas [2010\)](#page-29-1).

The idea of the study is based on the use of RS techniques on the interactions between the fallen radiation from the electromagnetic spectrum with the soil surface. The reflected radiation depends on the soil's physical and chemical properties such as texture, structure, moisture, soil minerals, organic matter, gypsum, and carbonate which considered an essential feature in recording the spectral signature of the soil. As a result of these interactions, different types of soils can be characterized and separated (AL-Rajehy [2002;](#page-29-2) Zinck [2008\)](#page-31-2).

For the purpose of codification, all the efforts, cost and time with the accuracy of the separation and classification of Bahar Al-Najaf soils are compared to the traditional methods, which includes the drilling of many profiles and Auger holes, The study is carried out to take advantage of these modern technologies (RS and GIS) to prepare soil map units and achieve the following objectives:-

- 1. The effect of the soil formation factors on the soil variations of Bahar Al-Najaf region.
- 2. Separating soil units depending on the variation of the earth's surface spectral reflections. Based on the soil morphological, physical, chemical and spectral characteristics.
- 3. Soil classification may contribute to the optimum use of good soil for the production of the ideal, and how to expand it within the vision of sustainable agricultural development.

### **4.2 Geology of the Study Area**

Bahar Al-Najaf region is a transition zone between the sedimentary plain and the Western desert region, i.e. between the stable pavement and the unstable pier (Sallom and Segar [1994\)](#page-31-3). The study area is located within the stable zone (Al-Rutbha-Al Jazeera) based on the geological divisions of Iraq (Barwari and Slewa [1995\)](#page-30-2). The area is characterized by its simplicity and the lack of surface and sub-surface structures with the extension of the sediments of the Pleistocene and the discoveries of the Tertiary formation. In most parts of the region, they differ in their thickness and texture, where the deposit of valleys extends over areas adjacent to the seasonal valleys, The secondary gypsum deposits are spread over a very large area as a result of carbon rocks erosion belonging to the Tertiary period and the past (Barwari and Slewa [1995\)](#page-30-2).

In some areas, the spread and extension of sand dunes and sediment deposits are noted. Also Bahar Al-Najaf contains natural plants, as well as Palm belt, is extended along Tar Al-Najaf, and it considered a separate limit between Tar Al-Najaf and Bahr Al-Najaf depression. Figure [4.1](#page-3-0) shows the sequence of the geological strata of Al-Najaf Governorate.

| Era      | Period     | Epoch       | Age    | Formation                | Lithology      |
|----------|------------|-------------|--------|--------------------------|----------------|
|          |            |             |        | Aeolian deposits         |                |
| CENOZOIC |            | Holocene    |        | Valley fill deposits     |                |
|          | Quaternary |             |        | Depression deposits      |                |
|          |            | Pleistocene |        | <b>Gypcrete deposits</b> |                |
|          |            |             |        | Dibdibba                 |                |
|          |            | Pliocene    |        |                          |                |
|          |            |             |        |                          |                |
|          | Tertiary   |             | Upper  | Injana                   |                |
|          |            |             |        |                          |                |
|          |            |             |        |                          |                |
|          |            |             |        |                          |                |
|          |            | Miocene     | Middle | Fatha                    |                |
|          |            |             |        |                          | <b>NNNNNNN</b> |
|          |            |             |        |                          |                |
|          |            |             | Lower  | <b>Euphrates</b>         | ---            |
|          |            |             |        |                          |                |
|          |            |             |        |                          |                |

<span id="page-3-0"></span>**Fig. 4.1** Sequences of the geological formations of Al-Najaf Governorate from (Barwari and Slewa [1995\)](#page-30-2)

# *4.2.1 Geomorphological and Hydrological Phenomena of the Study Area*

### **4.2.1.1 Bahar Al-Najaf Depression**

The Bahar Al-Najaf is a prominent hydrological phenomenon in Al-Najaf Governorate and the Middle Euphrates region. This is the precursor of Al-Najaf city since it was founded twelve centuries ago (Batatu [2002\)](#page-30-3). It is located in the south and southwest of Al-Najaf holy city, and overlooking its plateau at a rate of about 130 m above sea level, within the range of the Iraqi sedimentary plain, The Bahar Al-Najaf lands covered, at the northwestern edge, with deposits of the Quaternary with thickness of  $(10-20 \text{ m})$  (Yacoub et al. [1981\)](#page-31-4). The area of the Bahar Al-Najaf depression is about 360 km2, and the height of its lowest point is 11 m above sea level. On the geological side, the rocky remains of the Anjana and the Dabdaba appear on its western edges (Al-Kadhimi et al. [1996\)](#page-29-3).

The lands of Bahar Al-Najaf penetrated by the valleys, which descend from the western territory of Saudi Arabia Kingdom and have a length of hundreds of kilo-

meters. Most notably the valleys of Shu'ayb Husub, Al-Kher, Al-Jal, Abu Talha, Al-Huwaimi. The waters of these valleys move in the winter and most of it flows into depression, a seasonal lake fed by water from the seasonal streams sloping from the high western parts as well as water from the Tar Al-Najaf after the rains and the surface drainage from all sides in the direction of Bahar Al-Najaf depression (Hassan [1983\)](#page-30-4). The sediments of the Quaternary period, such as wind and mud, cover the slopes. The bottom of the depression is covered with clay, silt, sand and fine gravel with a quantity of salts in the upper layers. In addition, there is a large quantity of modern shells in the upper layers extending up to areas close to the foot of the Tar Al-Najaf indicating the maximum intensities of water level rise (Barwari and Slewa [1995\)](#page-30-2).

#### **4.2.1.2 Tar Al-Najaf**

It is an insular line, about 65 km long, extended along the eastern and northern east edge of the study area and Al-Najaf plateau, which almost linearly cut. Its highest level is 188 m, but the point of contact with Tar Al-Sayyid, which lies within the borders of Karbala province is 177 m above sea level. One of the most important phenomena observed in Tar Al-Najaf is the depletion of water from some rock layers, Its groundwater stored in the Al-Najaf plateau (Al-Rawi [2007\)](#page-29-4).

#### *4.2.2 Classification of Iraqi Alluvial Soils*

Buringh [\(1960\)](#page-30-5) used the old American genetic classification system for classifying the Iraqi alluvial soils, identified as undeveloped soil of Azonal order and the Alluvial Soils for great group. Al-Agidi [\(1976\)](#page-29-0) proposed a system for the classification of alluvial soils at the series level. It is the first Iraqi classification system to classify these soils which used two classes to diagnose and distinguish the series: Soil texture within the effective series for soil depth of the 120 cm thick and between (30–150 cm) depth as well as the internal drainage is expressed in mottling depth. These two characteristics are associated with many other soil properties.

