Assessment of Soil Suitability for Improvement of Soil Factors and Agricultural Management

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Abstract: The dramatic growth of the world’s population is increasing the pressure on natural resources, particularly on soil systems. At the same time, inappropriate agricultural practices are causing widespread soil degradation. Improved management of soil resources and identification of the potential agricultural capability of soils is therefore needed to prevent further land degradation, particularly in dryland areas such as Egypt. Here, we present a case study in the El-Fayoum depression (Northern Egypt) to model and map soil suitability for 12 typical Mediterranean crops. Two management scenarios were analyzed: the current situation (CS) and an optimal scenario (OS) of soil variables. The Almagra model was applied to estimate soil suitability under CS and OS. Management options based on the CS assessment were proposed to reduce some limiting factors: a fixed value of 2 dSm\(^{-1}\) for soil salinity and 5% for sodium saturation; these defined the OS. Under optimal management, the OS scenario showed potential, where a notable increase of the area covered by a high suitability class (around 80%) for annual and semi-annual crops was observed. There was also a marked increase (about 70% for CS and 50% for OS) for perennial crops shifting from the marginal to moderate soil suitability class. The results reveal the importance of proper management to massively alter soil suitability into better states in order to achieve sustainable land use in this fertile agro-ecosystem.

Keywords: sustainable agriculture; MicroLEIS DSS; land-use planning; soil reclamation
1. Introduction

The dramatic increase of the world’s population is inducing enormous pressure on natural resources [1,2]. This pressure causes multiple environmental problems for land and water systems [3], and therefore, appropriate land-use and management strategies are needed to reduce the magnitude of these human impacts [4,5]. Agricultural activities have a direct effect on soils’ physical, chemical, and biological properties [6], resulting in environmental problems such as soil degradation [7], waterlogging [8], salinization/alkalization [9], and contamination [10,11]. These environmental problems affect soil quality and crop productivity, reducing food production capacity and food security.

Water scarcity is also a serious constraint for agriculture in drylands, where land is highly vulnerable to land degradation due to aridity [9]. At the global scale, Africa is highly affected by desertification, with over 45% of the land exposed to this process [12]. This issue calls for the urgent improvement of agricultural practices and water use efficiency [13] in order to reduce environmental problems and prevent further degradation [14]. Irrigation is often used to alleviate water scarcity, and often inappropriate irrigation practices induce soil salinization and soil sodicity, which are major causes of land degradation [15]. The excess of salts may cause clay dispersion and create soil crusts, decreasing permeability and soil productivity, but suitable agricultural land management may also improve conditions [16]. Although drainage problems occur naturally, agricultural land management can exacerbate it [16]. Improving soils’ physical properties facilitates sodium leaching, and decreases salinity and sodicity effects in soils [17].

Agriculture is the main economic activity in Egypt and supports the livelihoods of approximately 55% of the population, contributing to around 20% of foreign exchange earnings, and approximately 30% of Egypt’s commodity exports [18]. Agricultural land in Egypt represents approximately 3.8 Mha and the main crops cultivated are wheat, cotton, maize, sunflower, clovers, tomatoes, aromatic and medicinal plants, mangoes, olives, and citrus [18]. In the largely fertile Nile valley, soil productivity is restricted by salinity as result of irrigation and by urban sprawl over productive soils. Such unsustainable management results in land degradation, with implications for soil productivity, food production, and food security [19–21].

Land evaluation involves determination of the land potential for agricultural purposes [22,23] and its main objective is to manage and improve land in a sustainable way to increase its potential for human uses [22,24]. Land suitability status is based on intrinsic properties of soils (e.g., parent materials, soil texture and depth) and characteristics that can be altered by human management (e.g., drainage, salinity, nutrient concentration and vegetation cover) [25,26]. Several land evaluation guidelines have been created in the last decades (e.g., [22,25–31]).

The evolution of technology and models capable of analyzing a large number of variables has increased the sophistication of land suitability analyses [32,33]. Furthermore, the development of geographical information systems (GIS) and geostatistical techniques has allowed improved spatial processing of information on the variables that affect land degradation and land suitability, which are important for sustainable use of agricultural areas [6]. Soil and land evaluation models, hypothetical scenarios of agriculture management, and spatial analyses are valuable tools for land managers and decision makers to achieve sustainable land-use planning and management for targeted areas [34–36]. The Micro Land Evaluation Information System has been extensively used worldwide for land suitability assessment (MicroLEIS, [29,30,37]). The MicroLEIS focuses on an integrated system of soil, climate, and agriculture management databases for land evaluation and contains two sets of models related to land suitability and land vulnerability. These models have been widely used for: (i) land degradation evaluation and prediction of optimal land-use and management practices, and (ii) assisting decision makers in solving agro-ecological problems (e.g., [36,38–44]). The Almagra model is one of the main components of the MicroLEIS DSS that was designed for land suitability evaluation [45].

