



Article

Efficiency of Geospatial Technology and Multi-Criteria Decision Analysis for Groundwater Potential Mapping in a Semi-Arid Region

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Abstract: The increasing water demand in Egypt causes massive stress on groundwater resources. The high variability in the groundwater depth, aquifer properties, terrain characteristics, and shortage of rainfall make it necessary to identify the groundwater potentiality in semi-arid regions. This study used the possibilities of multi-criteria decision approaches (MCDA), geographical information system (GIS), and groundwater field data to delineate potential groundwater zones in the Tushka area, west of Lake Nasser, South Egypt. Furthermore, groundwater potentiality identification can help decision-makers better plan and manage the water resources in this promising area. Eight controlling factors were utilized to achieve the objective of the present work using multi-criteria decision analysis (MCDA) approaches, namely the analytical hierarchy process (AHP) and frequency ratio (FR) models. The controlling parameters were integrated with the geographic information system (GIS) to develop the zones of groundwater potentialities. The results revealed that high and moderate-potential zones cover approximately 61% and 52% of the total area in the AHP and FR models, respectively. A total of 44 groundwater production wells along with the well yield were collected and used to validate the models. The results were evaluated using the receiver operating characteristics (ROC) curve. The best-performing prediction rates achieved by AHP and FR were 83% and 81%, respectively. Finally, the obtained results indicated that the AHP model achieved better performance than the FR model.

Keywords: groundwater potential zone; the analytic hierarchy process; frequency ratio model; GIS; Lake Nasser



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1. Introduction

Water shortage is one of the most critical global issues, especially in semi-arid and arid areas. The availability of groundwater serves as a key component in land development and sustainability for domestic and agricultural purposes [1]. Therefore, and as a part of resource management and planning, the exploration of groundwater as a source of freshwater is essential. In Egypt, groundwater is the second main water source, after the River Nile. For decades, the Egyptian government has attempted to expand its fertile area by reclaiming desert land depending on River Nile channels or groundwater. Tushka is the largest reclamation project in Egypt, which aims to transform the dry land of Upper Egypt into a comprehensive agricultural zone as part of a repatriation effort to ease congestion in the Nile River Basin. The area west of Lake Nasser is a part of the Tushka project, in which groundwater is the main water resource. Without proper management and evaluation of groundwater potentiality, the groundwater exploitation in this area has already begun through drilled wells for a long time, resulting in a decrease in the groundwater yield of

Water 2022, 14, 882 2 of 18

some wells after a while. In this regard, the determination and assessment of groundwater potential areas can play a crucial role in increasing the long-term viability of projects, minimizing the risk of water shortage, and reducing drilling costs. Traditional groundwater exploration techniques, such as geophysical methods and borehole data, are costly and time-consuming. Alternatively, the integration of geographic information systems (GIS) coupled with remote-sensing data and multi-criteria decision-making analysis approaches (MCDA) have been widely used for mapping potential groundwater zones [2–5]. Among these, weighted overlay (influencing factor) is a common method implemented by many researchers [6-8]. AHP, one of the MCDA approaches assessing multiple factors that was first developed by [9], is required for rapid evaluation, reliable prediction, and future planning for groundwater. Many studies have been conducted to delineate groundwater potential maps using the AHP model worldwide [10–15]. In addition, the AHP method has been widely used in many different environmental studies [16,17]. The frequency ratio model, as a tool to delineate groundwater potential, is not widely used; instead, a few studies have been carried out using frequency ratios [18–22]. Recently, several approaches and methodologies employed for groundwater potential, including the weights of evidence model [23,24], logistic model tree [25,26], Dempster-Shafer model [27], evidential belief function [28], certainty factor [21,29], logistic regression [30], and random forest model [31], have been successfully implemented. Alternative approaches have been used to study water stress applying GIS indicators, geostatistical analysis, and the stochastic model [32,33].

All of these methodological techniques have made groundwater investigation more straightforward, cost-effective, and time-efficient. Although all these approaches provide fairly accurate results, it is always better to follow statistical approaches, such as AHP or FR, in potential groundwater zone mapping.

Groundwater potentiality anywhere depends on geological, structural, hydrological, and climatic factors [34]. In areas of scarce rainfall, such as the area west of Lake Nasser, choosing appropriate criteria controlling groundwater recharge and its potentiality is important, as hydrogeological and structural factors become more effective. The aquifer characteristics and the interaction between groundwater and Lake Nasser water have been investigated in this area [35–38].