Al-Ani [\(2006\)](#page-29-5) studied the classification of agricultural land in the central region of the Iraqi alluvial plain, which represents the physiological units of the rivers and the irrigation channels prevailing in these areas, and based on the results of the morphological description as well as the physical, chemical and mineralogical characteristics of the studied soil pedons. It was found to be within the undeveloped soils of the Entisols order according to the modern American system, based on the presence of Ochric surface horizon and the absence of subsurface diagnostic horizons. The suborder were Fluvents, and Torrifluvents as great soil group and Typic Torrifluents as subgroup, These soils were classified into nine series based on the Al-Agidi [\(1976\)](#page-29-0) proposal, named: ME9, DW116, DW25, DE97, DM97, DE44, MW9, DW95, MF8.

In a study conducted in the Iraqi alluvial regions, Hamad [\(2009\)](#page-30-6) concluded that digital processing for satellite images show a new one that easily distinguishes soil types from other land uses and that the third, fifth and seventh spectral bands are suitable for determining soil units and water studies, With the eighth spectral package. Wahib [\(2012\)](#page-31-5) showed that soil salinity was significantly correlated with spectral bands 5 and 7 for Landsat 7 satellite with partial correlation coefficients (−0.4229, −0.4030), respectively in the Al-Salman depression, southern Iraq, and recommended using the values of reflectivity as an alternative to digital numbers because they were more accurate in expressing the results.

# *4.2.3 Remote Sensing*

The primary source of electromagnetic spectral radiation is the sun. When it reaches the earth's surface, it interacts with the atmosphere. Most of it is absorbed or dispersed by the atmosphere's components (water vapor and carbon dioxide and ozone), which prevents electromagnetic energy and specific wavelengths from reaching the earth's surface. The range of wavelengths in which the atmosphere is transmitted for energy, known as atmospheric windows, which also determines the type of sensors that can be used by the wavelength of these spectral bands.

### **4.2.3.1 Spectral Reflectivity of Soil**

A significant advantage of multi-spectral imagery is the ability to detect important differences between surface materials by combination spectral bands. Different materials may appear virtually the same within a single band. Selecting a particular band combination, various materials can be contrasted against their background by using colors (Yao et al. [2011;](#page-31-6) Zhang et al. [2017;](#page-31-7) Mustafa et al. [2019;](#page-30-7) Al-Wassai [2003\)](#page-29-6). The results showed a relationship between these characteristics and their reflectivity, depending on the degree of soil brightness (Katie and Mahimid [2002\)](#page-30-8). Studied types of desert soils in the Razzaza area, in Karbala Governorate/Iraq, They found that the sandy and gypsious soil showed the highest values of spectral reflectivity comparing to the rest soils found there.

Ben-Dor et al. [\(2008\)](#page-30-9) developed a method for characterizing soil profiles in the field by reading spectral reflectivity data for subsurface depths. It tested for moisture, organic matter, carbonate, free iron oxides, specific surface area for four sites in Palestine. This method is suitable for giving quantitative information to these pedons without digging trenches or sending soil samples to the laboratory. The results are for the studied sites only and need to be checked through the applications of the method in the future for other locations.

#### **4.2.3.2 Digital Classification of Spectral Images**

Classification using satellite imagery is one of the most essential techniques that contribute to accurate information about a given location in a short time frame. The classification aims to place all the pixels in groups according to their homogeneity and symmetry in the form of a classification map, through which the features, the types of the earth cover represented by these groups. Also, the accuracy in the production of such maps depends primarily on the accuracy of the process of classification of the study's images (Shuli [2008\)](#page-31-8). Generally, the digital soil classification has two main types; unsupervised and supervised classification.

#### Unsupervised Classification

An implemented automatically without supervision, depending on the degree of similarity and convergence between the spectral signatures without prior knowledge of the identity of these items in the study area. The computer handles the automatic processing of the assemblies and does not require the selection of training areas and is distinguished by the researcher's knowledge of the region, the success of this method depends on two crucial points:

- a. The number of spectral items assumed by the researcher must be sufficient to cover all visual parameters.
- b. This classification is exploratory and requires a little intervention according to our knowledge of the region through a comparison with reference data such as maps and others to make the interpretation successful.

#### Supervised Classification

It is implemented based on preliminary samples of the spectral signatures of each class and their statistical measurements (i.e., the classification is under supervision). Each unit of images is classified on the basis of its affinity and matching with the training samples in terms of spectral response and some statistical calculations. One of its methods is the maximum likelihood classification. It calculates the amount of correlation and variation of the spectral response of the training samples to each other and on the basis of the distribution of unknown pixels to those known on the assumption that the sampling units are systematically distributed and all the spectral bands used. The second method is the classification of the minimum distance to mean classification, where the points of location are chosen accurately. In order to give us a clear reflection, and in this case behave other pixels which unknown reversal (as an unknown goal). It takes a reflectivity of the shortest distance between them and the objectives approaching it. In the case of the using colors, it takes the color of the nearest target, so that the unclassified pixel is similar to the classification closest to the center, so-called (Mean center).

### *4.2.4 Accuracy Assessment of Digital Image Classification*

Numerous studies have carried out using both supervised and unsupervised classification methods. Some of which found that the unsupervised classification was more accurate (Borghuis et al. [2007\)](#page-30-10). However the vast majority of studies suggest the opposite is true (Alrababah and Alhamad [2006;](#page-29-7) Bahadur [2009;](#page-29-8) Mukherjee and Mukherjee [2009\)](#page-30-11). Accuracy assessment is a crucial part of the image classification process, as it evaluates the degree of acceptance between reference data and classified data (Tso and Olsen [2005\)](#page-31-9). Alrababah and Alhamed [\(2006\)](#page-29-7) achieved 69.1% overall accuracy of unsupervised classification without spatial improvements on satellite imagery and 73.7% accuracy with spatial improvements, while the overall accuracy of the supervised classification was  $(82.7, 78.8)\%$  with and without spatial improvements, respectively.

AL-A'araage [\(2012\)](#page-29-9) used the Landsat ETM  $+$  7 classification technology for Badra City in Wasit Governorate/Iraq, verified the validity of the classified results of five ground covers using 100 control points, the researcher reached an overall accuracy of 92.00% and the statistical Kappa coefficient of 0.8996. Ahmed and Ahmed [\(2014\)](#page-29-10) used supervised classification with maximum likelihood method of the Landsat TM satellite captured in 2007 for the land use-land cover map for an area of 5,700 Km**<sup>2</sup>** for Baghdad governorate, the data were classified into five categories. The classification accuracy was 93.7% and the statistical Kappa coefficient of 0.8833. Elkhrachy [\(2015\)](#page-30-12) used the unsupervised classification method for an area in Najran, Saudi Arabia, to separate three categories of ground coverings (urban areas, barren land, agricultural land) using Erdas imagine and ArcGIS. The classification accuracy of all varieties varied between (61–87%). In Delhi, India, Nain and Kumar [\(2016\)](#page-30-13) applied the unsupervised classification technology for Landsat  $ETM + 7$  using Erdas Imagine to classify ground coverings and the ArcGIS, with an overall accuracy of 91.67% and a statistical Kappa coefficient of 0.896, which means actual acceptance and good performance of the work.