Given the importance of the agricultural sector in Egypt, there is a strong need for assessments of the agricultural potential of the existing soils. This is especially important in an area extremely
vulnerable to land degradation and where soil productivity is important for food security. In this study, the Almagra model was used to evaluate soil suitability in the El-Fayoum depression (Northern Egypt). The main objectives were to: (i) assess soil suitability in relation to cultivation of 12 Mediterranean crops (annual, semi-annual, and perennial) under current scenario (CS) and optimal scenarios (OS) of soil management, and (ii) to obtain the spatial distribution of soil suitability under the proposed scenarios based on physiographic units.

2. Material and Methods

2.1. Study Area

The El-Fayoum depression is located within the El-Fayoum region (western desert of Egypt) at coordinates 29°02′–29°35′ N and 30°23′–31°05′ E (Figure 1). Monthly mean temperature and precipitation inside and outside of the depression are shown in Figure S1.

![Figure 1. Location of the El-Fayoum depression within Egypt.](image)

The El-Fayoum altitude ranges between −52 to 141 m in relation to mean sea level and has a flat topography (Figure 2A, B). According to Reference [46], the soils of the El-Fayoum depression belong to the Aridisol and Entisol soil groups. The dominant soil subgroup is the Vertic Torrifluvent that covers an area of 43% of the study area; the other subgroups are Typic Haplocalcids, Typic Torrifluvents, Typic Haplogypsids, Typic Haplosalids, and Typic Torripsamments [47]. The depression is linked to the Nile River by the canal of Hawara that transports the water to the depression. The study area involves six districts (Tamia, Sinnoris, Ibshawai, Fayoum, Yousef El Sadik, and Itsa). Agriculture is the main activity in the study area and the total cultivated land is 177,802 ha [18].

2.2. Soil and Climate Databases

Twenty-nine soil profiles were selected from the studies of [48,49] to analyze the soil characteristics in the current situation (CS). Each of these profiles was the average of three individual points (Figure 2C). For this study, we selected the most important soil variables (model input) to evaluate soil quality, e.g., electrical conductivity (EC), exchangeable sodium percentage (ESP%), pH, carbonate content (CaCO$_3$),
Agroclimatic parameters such as evapotranspiration and aridity index (ARi, numbers of arid months in which the actual precipitation was lower than the evapotranspiration) were calculated from El-Fayoum weather station data from the period 1962–2006 [50]. Water supply from the Nile for irrigation (2.64 × 10^9 m^3/year) was also considered in the models. The monthly distribution of these parameters is shown in Figure S1.

Figure 2. Study area: (A) elevation, (B) slope, (C) sampled sites and physiographic units, and (D) land use.

2.3. Soil Suitability Evaluation

Almagra Model

This model is considered as a qualitative biophysical evaluation model that uses the diagnostic criteria of soil variables and the favorable conditions for crop growth within MicroLEIS ([30,45,51], Figure S2). This model is more practical to use than other land suitability models [52]. The Almagra model has been used and calibrated in previous studies in Mediterranean regions [37,41,53–55]. In this study, the Almagra model defined soil suitability within five different classes (Figure S3): optimum (S1), high (S2), moderate (S3), marginal (S4), and not suitable (S5) for twelve traditional crops that included annuals, e.g., wheat, sunflower, sugar beet, melon, potato, soybean, cotton, and corn; semi-annuals, e.g., alfalfa; and perennials, e.g., olive, citrus, and peach. The level of generalization of the variables was dependent on the crop requirement for each soil variable using the “most limiting factor” approach to identify the soil suitability classes. Here, the Almagra model was applied to evaluate the CS of soil...
suitability for the twelve Mediterranean crops considering all the soil limiting factors in the El-Fayoum depression. The optimal scenario was based on the soil variables that could be managed, such as EC, ESP%, and drainage, without taking into account the interaction between them. Soil depth and texture were not considered as these variables are not easily modified. The soil suitability classes were further divided into 18 subclasses depending on the number of limiting factors in each soil suitability class (Figure S3).

The suggested optimal scenario (OS) was determined according to the following equation:

\[
\text{OS} = \text{CS} - \text{URs}
\]

where; OS: optimal scenario; CS: current situation; URs: units of reduction.

The units of reduction were defined based on the CS assessment to meet the proposed fixed value of the OS in order to increase final soil suitability to S2 (high soil suitability). The proposed fixed values of each manageable soil factor are highlighted in green boxes in Figure S2. Accordingly, when soil suitability under CS is S5 or S4 (marginal and not suitable, respectively), these will require higher URs compared to S3 (moderate), which need lower reduction units to meet the fixed value of OS for each soil variable. For soil salinity, the OS was proposed to decrease ES classes (ranged from slightly saline to high saline) to a fixed value of 2 dSm\(^{-1}\) (non-saline class). In terms of ESP%, the intended value of the OS was 5%. Finally, as the drainage factor is a qualitative parameter in the Almagra model, the OS was aimed to enhance the drainage from “very poor”, “poor”, and “moderate”, to “good” drainage status.

2.4. Spatial Analysis

A Landsat 8 satellite image (path/raw: 177/40 acquired in July 2016) was used as a base map for the spatial analysis of the studied variables. The image was masked to the study area based on the administrative boundaries of the El-Fayoum depression using ArcGIS 10.4 [56].