The key objective of the current work is to delineate groundwater potential zones in the area west of Lake Nasser, Egypt, using AHP and FR models, as new approaches used in this area, and compare these models to evaluate the efficiency of the application of groundwater potentiality mapping. This objective is achieved by preparing and overlying thematic layers for the most critical parameters that control groundwater recharge and flow in the area that affect groundwater potentiality.

This study would help to improve water for irrigation planning and management at a comparatively low cost. The applied approaches are well-known multi-criteria decision analyses and, to our knowledge, they have never been investigated in the study site.

2. Study Area Description

The study area is a part of southern Egypt, located in the southeastern corner of the Egyptian Western Desert, covering an area of 14250 km^2 . It is occupied the area west of Lake Nasser, between latitudes $22^{\circ}10'-23^{\circ}40'$ N and longitudes $31^{\circ}10'-33^{\circ}00'$ E, and bounded to the east by the shoreline of the lake (Figure 1).

Water 2022, 14, 882 3 of 18

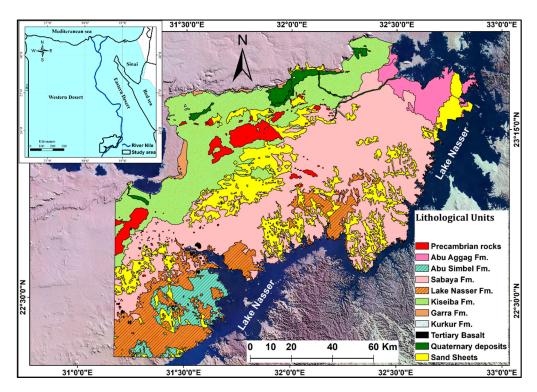


Figure 1. Location and geological map of the study area, modified from [39].

Geologically, the rock units exposed in the area are from the Precambrian basement to Cenozoic sedimentary successions (Figure 1). The area covered with the Precambrian basement rocks (Arabian Nubian Shield), including granites, granodiorite, gneiss, and Schist [40,41], is overlain by Mesozoic successions of limestones and the Cretaceous sandstones of Abu Agag, Abu Simbel, Sabaya, and Lake Nasser Formations. All of these are covered by Cenozoic rocks of the Tertiary Dakhla, Kiseiba shale, Kurkur, Garra, and Thebes formations; the whole successions are overlain by Quaternary deposits composed of the piedmont gravels, sand sheets, Tufa, and Nile deposits, and are intruded by Oligocene and Quaternary volcanic [42–44]. The study site generally belongs to the North Africa arid belt and is characterized by arid conditions with occasional rainfall. These conditions include a long, hot summer and warm winter, where the mean air temperature of the area ranges between 14.2 °C and 38.4 °C.

The Nubian sandstone aquifer in the area represents the principal groundwater aquifer [45]. Over 200 productive wells and piezometers have been drilled, tapping the Sabaya and Abu Simbel Formations of the Nubian Aquifer. It is exposed on the surface of the study region and directly overlies the basement rocks. It is composed of fine to very coarse-grained sandstone with claystone inter-beds. It is hydraulically connected with Lake Nasser water at some localities of the area. The average effective porosity and hydraulic conductivity of this aquifer are 26% and 4.5 m/d, respectively [37,46].

3. Materials and Methods

The groundwater potential zones were delineated by collecting and preparing the spatial database, including thematic layers of the key factors influencing groundwater potentiality, followed by the processing of these layers using the AHP and FR models, and, finally, the interpretation and validation of the results. The methodology of this research is shown in Figure 2.

Water 2022, 14, 882 4 of 18

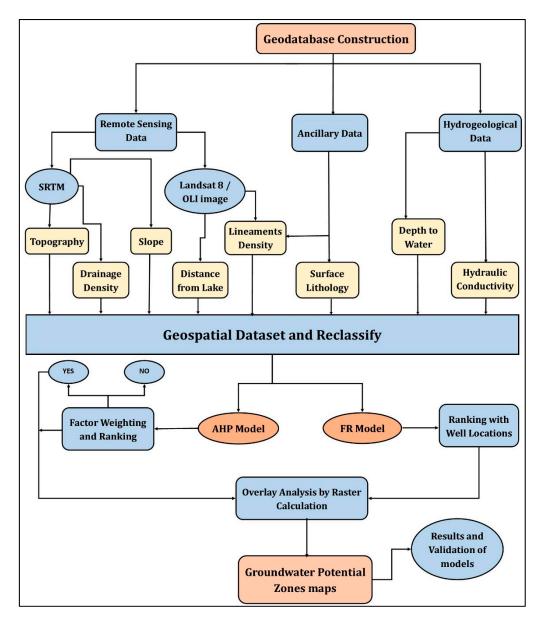


Figure 2. Flowchart showing the methodology applied in this study.