# *4.2.5 Applications of RS in the Field of Iraqi Environment and Soils*

Qadir [\(2007\)](#page-31-10) studied the reflectivity of land use varieties and their relationship to soil characteristics to determine their suitability for planting using B1, B2, B3 bands for Landsat TM. Twelve types of soils were identified in the study area depending on the state of variation in their reflectivity values and indicated a weak correlation between soil reflectivity and its characteristics. Sulaiman and Aboud [\(2012\)](#page-31-11) conducted a semi-detailed survey in the center of the sedimentary plain south of Babylon province/Iraq to assess the land productivity. They dug 10 pedons representative of the soil series, morphological described and the soil texture estimated with some chemical characteristics. They reach that the soil of the study area is newly formed, characterized as stratified, medium to fine texture, and soils classified to the Entisols order and subgroups Typic Torrifluvent, Aquic Torrifluvent, Vertic Torrifluvent. The land assessment showed the presence of all land varieties in different proportions.

Azeez [\(2013\)](#page-29-11) studied the use of Geo informatics techniques (GIS, RS and GPS) using Satellite images of Landsat 5 and 7 to identify and isolate soil map units in an area of 4,087.3 Km2, Kurdistan Region, north of Iraq. Soil samples were taken from depth (0–30 cm) for 88 locations to determine physical and chemical properties. A total of 12 pedons were allocated to different map units in order to test the reliability of the soil map units. All pedons at the study area represent developed soils. The supervised classification was more harmony than unsupervised classification, as well as the soil map units in Arc GIS software package was more accordance with physical, chemical and morphological soil properties in comparison with supervised ERDAS classification. Several studies have been conducted on the environmental problems, such as drought events, land degradation, soil salinity, sand dunes encroachments in Iraq using satellite images, remote sensing techniques and GIS analysis in various areas of Iraq (Fadhil [2011;](#page-30-14) Wu et al. [2019;](#page-31-12) Fadhil [2009;](#page-30-15) Almamalachy et al. [2019;](#page-29-12) Wu et al. [2018;](#page-31-13) Al-Quraishi et al. [2019;](#page-29-13) Fadhil [2013\)](#page-30-16).

# *4.2.6 Soil Surveying and Classification Studies of Bahar Al-Najaf Region*

Yousif [\(2004\)](#page-31-1) used Landsat 5 TM images to classify Al-Najaf governorate soil using three spectral bands. The results of the supervised classification using Maximum Likelihood Classifier method gave a good representation of the soil classes with a total accuracy of 92% while the unsupervised classification gave a good representation of some soil classes and merged them. The study found that the color composition 754 was optimal for soil moisture sensitivity and recommended that the thematic map is the best in soil classification and representation of the ground truth compared to the traditional method.

Mahmoud and Mahimid [\(2011\)](#page-30-17) studied the inheritance and development of some gypsic soils in Iraq included Bahar Al-Najaf soils. They studied the morphological, physical and chemical properties of the pedons. They found that the Bahar Al-Najaf soil series KME 121 is undeveloped and with A-C horizons due to the dry climate, low vegetation cover, and relatively modern sedimentary parent material. All of which have helped to form undeveloped soil.

# **4.3 Methodology**

### *4.3.1 The Geographical Location of Bahar Al-Najaf Region*

The Bahar Al-Najaf Region Located in the west to southwestern part of Al-Najaf Governorate/Iraq, limited to longitude  $(43^{\circ}$  47' $11^{\prime\prime}$ –44 $^{\circ}$  30' 15" E) and latitudes (31 $^{\circ}$ 39'  $16'' - 32^{\circ}$  08' 08" N), and its area is estimated at 2,000.2 km<sup>2</sup> (Fig. [4.2\)](#page-9-0).

The study area is in the hot and desert climate. The summer is characterized by its length and dryness, winter is relatively short and cold. The rainfall rate in winter is low, with an average of 7.2 mm per year, and the annual accumulated rainfall depth of the 86.2 mm distributed irregularly. The precipitation is interrupted in June, July, August, and September, which is accompanied by a small quantity and irregularity of rainfall in the fall with the rise in annual temperature rates at a value of 25.5 °C with a sharp increase in the summer months to reach the rate of 37.7  $\degree$ C in July. So the soil of study area classified as hyperthermic type, because the annual average temperature is greater than 22 °C and the differences between the average temperature in summer and winter months are more than 5 °C. Its moisture system is of a Torric (Aridic) type due to the keeping of study area in dry conditions for more than 90 days on a continuous basis Where the amount of rainfall is less than double of the average monthly temperature, as well as the existence of a moisture deficit for an extended period of the year (Soil Survey Staff [1999\)](#page-31-0).



<span id="page-9-0"></span>**Fig. 4.2** Maps of location of Bahar Al-Najaf region, Al-Najaf governorate, and Iraq

# *4.3.2 FieldWork*

Traffic and mobility routes were identified to determine and dug 16 pedons, as well as 21 auger holes, were drilled within the studied area. Their locations were recorded using the GPS type Garmin: GPSMAP76CSx via (Free Lance Soil Survey), while relying on the unsupervised classification for satellite image with some enhancements, as well as the use of land guidelines and directories such as variations in soil texture, natural plants in addition to topography. We relied on region's altitude levels based on Digital Elevation Model (DEM) files, which downloaded from the Indian Institute of Remote Sensing website (IIRS) (Fig. [4.3\)](#page-10-0) (Table [4.1\)](#page-11-0).

For surveying and isolating the map units representing the soil individuals expected to be present in the study area. The GPS dataset was projected to the geodetic system of UTM\_WGS84\_38N and saved into a text file. This file was imported into Arc GIS10.3 desktop and converted to shapefile layer to be overlapped with the Landsat 8 images. The pedons are morphologically described according to (Soil Survey Staff [1999\)](#page-31-0). Soil samples from each horizon and auger holes surface depths were collected, air-dried, crushed to pass through a 2 mm sieve and analyzed for their physical and chemical properties, the results are shown in (Tables [4.2](#page-12-0) and [4.3\)](#page-13-0).



<span id="page-10-0"></span>**Fig. 4.3** Digital Elevation Model (DEM) of the study area

<span id="page-11-0"></span>

*NIR* Near Infrared; *SWIR*<sup>1</sup> Short Wavelength Infrared; *SWIR*<sup>2</sup> Short Wavelength Infrared<sub>2</sub> *LIR*<sub>10</sub> Long Wavelength Infrared<sub>10</sub>; *LIR*<sub>11</sub> Long Wavelength Infrared<sub>11</sub> http://fas.org/irp/imint/docs/ rst/Front/overview.html

### *4.3.3 Office Work*

### **4.3.3.1 Classification and Distribution of Soil Units in Bahar Al-Najaf Region**

The soils were classified into order, suborder, great group and family levels according to the (Soil Survey Staff [1999\)](#page-31-0). While for the series level under the proposed classification of Al-Agidi [\(1976\)](#page-29-0) for Iraqi alluvial soils (Table [4.4;](#page-18-0) Figs. [4.4](#page-20-0) and [4.7\)](#page-24-0).