The Shuttle Radar Topographic Mission (SRTM) digital elevation model (DEM) produced by NASA (version 3, 30 m resolution, updated on 6 August 2015) was downloaded from the US Geological Survey website using EarthExplorer. The DEM of the El-Fayoum depression was subset to the study area to accurately allocate the physiographic units [49,57]. The slope percentage was calculated based on the DEM using the spatial analysis tools in ArcMap and the slope classes were generated based on the Soil Survey Manual [58,59].

In order to obtain more reliable results, the largest extended urban areas (>10 hectare) were excluded from the total area coverage. A supervised classification procedure by maximum likelihood classifier was used to separate the urban areas from other land cover classes. All area coverages (km\(^2\) and %) for the physiographic units and urban areas were calculated in the attribute table using ArcMap.

To display the spatial distribution of soil salinity and sodium saturation percentage of the soil profiles under CS and URs, these variables were interpolated using ArcMap. This interpolation considered the values of the soil input factor as well as the DEM using Co-Kriging. Additionally, inverse distance weighting (IDW) was executed to spatially represent the distribution of soil suitability for the twelve studied crops under CS and OS, as well as the total average of soil suitability.

2.5. Statistical Analysis for Soil Factors

All soil variables analyzed were tested for normality using the Shapiro–Wilk normality test. Data were log transformed as necessary for analysis, but all presented data were non-transformed for ease of interpretation. Differences among physiographical units for each soil property were tested using a one-way ANOVA test. The SPSS 23 software [60] was used at a 95% confidence interval (significance level of \(\alpha = 0.05\)). The main descriptive statistical parameters (maximum, minimum, and mean) were also generated.
3. Results and Discussion

3.1. Spatial Analysis and Soil Physiochemical Properties

3.1.1. Physiographic Units

Eight physiographic units were distinguished in the area besides the hilly and rocky areas (Figure 2C, Table 1). When considering the entire study area (with the urban area excluded), the fan physiographic unit represented one fifth of the study area (20%). The transition elevated plains (fluvial/lacustrine) covered an area of 30% and around the same area percentage (30%) was allocated to lake terraces (oldest, old, and recent). The uncultivated areas: rocky area (RU), and hilly area (HU) covering 105 and 22 km$^2$ respectively, were excluded from the geometric area calculation for the evaluation process. The total urban area obtained was 115 km$^2$ (Figure 2D) which represents around 6% of the area coverage.

Table 1. The main physiographic units (PU), soil taxonomy according to the United States Department of Agriculture (USDA, 2014) and area coverage by km$^2$ and % for the studied area. VTFE, Vertic Torrifluvents; TTFE, Typic Torrifluvents; TTPE, Typic Torripsamments; THCI, Typic Haplocalcids; THGI, Typic Haplogypsids; THIS, Typic Haplosalids.

| Unit                  | Taxonomy (USDA, 2014) | Area          |
|-----------------------|------------------------|---------------|
|                       |                        | Urban Included | Urban Excluded |
|                       |                        | km$^2$        | %             | km$^2$         | %             |
| PU1-Flood plain       | VTFE                   | 59.47         | 3.33          | 55.5           | 3.32          |
| PU2-Fan               | TTFE and VTFE          | 374.51        | 20.98         | 336.04         | 20.13         |
| PU3-Basin             | THCI and VTFE          | 295.16        | 16.54         | 283.43         | 16.98         |
| PU4-Transition A      | THCI, THGL, and VTFE   | 347.5         | 19.47         | 325.87         | 19.52         |
| PU5-Transition B      | VTFE and THCI          | 182.81        | 10.24         | 166.81         | 9.99          |
| PU6-Oldest lake terraces | TTFE and THCI        | 234.32        | 13.13         | 226.47         | 13.57         |
| PU7-Old lake terraces | TTFE and THCI          | 203.03        | 11.38         | 192.85         | 11.55         |
| PU8-Recent lake terraces | THIS                 | 87.87         | 4.92          | 82.58          | 4.95          |
| Total                 |                        | 1784.67       | 100           | 1669.54        | 100           |