3.1. Thematic Layer Preparation

Eight factors influence the groundwater potentiality in the studied aquifer: hydraulic conductivity, distance from the Lake, lineament density, surface lithology, slope, topography, drainage density, and depth to groundwater. These factors are supposed to control groundwater recharge and groundwater flow in the study area. These factors were processed and presented in the environment of GIS to create the database and apply spatial overlay analysis. Remotely sensed satellite images, hydrogeological data, and maps were provided by governmental institutions and previously published data to generate thematic layers that influence groundwater potentiality.

3.1.1. Hydraulic Conductivity

In general, hydraulic conductivity is an important factor when considering ground-water potentiality, as rock permeability directly affects groundwater flow and storage [47]. Hydraulic conductivity reflects the subsurface lithology of the aquifer and its potentiality, as the higher the permeability, the higher the potentiality. The hydraulic conductivity layer was created from the interpolation of available values resulting from pumping test analysis and based on the previous literature [36,37,48]. The hydraulic conductivity values

Water 2022, 14, 882 5 of 18

in the area range from 0.066 to 16 m/day, where the lowest values represent the igneous and metamorphic rocks and highest values represent areas where the sandstone aquifer is dominant. The resultant map of groundwater potentiality was classified and resampled into five classes (Figure 3a).

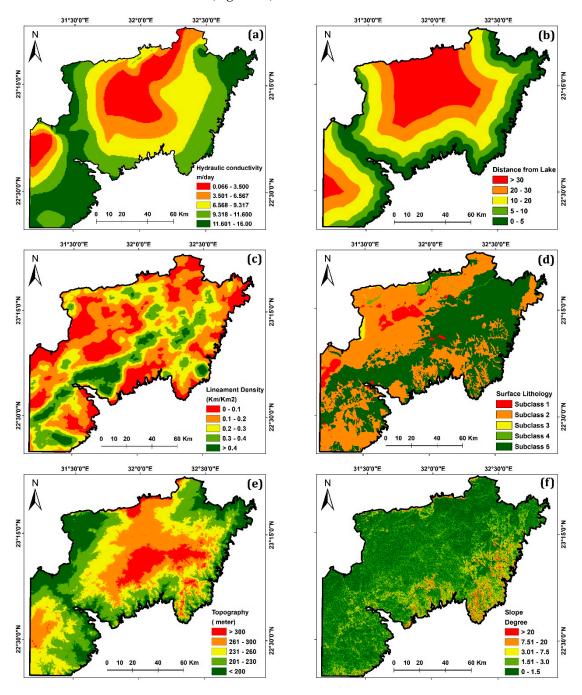


Figure 3. *Cont.*

Water 2022, 14, 882 6 of 18

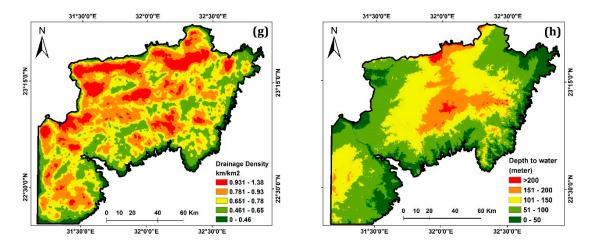


Figure 3. Thematic spatial maps (a–h) of study area: (a) Hydraulic conductivity; (b) distance from Lake; (c) lineament density; (d) surface lithology; (e) topography; (f) slope; (g) drainage density, and (h) depth to water.