#### **4.3.3.2 Remote Sensing Data Used in the Study**

The Landsat 8 satellite, which carries two sensors (OLI and TIRS) with 11 bands was adopted to capture study area image on 13-7-2014 was downloaded from the USGS site (Table [4.1\)](#page-11-0). The study area was deducted using the ERDAS Imagine to ensure a complete and accurate form for the study area. RGB (753) False Color Composite was chosen to distinguish terrestrial and geomorphological phenomena as being specialized for soil studies, separation of its units and minerals (Martinez-Rios and Monger [2002;](#page-30-18) Yacoub et al. [2008;](#page-31-14) Faleh and Shawan [2012;](#page-30-19) Al-Issawi et al. [2012\)](#page-29-14).

 $\mathbf{L}$ 



 $\sim$ 

<span id="page-12-0"></span>J.

4 Characterization and Classification of Soil Map Units by Using … 73



<span id="page-13-0"></span> $\overline{a}$ 



(continued)



76 A. S. Al-Hamdani and H. M. Al-Shimmary



4 Characterization and Classification of Soil Map Units by Using … 77

(continued)

(continued)



Table 4.3 (continued)

*CL* Clayey Loam; CL Clayey Loam; C Clayey; SCL Sand Clay Loam; SL Sandy Loam; LS Loamy Sand; L Loamy; S Sandy; SiCL Silty Clay Loam; SiL Silty Loam<br>CL Clayey Loam; C Clayey; SCL Sand Clay Loam; SL Sandy Loam; LS Loamy; Sand; L Loamy; S San Clayey; *SCL* Sand Clay Loam; *SL* Sandy Loam; *LS* Loamy Sand; *L* Loamy; *S* Sandy; *SiCL* Silty Clay Loam; *SiL* Silty Loam

<span id="page-18-0"></span>





<span id="page-20-0"></span>**Fig. 4.4** Classification of the study soil according to the modern American system Soil Survey Staff [\(1999\)](#page-31-0) and Al-Agidi's [\(1976\)](#page-29-0) proposal for sedimentary soil series

### **4.3.3.3 Satellite Image Enhancement Processing**

A series of enhancement processing was applied to the satellite image, which aimed to make the image's information more visible for purposes of visual interpretation. They included:-

Radiometric Enhancement

Means the removal of noise or interference of image elements because it affects the values of their spectral reflectivity, especially for red and near-infrared spectral bands.

Spectral Correction

Reflectivity is expressed by its true radiation by reversing the numerical numbers of the reflectivity (Digital Number, DN) to the actual radiation values, Top Of Atmosphere radiance (TOA). It considered as a radiometric treatment method which was important in the development of mathematical models that physically attach images data to ground truth (Lillesand and Kiefer [2000\)](#page-30-20).

### Spatial Enhancement

The integration of the spectral band 8 with other satellite bands, facilitated the acceleration of application and the maintenance of information accuracy and results to serve the purposes of the study in terms of the ability to distinguish well the types of soils in the image and to clearly demonstrate the process of integrating layers (Layer Stack) using ERDAS Imagine 2013.

### **4.3.3.4 Digital Classification of Image**

The classification methods include algorithms of both supervised and unsupervised classification. In both cases, the image was divided and classified into categories that give each category a specific color or code for thematic map production.

Unsupervised Classification

Unsupervised classification applied using the K-Mean method, to determine the varieties without external interference to obtain an overview of the differences in the reflectivity of the study area soils. The calculator performs the classification process with higher accuracy and less distortion in the isolation of ground parameters (Fig. [4.5\)](#page-21-0), because it works on the basis of pixels values represented by terrestrial



<span id="page-21-0"></span>**Fig. 4.5** Shows the unsupervised classification results of the study area



<span id="page-22-0"></span>**Fig. 4.6** Supervised classification to the study area soils based on pedons sites

features, which are separated and isolated by ERDAS Imagine 2013. This process was repeated several times to obtain satisfactory results (Lillesand and Kiefer [2000\)](#page-30-20).

#### Supervised Classification

Was carried out using the Maximum Likelihood method, after obtaining the results of samples analysis taken for the pedons sites as in (Fig. [4.2\)](#page-9-0). Training samples were then identified, which included all variances in the study area. The spectral signatures were taken for each training sample. This type of classification includes two stages (training and classification) (Fig. [4.6\)](#page-22-0).

### *4.3.4 Accuracy Assessment for Digitalmap Classification*

The accuracy of the classification was calculated by choosing 256 basis points, and the points outside the study area were excluded. Therefore, 140 points were adopted within the boundaries of the study area which compared with the points and locations of the classes values of the study area image which classified as supervised. The soil category code for each random point was fixed in the reference column of the accuracy assessment window. After the completion of the points, the final report was prepared to assess the accuracy of the classification, including the calculation of the following methods of accuracy: (Kappa coefficient, Users accuracy, Producers accuracy, Overall accuracy) (Table [4.7\)](#page-27-0).

### *4.3.5 Preparation of GIS Database*

This was done through the creation and implementation of the following measures.

#### **4.3.5.1 Entering Quantity Data**

Tabulated data containing the characteristics of spatial information, represented by the results of physical and chemical analysis of the study pedons were entered into Microsoft-MS Excel, and convert them into a database. So that, they can be joined later with the plan of the units and locations of soil pedons representing soil characteristics required in the production of spatial distribution of soil map units of the study area.

#### **4.3.5.2 Map Digitizing**

Performed by using the Arc GIS 10.3 software and implementing the screen digitization method for mapping soil units and pedons locations, and convert them into digital maps (shp.file) with two types of features: points representing the locations of pedons while polygons for soil individuals.

### **4.4 Results and Discussion**

# *4.4.1 Classification and Distribution of Soil Units in the Study Area*

The soils of the study area were classified at the levels of order, suborder, great soil group, subgroup and family according to (Soil Survey Staff [1999\)](#page-31-0). Moreover at a series level according to the proposed classification of Iraqi alluvial soils by Al-Agidi [\(1976\)](#page-29-0) (Figs. [4.4](#page-20-0) and [4.7;](#page-24-0) Table [4.4\)](#page-18-0), which depends on the presence of three different layers or textures except for the upper surface which goes to soil type. The thickness requirements of the first layer about  $35 + 5$  cm and  $40 + 5$  cm for the second and third layers (Table [4.3\)](#page-13-0). The soil series was named with the following symbols:

a. Number of layers: symbolized as (M, D, T) for (1, 2, 3) layers respectively.



<span id="page-24-0"></span>**Fig. 4.7** Distribution of the soil series of the study area by adopting the supervised classification

b. Natural drainage class: This indicates the movement of water in the soil depending on the depth at which the mottling appear via 50% or more of the test exposed surface.