3.1.2. Soil Properties

Almost all physiographic units were represented by four profiles except the flood plain unit (PU1), which was represented by only one soil profile, and the fan unit (PU2), which was represented by five. Table 2 illustrates the mean values of the soil factors in each physiographic unit. The elevated mean value of the EC was assigned to the recent lake terraces unit (PU8), adjacent to Qarun Lake, with a value of 31.9 dSm$^{-1}$. The elevated EC in the soil close to Qarun lake could be caused by the percolation of the saline lake water, which can eventually lead to a rising water table within the soil profile depth [14]. Generally, the soils’ pH ranged from 7.54 to 8.87 (from slightly to strongly alkaline). The highest mean value of CaCO$_3$ content in the El-Fayoum soils was 19.71%, which was observed in the elevated plain (Lacustrine/Fluvial) unit (PU5), while the lowest mean value (3.96%) was observed in the PU1-Flood plain. Several soil texture classes were found in the study area, from sand (>90% sand) to clay (around 65% clay). The heavy texture (clay content of 45%-60%) was primarily found in the PU2-fan physiographic unit. The mean values of exchangeable sodium percentage ranged from 8.72 to 19.83% based on the type of the physiographic unit, while the highest CEC value occurred within the fan physiographic unit with highest amount of clay. Generally, the El-Fayoum depression soils were poor in organic matter, with around 80% of the studied area having less than 1.5%. Deeper profiles (120–150 cm) were observed in the flood plain, fan, and basin physiographic units. Sand, silt, and clay contents differed significantly ($p < 0.05$) among physiographic units, but the rest of the variables (e.g., EC, pH, CaCO$_3$, OM, and ESP) were not significantly different.
Table 2. The mean values of the soil factors that were used in the suitability model for each physiographic unit. PU1, Flood plain; PU2, Fan; PU3, Basin; PU4, Transition elevated plain A; PU5, Transition elevated plain B; PU6, Oldest lake terraces; PU7, Old lake terraces; PU8, Recent lake terraces; EC, electric conductivity; OM, organic matter; ESP, exchangeable sodium percentage; CEC, cation exchange capacity.

| Physiographic Unit | Particle Size Distribution, % | OM, % | EC, dS/m | pH | CaCO₃, % | ESP, % | CEC, meq/100 g |
|--------------------|-------------------------------|-------|----------|-----|----------|--------|--------------|
| PU1                | 43.97 24.17 31.86             | 2.05  | 2.56     | 7.55| 3.69     | 8.72   | 24.96        |
| PU2                | 19.13 28.74 52.13             | 1.59  | 5.79     | 8.21| 4.9      | 14.55  | 41.16        |
| PU3                | 57.54 19.85 22.61             | 1.54  | 5.92     | 7.94| 13.45    | 12.03  | 19.8         |
| PU4                | 32.25 34.92 32.83             | 1.28  | 2.86     | 8.07| 14.81    | 17     | 26.92        |
| PU5                | 43.74 21.8 34.45              | 1.21  | 10.86    | 8.11| 19.71    | 10.1   | 26.02        |
| PU6                | 77.66 8.73 13.61              | 0.72  | 2.43     | 7.72| 7.18     | 8.87   | 13.34        |
| PU7                | 35.95 26.33 37.72             | 1.17  | 8.98     | 8.08| 18.33    | 14.1   | 35.75        |
| PU8                | 42.6 16.68 40.73              | 2.2   | 31.89    | 8.41| 11.38    | 19.83  | 30           |

3.2. Soil Factors under Current and Optimal Scenarios

3.2.1. Soil Salinity

Around 70% of the study area had a slight-to-moderate soil salinity with EC ranging between 2 and 6 dSm⁻¹ under CS (Table 3). Approximately 5% of the study area showed extreme salinity (>16 dSm⁻¹) (Figure 3a).
Table 3. Area (km$^2$ and %) of each soil salinity class and sodium saturation in the study area.

| Soil Salinity (dSm$^{-1}$) $^a$ | Area (km$^2$) | Sodium Saturation (%) | Area (km$^2$) |
|--------------------------------|--------------|------------------------|--------------|
| 1.65–2                        | 0.79         | 5.36–7                 | 1.27         |
| 2–4                           | 586.37       | 7–10                   | 282.72       |
| 4–6                           | 592.08       | 10–12                  | 326.89       |
| 6–8                           | 213.74       | 12–15                  | 658.23       |
| 8–10                          | 84.68        | 15–17                  | 215.60       |
| 10–16                         | 106.46       | 17–20                  | 113.84       |
| >16                           | 85.42        | 20–22                  | 41.88        |
| Total                         | 1669.54      | 22–25                  | 29.09        |
|                               |              | >25                    | 0.01         |
|                               |              | Total                  | 1669.54      |

$^a$ 0–2: non-saline; 2–4: slightly saline; 4–8: moderately saline; 8–16: highly saline; >16: very highly saline.

Under the optimal scenario, all salinity classes were decreased to 2 dSm$^{-1}$ (non-saline soil) with more than two-thirds of the study area assigned as slight-to-moderately saline soil (Table 3). The maximum soil salinity values (>16 dSm$^{-1}$) were observed in PU8 under CS and OS (Figure 4A,B). On the other hand, the non-saline soil (<2 dSm$^{-1}$) resulted in no reduction and the highest salinity classes (10–16 dSm$^{-1}$) in 8 to 14 units of reduction (Figure 4A,B).

Figure 4. Spatial distribution of soil salinity under the current situation (CS) and the projected reduction units in the optimal scenario (OS) scenario (A,B); and the spatial distribution of sodium saturation under the current situation and the projected reduction in the OS scenario units (C,D).
Different soil management options have been suggested to reduce soil salinity, such as removal of salts from the soil root zone through a leaching process via low salinity water [61]. However, the growth rates of plants under salt stress often vary between cultivars and strongly differ between plant species [62,63].

3.2.2. Sodium Saturation

In around 39% of the study area, sodium saturation ranged from 12 to 15% under CS (Table 3). Low-sodium saturation (5–12%) represented around 37% of the study area, and high-sodium saturation (20–25%) covered <5% of the area (Figure 3b). Therefore, 37% of the study area can be improved to the intended value (5%) with 5–7 reduction units, and 7–10 reduction units are needed to enhance 39% of the study area. The spatial distribution of sodium saturation under CS and the projected reduction units of sodium saturation for each class are shown in (Figure 4C,D).

The increased sodium saturation and salinization in arid and semiarid regions negatively affect soils’ physical properties such as soil structure and hydraulic conductivity, and consequently adversely impact crop yield, and if the exchangeable sodium percentage passes 15, these soils are classified as sodic soils [64,65]. On the other hand, high-sodium saturation could be improved by the addition of gypsum [16,17,66]. The addition of gypsum leads to the replacement of sodium with calcium on soil particles, which directly affects soil aggregation and reduces pH [17,67]. Thus, the incorporation of calcium using gypsum has proved to enhance soil particle aggregation, thus creating an adequate soil physical condition for nutrient uptake in sodic soils [68].

3.2.3. Drainage

Agriculture drainage in the studied area ranged from excessive (PU6) to very poor (PU8) (Figure 3c), and the dominant drainage status was the poor class under CS. Adequate drainage is essential to discharge leached salts from locations that have suffered from salinity problems [69,70]. In El-Fayoum, the main source of irrigation is Bahr Youssif (an old branch of the River Nile) with a water salinity value < 0.4 dSm$^{-1}$. An optimum drainage status may be achieved by the addition of OM and gypsum, that would directly enhance soils’ physiochemical properties and consequently improves soil suitability [71,72]. Additionally, reducing tillage in sandy soils [73] and adding sand in the cases of clayey soil with very poor drainage [74,75], could lead to increased agriculture suitability under OS. An improved irrigation and drainage system could additionally help to recover and enhance soil properties, although over-irrigation could lead to decreasing soil suitability [76].

3.3. Soil Suitability

3.3.1. General Evaluation of Current Situation (CS) and Optimal Scenario (OS)

Table 4 shows how the Almagra model classified the three soil profile examples, and also illustrates the improvement in suitability classifications under the OS compared with the CS. Generally, the non-suitable class (S5) occurred mainly in the physiographic unit PU8 (Figure 5). For the perennial crops (citrus, peach, and olive), the main area assigned was the marginal suitability class (S4), and for the rest of the studied crops, the dominant classes were moderate (S3) and marginal (S4). High salinity, elevated sodium saturation, poor drainage, and heavy texture were observed as limiting factors in the physiographic units PU4 and PU5.

According to the spatial distribution (Figure 5) under CS, the average of soil suitability for each crop was classified as: cotton > sugar beet > wheat, melon, potato, soybean sunflower, and alfalfa > corn > olive > citrus, and peach. Soil suitability assessments for annual and semi-annual crops under CS were mostly classified as moderate in the adjacent south-western areas of Qarun Lake to not suitable for almost all crops in the south-eastern soils (Figure 5).
Table 4. Soil suitability classifications according to the Almagra model. Each soil profile was classified on a scale from 1 (best) to 5 (worst) for each subclass, based on the specific requirements for each crop. The final classification was determined by the worst subclasses (in red), which was indicated by their letter. The crops were: I, wheat; II, corn; III, melon; IV, potato; V, soybean; VI, cotton; VII, sunflower; VIII, sugar beet; IX, alfalfa; X, peach; XI, citrus fruits; XII, olive. CS, Current Situation; OS, Optimal Scenario.