The results of the physical and chemical properties in (Table [4.3\)](#page-13-0) indicate that The soils of the study area were undeveloped, classified as Entisols order. This is due to the nature of the environmental conditions prevailing in the study area as a result of the nature of the dry climate and the non-dense vegetation cover. These factors do not help the activity of the pedogenic processes responsible for the development of soil at an adequate level as well as the shorten chronological age of the parent material of these soils. The results were consistent with Mahmoud and Mahimid [\(2011\)](#page-30-17) when studying the evolutionary status of Bahar Al-Najaf soils. Since all these soils are located in the alluvial plain, which receives new deposits of sedimentary material at frequent intervals of time. This has helped to weaken the pedogenic activity and characterized as A-C horizons alternation (Buringh [1960\)](#page-30-5).

These soils were recognized by the presence of the surface horizon Ochric. It characterized by light color with low organic matter content and color value 4 or more in the moist state and 6 or more in the dry state, with the dominant of dry, moist system add to the Control section is not moist for most days in the year as well as the soil is unsaturated within the first meter of soil body. The soils order was classified into two suborders. The first Fluvents, which their parent materials are the result of added sediments, while the second Psamments which is located

within the desert region and the parent materials are ancient deposits rich in coarse particles represented by gravel and sand. The main soil groups are, torrifluvents, torripsamments, quartzipsamments and three sub great groups are typic torrifluvents, typic torripsamments, Typic quartzipsamments and nine families, as well as 15 soil series, were named: DE126, DE33, DE34, DE74, DE47, DM14, TE334, DW124, TE354, MW3, DM44, TW446, TW964, DW56, ME1. The largest soil series area DE47 Occupied 707.9 Km<sup>2</sup> from the study area, The smallest area of 10.31 Km<sup>2</sup> was for the DE126 series. The illustrations of these details are shown in (Figs. [4.4](#page-20-0) and [4.7;](#page-24-0) Table [4.4\)](#page-18-0).

# *4.4.2 Digital Image Classification Using Remote Sensing Techniques*

The most common methods are:

#### **4.4.2.1 Unsupervised Classification**

Figure [4.5](#page-21-0) shows the results of the unsupervised classification of the study area, comprising 17 soil classes as well as unclassified sites located outside the boundaries of the study area.

#### **4.4.2.2 Supervised Classification**

Figures [4.6](#page-21-0) and [4.7](#page-24-0) illustrate the classification method for the study area soils in addition to the final map sorted to the level of the series with their respective area in  $Km<sup>2</sup>$  units.

### *4.4.3 Accuracy Assessment of Digital Classification*

The classification accuracy technique is considered to be a quantitative analysis of the results obtained, It is necessary and important to assess the degree of error at the end of the work and to indicate the compatibility of the items titles in the thematic (classified) map with ground truths or reference data for the same sites (Jensen [2007;](#page-30-21) Taruvinga [2008\)](#page-31-15).

The primary objective of assessing accuracy is to estimate the elements and components of the error or confusion matrix and different descriptive measurements of terrestrial data within the field of study and gives a decision on the accuracy of the spectral data and the chosen method (Nain and kumar [2016;](#page-30-13) Stehman [2013\)](#page-31-16).

|                       | <b>Reference Data</b> |                |                |           |                  |                |                  |                 |    |                |                |              |                 |                |              |            |                |                                |
|-----------------------|-----------------------|----------------|----------------|-----------|------------------|----------------|------------------|-----------------|----|----------------|----------------|--------------|-----------------|----------------|--------------|------------|----------------|--------------------------------|
| Classified<br>data    | P <sub>2</sub>        | P <sub>5</sub> | P <sub>3</sub> | <b>P7</b> | P <sub>4</sub> A | P <sub>6</sub> | P <sub>4</sub> B | P <sub>14</sub> | P8 | P <sub>9</sub> | P10            | P16          | P <sub>12</sub> | P11            | P13          | <b>P15</b> | Row<br>total   | <b>Mapping</b><br>accuracy (%) |
| P <sub>2</sub>        | 1                     | 0              | 0              | 0         | 0                | 0              | 0                | 0               | 0  | 0              | $\mathbf{0}$   | 0            | 0               | 0              | 0            | 0          | $\mathbf{1}$   | 50                             |
| <b>P5</b>             | 0                     | 38             | 0              | 1         | 0                | 0              | 0                | 0               | 0  | 0              | $\mathbf{0}$   | 0            | 0               | $\mathbf{0}$   | 0            | 0          | 39             | 95                             |
| P <sub>3</sub>        | 0                     | 0              | 9              | $\Omega$  | 1                | 0              | 0                | 0               | 0  | 0              | 0              | 0            | 0               | $\mathbf{0}$   | 0            | 0          | 10             | 81.8                           |
| <b>P7</b>             | 0                     | 1              | 0              | 10        | 0                | 1              | 0                | 0               | 0  | 0              | 0              | 0            | 0               | 0              | 0            | 0          | 12             | 77                             |
| P <sub>4</sub> A      | 0                     | 0              | $\mathbf{1}$   | 0         | 3                | 0              | 0                | 0               | 0  | 0              | 0              | 0            | 0               | $\mathbf{0}$   | $\mathbf{1}$ | 0          | 5              | 43                             |
| <b>P6</b>             | 0                     | 0              | 0              | 0         | 0                | 7              | 0                | 0               | 0  | 0              | 0              | 0            | 0               | $\mathbf{0}$   | 0            | 0          | 7              | 87.5                           |
| P <sub>4</sub> B      | 0                     | 0              | 0              | 0         | 0                | 0              | 7                | 0               | 0  | 0              | 0              | 0            | 0               | $\mathbf{0}$   | 0            | 0          | 7              | 87.5                           |
| P <sub>14</sub>       | 0                     | 0              | 0              | 0         | 1                | 0              | 1                | 9               | 0  | 0              | 0              | 0            | 0               | $\mathbf{0}$   | 0            | 0          | 11             | 75                             |
| P <sub>8</sub>        | 0                     | 0              | 0              | 0         | 0                | $\mathbf{0}$   | 0                | 0               | 5  | 0              | 0              | 0            | 0               | $\mathbf{0}$   | 0            | 0          | 5              | 83.3                           |
| P <sub>9</sub>        | 0                     | 0              | 0              | $\Omega$  | 0                | 0              | 0                | 0               | 0  | $\mathbf{1}$   | 0              | 0            | 0               | $\mathbf{0}$   | 0            | $\Omega$   | 1              | 100                            |
| P <sub>10</sub>       | 1                     | $\mathbf 0$    | 0              | $\Omega$  | 0                | $\mathbf 0$    | 0                | 0               | 1  | 0              | $\overline{7}$ | 0            | 0               | 0              | 0            | 0          | 9              | 63.3                           |
| P <sub>16</sub>       | 0                     | 0              | 0              | 0         | 0                | 0              | 0                | 0               | 0  | 0              | 1              | $\mathbf{0}$ | 0               | 0              | 0            | 0          | $\mathbf{1}$   | $\mathbf{0}$                   |
| P <sub>12</sub>       | 0                     | 0              | 0              | 0         | 0                | 0              | 0                | 0               | 0  | 0              | 0              | 0            | 1               | $\mathbf{0}$   | 0            | 0          | 1              | 100                            |
| P <sub>11</sub>       | 0                     | 0              | 0              | 0         | 0                | 0              | 0                | 0               | 0  | 0              | 0              | 0            | 0               | $\overline{2}$ | 0            | 0          | $\overline{2}$ | 66.7                           |
| P13                   | 0                     | 0              | 0              | 0         | 0                | 0              | 0                | 0               | 0  | 0              | 1              | 0            | 0               | $\mathbf{1}$   | 20           | 0          | 22             | 87                             |
| P <sub>15</sub>       | 0                     | 0              | 0              | 0         | 0                | 0              | 0                | 1               | 0  | 0              | 0              | 0            | 0               | $\mathbf{0}$   | 0            | 5          | 6              | 83.3                           |
| Col.<br><b>Totals</b> | 2                     | 39             | 10             | 11        | 5                | 8              | 8                | 10              | 6  | $\mathbf{1}$   | 9              | 0            | 1               | 3              | 21           | 5          | 140            |                                |