| Profile Code | Soil Factors/Classification | Crops |
|--------------|-----------------------------|-------|
| F10          | Useful depth (p)             |       |
|              |                              | I     | II    | III   | IV    | V     | VI    | VII   | VIII  | IX    | X     | XI    | XII   |
|              | Texture (t)                  | 1     | 1     | 1     | 1     | 1     | 1     | 1     | 1     | 1     | 1     | 1     | 1     |
|              | Drainage (d)                 | 3     | 2     | 2     | 2     | 3     | 2     | 2     | 3     | 3     | 4     | 4     | 4     |
|              | Carbonate (c)                | 2     | 1     | 1     | 1     | 2     | 1     | 2     | 2     | 2     | 1     | 1     | 2     |
|              | Salinity (s)                 | 1     | 1     | 2     | 2     | 1     | 2     | 1     | 2     | 2     | 2     | 2     | 2     |
|              | Sodium sat (a)               | 2     | 2     | 2     | 2     | 2     | 1     | 2     | 2     | 2     | 2     | 2     | 2     |
|              | Profile dev (g)              | 1     | 1     | 1     | 1     | 1     | 1     | 1     | 1     | 1     | 2     | 2     | 1     |
|              | CS classification            | S3d   | S2da  | S2tdsa| S2tdsa| S3d   | S2tda| S2dcsa| S3d   | S3d   | S4d   | S4d   | S4d   |
|              | OS classification            | S2c   | S1    | S2t   | S2t   | S2c   | S1    | S2c   | S2t   | S2c   | S2t   | S2t   | S2t   |
| F22          | Useful depth (p)             | 1     | 1     | 1     | 1     | 1     | 1     | 1     | 1     | 1     | 1     | 1     | 1     |
|              | Texture (t)                  | 1     | 1     | 2     | 2     | 1     | 2     | 1     | 1     | 1     | 1     | 1     | 2     |
|              | Drainage (d)                 | 1     | 1     | 1     | 1     | 1     | 1     | 1     | 1     | 1     | 1     | 2     | 2     |
|              | Carbonate (c)                | 2     | 1     | 1     | 1     | 2     | 1     | 2     | 2     | 2     | 2     | 1     | 1     |
|              | Salinity (s)                 | 4     | 4     | 4     | 4     | 3     | 4     | 3     | 4     | 3     | 5     | 5     | 3     |
|              | Sodium sat (a)               | 2     | 3     | 2     | 2     | 2     | 1     | 2     | 1     | 2     | 2     | 2     | 2     |
|              | Profile dev (g)              | 1     | 1     | 1     | 1     | 1     | 1     | 1     | 1     | 1     | 2     | 2     | 1     |
|              | CS classification            | S4s   | S4s   | S4s   | S4s   | S4s   | S4s   | S4s   | S4s   | S4s   | S4s   | S4s   | S4s   |
|              | OS classification            | S2c   | S1    | S2t   | S2t   | S2c   | S2c   | S1    | S2c   | S2t   | S2c   | S2t   | S2t   |
| F26          | Useful depth (p)             | 1     | 1     | 1     | 2     | 2     | 2     | 2     | 2     | 2     | 2     | 2     | 1     |
|              | Texture (t)                  | 2     | 2     | 2     | 2     | 2     | 2     | 2     | 2     | 2     | 4     | 4     | 4     |
|              | Drainage (d)                 | 4     | 3     | 3     | 4     | 3     | 4     | 3     | 4     | 4     | 5     | 5     | 5     |
|              | Carbonate (c)                | 1     | 2     | 2     | 2     | 1     | 2     | 1     | 1     | 1     | 2     | 2     | 1     |
|              | Salinity (s)                 | 5     | 5     | 5     | 5     | 5     | 5     | 5     | 5     | 5     | 5     | 5     | 5     |
|              | Sodium sat (a)               | 3     | 4     | 3     | 3     | 3     | 3     | 3     | 3     | 3     | 3     | 3     | 3     |
|              | Profile dev (g)              | 2     | 2     | 2     | 2     | 2     | 2     | 2     | 2     | 2     | 2     | 2     | 2     |
|              | CS classification            | S5s   | S5s   | S5s   | S5s   | S5s   | S5s   | S5s   | S5s   | S5s   | S5s   | S5s   | S5s   |
|              | OS classification            | S5s   | S5s   | S5s   | S5s   | S5s   | S5s   | S5s   | S5s   | S5s   | S5s   | S5s   | S5s   |
Figure 5. Spatial distribution of the soil suitability classes under the current situation of soil factors, according to the application of the Almagra model in the El-Fayoum depression. The Main limiting factors; s, salinity; d, drainage; a, sodium saturation; t, texture; c, carbonate content; g, profile development.

While under the OS, the marginal and moderated classes changed to high suitability (Figures 5 and 6). Importantly, there was at least one factor limiting soil suitability and preventing classification as optimum (Table 4). Although for all annual and semi-annual crops, the dominating soil suitability class was S2 (high), where the moderate and marginal suitability classes were predominant under perennials crops due to the difficulty of managing or improving soil factors, such as soil texture, high carbonate content, and soil depth (Figure 6).
with the implementation of the OS, the S2 class improved to cover an area of 1287 km² (77%) (Table 5, Figures 4, 5, and 7). Cotton is one of the strategic economic crops in Egypt [79]. In the El-Fayoum depression, the soil suitability for cotton under CS ranged from S2, S3, and to S4, with a coverage of 29%, 58%, and 12%, respectively. Applying the OS enhanced the area coverage for S2 to represent 81% (1357 km²). The S3 and S4 classes for sunflower under CS represented 90% of the study area (70% and 21%, respectively), whereas with OS, the area coverage for S2 was improved to cover 1299 km² (78%) of the study area. For the sugar beet crop, 93% of the study area was allocated for the S3 and S4 classes (77% and 16%, respectively), while under the OS, the area of S2 increased to 1299 km² (78%). Finally, the S3 and S4 classes for alfalfa under CS covered 96% of the study area (79% and 17%, respectively); however, with OS, the area coverage for S2 was improved to cover 1299 km² (Table 5, Figures 4, 5, and 7).

Regarding subclasses, the subclass 4 represented the large area for almost all annual and semi-annual crops studied, except for cotton and sunflower the subclass 3 represented the maximum extension.

Figure 6. Spatial analysis of soil suitability classes under the projected improvement of soil factors, according to the application of the Almagra model in the El-Fayoum depression. The main limiting factors: t, texture; c, carbonate content; p, soil depth; g, profile development; s, salinity; d, drainage; a, sodium saturation.

3.3.2. Soil Suitability for Studied Crops

El-Fayoum soils are inherently very fertile, but with poor management and environmental conditions, the salinity levels and the ESP have increased in the area, which significantly affects the suitability of the studied crops [77,78]. Reducing the severity of the manageable soil limiting factors (EC, ESP, and drainage), where possible, resulted in an increase of soil suitability for all the studied crops under OS. Only areas with extreme salinity and shallow soil profile depths were excluded from the OS, as the modification of these factors would be unfeasible (Figure 7). Under CS, the main soil suitability subclasses were represented by subclasses 7 to 17, which covers the suitability classes of S3 and S4 for the most evaluated crops (Figure 4). With the application of the OS, the moderate and marginal suitability classes (S3 and S4, respectively) shifted to the high suitability class (S2) (Table 5).
with the implementation of the OS, the S2 class improved to cover an area of 1287 km² (1357 km² under CS). The subclass 3 represented the maximum extension. Subclasses 4 represented the large area for almost all annual and semi-annual crops. Under CS, the S3 and S4 soil suitability classes for alfalfa under CS covered 96% of the study area (79% and 17%, respectively); however, with the application of OS, the S2 class represented 1299 km², whereas with CS, 1291 km² (77% and 21%, respectively). Applying the OS enhanced the area coverage for S2 to represent 81% (1314 km²) of the study area. Under the CS, 88% of the soil area was classified as S3 and S4 classes and covered areas of 533 km² (32%) and 44% (739 km²), respectively. The S3 and S4 classes for sunflower under CS represented 90% of the study area (70% and 20%, respectively), whereas with OS, the S2 class represented 1299 km² (78%). Finally, the S3 and S4 classes for potato crop (1150 km² and 317 km², respectively) under CS. However, with the application of OS, the area of the S2 class increased substantially from 119 km² to 1350 km², which represented 81% of the study area. Under the CS, 88% of the study area was classified as S3 and S4 for potato crop (1150 km² and 317 km², respectively), which was enhanced under the OS to the S2 class covering an area of 1342 km². The S3 and S4 classes covered 90% of the study area (1175 km² and 431 km², respectively) for soybean under CS. However, with the implementation of the OS, the S2 class improved to cover an area of 1287 km² (77%) (Table 5, Figures 5, 6 and 8). Cotton is one of the strategic economic crops in Egypt [79]. In the El-Fayoum depression, the soil suitability for cotton under CS ranged from S2, S3, and to S4, with a coverage of 29%, 58%, and 12%, respectively. Applying the OS enhanced the area coverage for S2 to represent 81% (1357 km²). The S3 and S4 classes for sunflower under CS represented 90% of the study area (70% and 21%), whereas with OS, the S2 class represented 1299 km² (78%) of the study area. For the sugar beet crop, 93% of the study area was allocated for the S3 and S4 classes (77% and 16%, respectively), while under the OS, the area of S2 increased to 1299 km² (78%). Finally, the S3 and S4 classes for alfalfa under CS covered 96% of the study area (79% and 17%, respectively); however, with OS, the area coverage for S2 was improved to cover 1299 km² (Table 5, Figures 4, 5 and 7). Regarding subclasses, the subclass 4 represented the large area for almost all annual and semi-annual crops studied, except for cotton and sunflower the subclass 3 represented the maximum extension.
| Class/Scenarios | Crops            | Wheat | Corn  | Melon | Potato | Soybean | Cotton | Sunflower | Sugar Beet | Alfalfa | Peach | Citrus Fruits | Olive |
|-----------------|------------------|-------|-------|-------|--------|---------|--------|-----------|-----------|---------|-------|----------------|-------|
|                 |                  | A     | B     | A     | B      | A       | B      | A         | B         | A       | B     | A               | B     |
| S2-high         | 2.48             | 78.7  | 3.77  | 3.77  | 71.4   | 80.8    | 9.92   | 80.4      | 1.43      | 77.1    | 29.1  | 81.3            | 7.38  |
|                 | 71.1             | 15.8  | 52.8  | 16.8  | 68.9   | 13.8    | 14.1   | 70.4      | 17.2      | 57.6    | 13.4  | 69.7            | 16.6  |
| S3-moderate     | 24.1             | 4.8   | 39.8  | 4.81  | 21.7   | 4.72    | 19     | 4.78      | 25.8      | 5.01    | 11.9  | 4.66            | 20.7  |
|                 | 2.28             | 0.68  | 3.69  | 0.67  | 2.3    | 0.67    | 2.21   | 0.68      | 2.38      | 0.71    | 1.38  | 0.66            | 2.22  |
| S4-marginal     | 100              | 100   | 100   | 100   | 100    | 100     | 100    | 100       | 100       | 100     | 100   | 100             | 100   |
| S5-not suitable | 100              | 100   | 100   | 100   | 100    | 100     | 100    | 100       | 100       | 100     | 100   | 100             | 100   |
| Total           | 100              | 100   | 100   | 100   | 100    | 100     | 100    | 100       | 100       | 100     | 100   | 100             | 100   |
Figure 8. Soil suitability subclasses area (km$^2$) under CS and OS.