<span id="page-26-0"></span>**Table 4.5** The error matrix for the soil series units of the study area from Landsat 8 image

Table [4.5](#page-26-0) shows the error matrix for fifteen soil units classified at the series level as suggested by Al-agidi [\(1976\)](#page-29-0) resulting from a supervised classification using maximum likelihoods classifier. The validity of this classification was based on 140 reference points selected within the study area out of a total of 256 random points using the ERDAS 2013. The error matrix provides information on the correct and incorrect prediction of the image classification by comparing the classified map with the ground information. It works on the basis of class-by-item comparison between source data known as ground truth and the corresponding results on the classified image (Martine—Rios and Monger [2002;](#page-30-18) Proklamasi et al. [2015;](#page-31-17) Nash [2016\)](#page-31-18).

The following methods are calculated for accuracy: Users accuracy, Producers accuracy, Overall accuracy, Kappa coefficient. Kappa coefficients are the most commonly used to measure the accuracy of the thematic map, as it shows the difference between actual agreement for reference data and the expected agreement of classified data by chance (Nain and kumar [2016\)](#page-30-13). Koutroumpas et al. [\(2010\)](#page-30-22) divided Kappa coefficient values into five categories of agreement classes (Table [4.6\)](#page-27-1).

Table [4.7](#page-27-0) shows that the producer accuracy of ME1, TW446, and TE354 soil series were 100%, this means that all the reference data pixels classified and located in the thematic map within those soil series. High values of producer accuracy (omission errors) as in the soil series DE47, DE33, TE334, TW964, DW56 are indicated that most of the terrestrial reference data pixels were classified in their corresponding positions in the classified thematic map. According to Koutroumpas et al. [\(2010\)](#page-30-22) classification. The accuracy of the soil series units map of the study area were classified as very good for soil series ME1, TW964, TW446, TE354, DW124, DE74,

<span id="page-27-1"></span>

| <b>Table 4.6</b> Accuracy ranges<br>of Kappa K <sub>hat</sub> coefficients | Agreement classes | Kappa K <sub>hat</sub> coefficient range |  |  |  |  |
|--|-------------------|--|--|--|--|--|
| and agreement classes by   | Poor              | $\epsilon = 0.2$                         |  |  |  |  |
| Koutroumpas et al. (2010)  | Fair              | $0.21 - 0.40$                            |  |  |  |  |
|  | Moderate          | $0.41 - 0.60$                            |  |  |  |  |
|  | Good              | $0.61 - 0.80$                            |  |  |  |  |
|  | Very good         | $0.81 - 1.00$                            |  |  |  |  |

<span id="page-27-0"></span>**Table 4.7** Accuracy evaluation of the error matrix data for soil series of the study area



Overall Classification Accuracy = 90.00% Overall Kappa Statistics  $= 0.8846$ 

DM14, DE47, DE33 and good for DM44, MW3, DW56, TE334 as well as Moderate for soil series DE34, DE126 (Tables [4.6](#page-27-1) and [4.7\)](#page-27-0).

Generally, it is noted that the overall accuracy of the soil units classification 90%, This means that the probability of 90% of the reference pixels is correctly classified within the classified map (Jensen [2007\)](#page-30-21). The value of the Kappa coefficient is 0.8846. This means that there is a statistical acceptance of 88.46% between the pixels or reference points and their corresponding on the classified map. According to the Kappa equivalent value and the classification proposed by Koutroumpas et al. [\(2010\)](#page-30-22). The overall accuracy of supervised classification of all sites in the study area falls within the category very good. It means, that is a very good agreement between ground truths and classified data.

## **4.5 Conclusions**

- 1. There is a weak activity of the pedogenic process affecting the soil formation of the study area. Within the full dominion of the undeveloped soils (Entisols) order, as well as the influence of geomorphic processes. The nature of sedimentation sources from the highlands in the western and southwestern parts of Bahar Al-Najaf Region affects the quantity and nature of the distribution of soil components vertically and horizontally between pedons.
- 2. The remote sensing and GIS techniques contributed effectively and distinctly the in identification and separation of soil units. The overall accuracy of the matrix error obtained from the supervised classification using was 90% as well as 0.8846 for Kappa statistical coefficient. This means a very good acceptance of the thematic map to classify studied soils.
- 3. The study demonstrated the importance of remote sensing and GIS techniques for studying natural resources, including soil, its characteristics, depending on the analysis of the satellite images digitally and visually. As well as economic feasibility in the field of survey and classification of soils through the reduction of effort, cost, speed up the completion of work and reducing the difficulties compared to traditional surveys.

### **4.6 Recommendations**

- 1. Emphasize the importance of using RS and GIS techniques in soil survey and classification studies because of the speed and accuracy of the work, preparation of soil survey and classification maps, add to reducing the effort and cost compared to field surveys.
- 2. Take advantage of remote sensing technologies and digital processing processes by selecting suitable spectral bands and integrating them into combinations that contribute to the study, detection and identification of desertification and land degradation areas.
- 3. In order to improve the results and to obtain the more specialized spectral signature curves as well as more details of the characteristics studied, especially in the field of surveying and classification of soils. We recommend using data from more than one sensor at the same time as their basis, or using images of sensors with high spectral and spatial discrimination capabilities that have multiple spectral bands with a narrow width (Hyper spectral imagery).
- 4. The need to use active remote sensing techniques to sense the internal soil characteristics, as well as the possibility of pairing with the passive technology to increase the efficiency of using remote sensing in soil studies.
- 5. The field measurements of the reflectivity are more accurate than the calculations of the satellite image because of many overlaps between the conditions and time of the images capture and the surface components of the earth coverings, which greatly affect the real values of their reflectivity. This requires calibration of the spectral reflectivity values calculated from the satellite images with the field reading using the spectroradiometer device.