Three perennials crops were studied; peach, citrus, and olive. Under CS, around 91% of the study area was covered with S3 and S4 classes (21% and 71%, respectively) for peach and citrus. After implementing the OS, the area of S2 and S3 increased to around 28% and 50%, respectively. The S3 and S4 classes for olive under CS covered 95% of the study area (32% and 63%, respectively), which enhanced to S2 and S3 classes and covered areas of 533 km$^2$ (32%) and 44% (739 km$^2$), respectively (Table 5, Figures 4, 5 and 7).

Under CS, the average of soil suitability for the 12 studied crops was assigned to moderate and marginal suitability classes representing 71% and 25% of the study area, respectively (Figure 9). At the subclass level, the maximum area coverage of 378 km$^2$ (23%) was occupied by subclass 10 (S3 class). Remarkably, with the application of the OS, a notable improvement was observed, as approximately 75% (1248 km$^2$) of the study area was highly suitable for all evaluated crops. Conversely, the marginal suitability class was assigned to <5% of the area. The subclass 5 (S2 class) had a maximum area of 622 km$^2$ (37%) compared to the rest of soil suitability subclasses.
Successful sustainable agricultural management programs rely on the best choice of suitable crops for specific land in specific environmental conditions. There is robust interaction between soil characteristics and crop requirements. Thus, when the soil qualities fail to meet the requirements of the crop (soil suitability assigned as marginal or not suitable), this can be amended in some cases by improving soil management [80]. Enhancing agricultural management practices via improving soil physiochemical characteristics helps to raise soil suitability for optimal production of maize, melon, and olive [81–83]. Improving soil suitability for potato, maize, and alfalfa needs substantial efforts of soil management to increase crop production [84]. Reference [85] found that the most limiting factors of soil suitability for wheat production were elevated soil salinity and alkalinity. Additionally, they applied a qualitative model for evaluating soil suitability to assist decision makers for sustainable agriculture planning and economic productivity of soil resources in Iran. While, in the north-western region of Libya, Reference [86] reported that the main limiting factors of soil suitability for wheat, maize, alfalfa, sunflower, soybean, potato, citrus, and olives crops were soil salinity, soil texture, alkaline pH, calcium carbonate, and soil depth. Their findings are consistent with this research’s observations.

Several scientists have utilized a computer/GIS-based model for mapping soil suitability [45,85,87–89]. The spatial analysis of soil suitability has a vital role in carrying out agriculture management processes [85,90]. Reference [91] used a soil evaluation model to estimate the land suitability of wheat, sugar beet, potato, and alfalfa in Iran. On the other hand, Reference [92] introduce a Bayesian network model, ALECA (Agroecological Land Evaluation for Coffea arabica L.), to evaluate land suitability for coffee production. Other scientists have utilized multi-criteria decision-making processes to provide accurate estimations of land suitability for different crops such as citrus [93], tobacco [94], and tea [95].

4. Conclusions

In this study, the application of the Almagra soil suitability model, a component of the MicroLEIS, allowed for predictions of agriculture soil suitability for twelve Mediterranean crops. The model identified the most suitable crops and the optimal spatial distribution under current and optimal management scenarios. The assessment of soil suitability under the current situation (CS) of soil factors helps to distinguish the most limiting soil factors. Likewise, applying the soil suitability model under the improvement of manageable limiting factors such as soil salinity, sodium saturation, and drainage assists in predicting the degree of improvement in soil suitability under the proposed optimal scenario (OS).
Consequently, the Almagra model assisted in identifying which of the studied crops were suitable to grow and where under the CS and the OS. The soil suitability classes varied spatially from optimum to non-suitable classes depending on the crop type and the possibility of soil factor improvements in case of the OS. Generally, under the CS of soil factors, the dominant soil suitability classes in the El-Fayoum depression were the moderate and marginal classes. On the other hand, the highly suitable class was dominant under the projected improvement of the manageable soil factors. Cotton was the most suitable current crop, while under optimal management, the suitability for all crops improved except for the perennial crops (i.e., peach, citrus fruits, and olive trees), where the most limiting factors for these crops are soil texture, depth, soil profile development, and carbonate content which are inflexible to modification. The total average of soil suitability for the twelve Mediterranean crops under CS were cotton > sugar beet > wheat, melon, potato, soybean sunflower, and alfalfa > corn > olive > citrus, and peach.

The assessment of soil suitability can help decision makers recognize the most limiting soil factors. Assessing the potential for improvement of manageable limiting factors such as soil salinity, sodium saturation, and drainage may assist in predicting the degree of improvement in soil suitability under the proposed OS. Mapping current soil suitability in this study and its improvement potential therefore provides valuable information to decision makers for appropriate land-use planning and sustainable development in the El-Fayoum depression.

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