### **References**

- <span id="page-29-1"></span>Abbas AH (2010) Characterization and classification of soil units of North Kut Project and prediction of some soil physical properties by using GIS and Remote Sensing. Ph.D. thesis, College of Agriculture, University of Baghdad, Iraq
- <span id="page-29-10"></span>Ahmed MA, Ahmed WA (2014) Integration remote sensing and GIS techniques to evaluate land use/land cover of baghdad region and nearby areas. Iraqi J Sci 55(1):184–192
- <span id="page-29-9"></span>AL-A'araage AA (2012) Monitoring desertification in Badra Area Eastern Iraq by using landsat image data. M. Sc. thesis, Geology Science, University of Baghdad, Iraq
- <span id="page-29-0"></span>Al-Agidi W (1976) Proposed soil classification at series level for Iraqi soils: I-Alluvial soils. Faculty of Agriculture, University of Baghdad, Technical bulletin No. 1
- <span id="page-29-5"></span>Al-Ani AM (2006) Applications of numerical taxonomy to classify some soil series of river levees in Iraqi alluvial plain. Ph.D. thesis, Faculty of Agriculture, University of Baghdad, Iraqi
- <span id="page-29-14"></span>Al-Issawi DF, Ibrahim MK, Sayel KN (2012) The use of heterogeneity of the Values of spectral reflectivity to separate the soil units of Al-Hawija area south of Al-Fallujah city/Iraq. Iraqi J Desert Studies 4(1):26–35
- <span id="page-29-3"></span>Al-Kadhimi JM, Sissakian VK, Sattar AF, Deikran DB (1996) Tectonic Map of Iraq, 2nd edn. scale 1: 1000 000. GEOSURV., Baghdad, Iraq
- <span id="page-29-13"></span>Al-Quraishi AMF, Qader SH, Wu W (2019) Drought monitoring using spectral and meteorological based indices combination: a case study in Sulaimaniyah, Kurdistan Region of Iraq. In: Al-Quraishi AMF, Negm AM (eds) Environmental remote sensing and GIS in Iraq. Springer Water
- <span id="page-29-12"></span>Almamalachy YS, Al-Quraishi AMF, Moradkhani H (2019) Agricultural drought monitoring over Iraq utilizing MODIS products. In: Al-Quraishi AMF, Negm AM (eds) Environmental remote sensing and GIS in Iraq. Springer Water
- <span id="page-29-7"></span>Alrababah MK, Alhamad MN (2006) Land use/cover classification of arid and semi-arid Mediterranean landscapes using Landsat ETM. Int J Remote Sensing 27(13):2703–2718
- <span id="page-29-2"></span>AL-Rajehy AM (2002) Relationships between soil reflectance and soil physical and chemical properties, M.Sc. Thesis, Mississippi State University, Mississippi, USA, p 75
- <span id="page-29-4"></span>Al-Rawi MB (2007) studying the Hydrologic and morphometric properties of Bahr Al-Najaf Basin using Geographic Information system. M.Sc. thesis, College of Education (Abn-Rushd), University of Baghdad, Iraq
- <span id="page-29-6"></span>Al-Wassai FA (2003) Comparison between different method of satellite merge resolution. M.Sc. thesis, University of Baghdad, Iraq
- <span id="page-29-11"></span>Azeez SN (2013) Used Geo informatics techniques (GIS, RS and GPS) for identification and isolation of soil map units. In Garmyan district, Kurdistan Region—north of Iraq, Ph.D. Thesis, University of Sulaimania, Iraq
- <span id="page-29-8"></span>Bahadur KC (2009) Improving Landsat and IRS image classification: Evaluation of unsupervised and supervised classification through band ratios and DEM in a mountainous landscape in Nepal, Remote Sensin*g* 1(4):1257–1272
- <span id="page-30-2"></span>Barwari AM, Slewa NA (1995) Geology of the Najaf plate, general company for geological survey and mining. Ministry of Industry and Minerals, Report No 20
- <span id="page-30-3"></span>Batatu HT (2002) The old social classes and revolutionary movement of Iraq. A study in Iraqs old landed and communist pp 88–91
- <span id="page-30-9"></span>Ben-Dor E, Heller D, Chudnovsky A (2008) A novel method of classifying soil Profiles in the field using optical means. Soil Sci Soc Am J 72:1113–1123
- <span id="page-30-10"></span>Borghuis AM, Chang K, Lee HY (2007) Comparison of automated and manual mapping of typhoontriggered landslides from SPOT-5 imagery. Int J Remote Sens 28(8):1843–1856
- <span id="page-30-5"></span>Buringh P (1960) Soils and soil conditions in Iraq, soil survey and classification specialist, Ministry of Agriculture, Baghdad, Iraq, p 322
- <span id="page-30-0"></span>Elachi C, Van J (2006) Introduction to the physics and of Remote sensing (Wiley series in Remote sensing and Image Processing). Wiley-Inter science, p 18
- <span id="page-30-12"></span>Elkhrachy I (2015) Land use change detection using satellite images for Najran City, Kingdom of Saudi Arabia (KSA), WCS-CE-The World Cadastre Summit, Congress & Exhibition Istanbul, Turkey, 20–25 April
- <span id="page-30-15"></span>Fadhil AM (2009) Land degradation detection using geo-information technology for some sites in Iraq. J Al-Nahrain Univ Sci 12(3):94–108
- <span id="page-30-14"></span>Fadhil AM (2011) Drought mapping using Geoinformation technology for some sites in the Iraqi Kurdistan Region. Int J Digit Earth 4(3):239–257
- <span id="page-30-16"></span>Fadhil AM (2013) Sand dunes monitoring using remote sensing and GIS techniques for some sites in Iraq. In: Proceedings SPIE 8762, PIAGENG 2013: intelligent information, control, and [communication technology for agricultural engineering, p 876206.](http://dx.doi.org/10.1117/12.2019735) http://dx.doi.org/10.1117/12. 2019735
- <span id="page-30-19"></span>Faleh A, Shawan J (2012) GIS and remote sensing: principles and applications. Anvo Brandt Press. Fas. The Kingdom of Morocco
- <span id="page-30-6"></span>Hamad AI (2009) The use of remote sensing and geographic information systems of land evaluation in the center of the Iraqi alluvial plain.MSc. Thesis, Faculty of Agriculture, University of Baghdad, Iraq
- <span id="page-30-4"></span>Hassan AH (1983) Hydrological, hydrogeological and hydro chemical investigation of Bahar AL-[Najaf area. MSc. Thesis, College of Science, University of Baghdad, Iraq, p \(91\).](http://www.redalyc.org/articulo.oa?id=57320201) http://www. redalyc.org/articulo.oa?id=57320201
- <span id="page-30-21"></span>Jensen JR (2007) Introductory to digital image processing: a remote sensing perspective. Prentice Hall Series in Geographic Information Science
- <span id="page-30-8"></span>Katie HH, Mahimid AS (2002) The relationship between the geo morphological units and the distribution of soil units in the area west of Al-Razazah lake. Iraqi J Soil Sci 1(2):129–140
- <span id="page-30-22"></span>Koutroumpas A, Alexiou I, Vlychou M, Sakkas L (2010) Comparison between clinical and ultrasonographic assessment in patients with erosive osteoarthritis of the hands. Clin Rheumatol 29:511–516
- <span id="page-30-20"></span>Lillesand TM, Kiefer RW (2000) Remote Sensing and Image Interpretation, 4th edn. Wiley, New York, USA
- <span id="page-30-17"></span>Mahmoud RA, Mahimid AS (2011) Inheritance and evolution of some gypsum soils in Iraq. Iraqi J Desert Res 24(5):88–99
- <span id="page-30-18"></span>Martinez-Rios JJ, Monger HC (2002) Soil classification in arid lands with thematic mapper data, TERRA, 20(2):89–100. Available in: <http://www.redalyc.org/articulo.oa?id=57320201>
- <span id="page-30-11"></span>Mukherjee S, Mukherjee P (2009) Assessment and comparison of classification techniques for forest inventory estimation: A case study using IRS-ID imagery. Int J Geo Info 5(2):63–73
- <span id="page-30-7"></span>Mustafa BM, Al-Quraishi AMF, Gholizadeh A, Saberioon M (2019) Proximal soil sensing for soil monitoring. In: Al-Quraishi AMF, Negm AM (eds) Environmental remote sensing and GIS in Iraq. Springer Water
- <span id="page-30-1"></span>Mutter AN (2008) Efficiency methods calculating plain soil mapping units of soil mid mesopotamian plain. MSc. Thesis, College of Agriculture, University of Baghdad, Iraq
- <span id="page-30-13"></span>Nain P, Kumar K (2016) Study for accuracy assessment of land use and land cover classification of New-Delhi, North India. Int J Comput Sci Trends Technol (I JCS T) 4(3):137–143
- <span id="page-31-18"></span>Nash N (2016) Detection and accuracy assessment of mountain pine beetle infestations using landsat 8 OLI and WorldView02 satellite imagery Lake Tahoe Basin-Nevada and California. Msc. Thesis, Geographic Information Science and Technology, Faculty of the USC Graduate School, University of Southern California, USA
- <span id="page-31-17"></span>Proklamasi SA, Indonesia G, Myint M (2015) Conducting initial vegetation classification through image analysis. In: Chapter 3, Version 1.0 March, Natural Resources Information Integration; Ihwan Rafina, TFT; and Tri A. Sugiyanto, PT SMART/TFT
- <span id="page-31-10"></span>Qadir MH (2007) Study of land cover—land use and its reflectivity in Shahrazur plain by using remote sensing techniques. MSc. Thesis, Collage of Agriculture, University of Sulaimania, Iraq
- <span id="page-31-3"></span>Sallom AJ, Segar RH (1994) Semi-detail of soil survey and hydrological investigations in the Najaf Sea Project. Department of Investigation and Soil, Ministry of Water Resources
- <span id="page-31-8"></span>Shuli MA (2008) Study of land coverings in the Nablus area using remote sensing technology. An-Najah National University in Nablus, Palestine
- <span id="page-31-0"></span>Soil Survey Staff (1999) Soil Taxonomy, A basic system of soil classification for making and interpreting soil survey, 2nd edn. Agriculture Handbook No. 436, USDA
- <span id="page-31-16"></span>Stehman SV (2013) Estimating area from an accuracy assessment error matrix. Remote Sens Environ 132:202–211
- <span id="page-31-11"></span>Sulaiman AA, Abboud NF (2012) Classification and evaluation some of alluvial Soils in the center of the sedimentary plain. Tikrit Univ J Agric Sci 12(3):155–162
- <span id="page-31-15"></span>Taruvinga K (2008) Gully mapping using remote sensing: case study in KwaZulu-Natal, South Africa. M.Sc. thesis, Geography Science, University of Waterloo, Ontario, Canada
- <span id="page-31-9"></span>Tso B, Olsen R (2005) Combining spectral and special information into hidden Marcov odels for unsupervised image classification. Int J Remote Sens 26:211–2133
- <span id="page-31-5"></span>Wahib QA (2012) Characteristics of spectral reflectivity of surface soil and ground coverings of Al-Salman depression in southern Iraq. J Iraqi Agric Sci 43(4)(special number):129–140
- <span id="page-31-13"></span>Wu W, Zucca C, Muhaimeed AS, Al-Shafie WM, Al-Quraishi AMF, Nangia V, Zhu M, Liu G (2018) Soil salinity prediction and mapping by machine learning regression in central mesopotamia, Iraq. Land Degrad Dev 29(11):4005–4014
- <span id="page-31-12"></span>Wu W, Muhaimeed AS, Al-Shafie AM, Al-Quraishi AMF (2019) Using radar and optical data for soil salinity modeling and mapping in central Iraq. In: Al-Quraishi AMF, Negm AM (eds) Environmental remote sensing and GIS in Iraq. Springer Water
- <span id="page-31-4"></span>Yacoub SY, Purser BH, Al-Hassani NH, Al-Azzawi M, Orzag-Sperber F, Hassan KM, Plaziat JC, Younis WR (1981) Preliminary study of the Quaternary Sediments of SE Iraq. Joint project between the geological survey of Iraq and University of Paris XI, Orsay, GEOSURV, int. rep. no. 1078
- <span id="page-31-14"></span>Yacoub SY, Abdul Jabbar MF, Shehab AT (2008) Study of determination of ferrous rocks explored using remote sensing techniques in Al-Kaara Area. Western Iraq, J Iraqi Geol Mining 4(1):43–52
- <span id="page-31-6"></span>Yao Y, Qin Q, Fadhil AM, Li Y, Zhao S, Liu S, Sui X, Dong H (2011) Evaluation of EDI derived from the exponential evapotranspiration model for monitoring China's surface drought. Environ Earth Sci 63(2):425–436
- <span id="page-31-1"></span>Yousif BF (2004) The use of remote sensing techniques in the classification of Al-Najaf Soil. M.Sc. Thesis, Building and Construction Engineering, University of Technology, Baghdad, Iraq
- <span id="page-31-7"></span>Zhang Z, Kang H, Yao Y, Fadhil AM, Zhang Y, Jia K (2017) Spatial and decadal variations in satellite-based terrestrial evapotranspiration and drought over Inner Mongolia Autonomous Region of China during 1982–2009. J Earth Syst Sci 126(8)
- <span id="page-31-2"></span>Zinck J (2008) Remote Sensing of soil salinization: impact on Land management, CRC Press, Technology & Engineering, pp 374