3.1.2. Distance from the Lake (Lake Nasser)

The Nubian aquifer in the study area is recharged by large quantities of Lake Nasser water, causing an increase in the groundwater level of the aquifer. The detected hypothetical salts in Lake Nasser were the same as those in the Nubian aquifer, indicating the hydraulic connection between the Lake and its adjacent aquifer [37]. Ref. [38] concluded that feeding from Nasser Lake leads to an increase in the groundwater level for a distance up to 30 km from the lake; however, the areas further away from the lake are not affected (during the past 30 years). The distance from the lake layer was extracted using the distance extension and buffer tools in ArcGIS and reclassified into five classes, where the shorter the distance from the lake, the greater possibility of groundwater recharge (Figure 3b).

3.1.3. Lineaments Density

The lineament density reflects rock structures, such as fractures, faults, or joints, denoting a permeable zone that water can infiltrate [49,50]. Therefore, high-groundwater-potential zones are characterized by a high density of lineaments. Lineaments were digitized manually from the geological map of Egypt [39] and were automatically extracted from the Landsat 8 OLI image (Band 8) using PCI Geomatica software (LINE module) [51]. The linear structures were imported and processed into the ArcGIS environment to create a lineament thematic map later used to create the lineament density map in ArcGIS. The lineament density of the study area ranged between 0 and 0.67 km/km², and the area was classified into five classes, as shown in Figure 3c.

3.1.4. Surface Lithology

Surface lithology is a vital factor for groundwater potentiality, as it controls the subsurface flow of groundwater. The lithology layer was digitized from the geological map of Egypt [39]. The lithology of the study area was classified based on its influence on groundwater potentiality into five classes (Figure 3d). Groundwater activities are thought to be lower towards the north and west due to the presence of igneous and metamorphic rocks, owing to the rocks' low permeability (subclass 1). Granites, granodiorite, and gneiss are considered impermeable units. Although the sedimentary successions represented moderate to high groundwater potential, the sand sheets and clay deposits were considered unsuitable areas for drilling wells (subclass 2). The zones of Nubian sandstone rocks were considered the high potential area of groundwater (subclass 5).

Water 2022, 14, 882 7 of 18

3.1.5. Topography and Slope

Both the topography and slope layers were generated from NASA's Shuttle Radar Topography Mission (SRTM) with a 30 m resolution using the spatial analysis tool in ArcMap. Before processing the digital elevation model (DEM), a "fill" function was used to eliminate pixel value problems. Because the gravitational forces of the Earth draw water streams down, high-altitude places have a low possibility of groundwater occurrence, and low-altitude regions have a more significant probability [52]. In addition, high-elevation areas are not preferable for drilling wells due to the high cost of water extraction. The elevation of the study area ranges from 135 to 459 m.a.s.l (Figure 3e). The highly elevated areas are located in the middle region of the study area, while low-land areas are spread around the outskirts of the area, especially near the lake. The study area was reclassified into five topographic classes, as shown in Figure 3e.

Slope angle has a significant impact, as the slope determines how much surface water percolates into the groundwater [13]. The steeper the slope, the lower the infiltration rate will be due to the more considerable runoff potential [53]. The slope map was reclassified into five classes (Figure 3f), where high weight was given to flat and gentle slopes, and low weight was given to steep slopes.

3.1.6. Drainage Density

The drainage density is linked to the infiltration rate, with high drainage density representing less infiltration. The drainage density is obtained mathematically by dividing the total lengths of all the streams in a drainage basin by the drainage basin's total area. The drainage density map was prepared from the drainage map using the ArcGIS platform (line density tool) (Figure 3g). The density values of the study area ranged between 0 and 1.38 km/km². However, for groundwater potentiality, high weight was attributed to low drainage density and low weight was assigned to high drainage density.

3.1.7. Depth to Water

The depth to groundwater is defined as the vertical distance from the ground's surface to the static water level. It measures the potential distance for water to travel to the aquifer and controls the cost of water consumption and abstraction. In general, the deeper the water is, the lower the potentiality of groundwater. Water level measurement data from 44 boreholes during December 2018 were used to prepare the depth to water map of the area shown in Figure 3h. The inverse distance-weighted (IDW) interpolation method of the spatial analysis tool in ArcGIS was used to fill the gaps in null data areas. Depth to water exhibited significant variations in the study area; it ranged between 4 and 258 m. The depth to water map was classified into five classes from 1 to 5, representing potentiality classes from high to low, respectively (Figure 3h).

3.2. Analytical Hierarchy Process (AHP) Method

The groundwater potential zones in the AHP model were classified based on the index value calculated using Equation (1):

$$GWPI = C_wC_r + D_wD_r + Ld_wLd_r + G_wG_r + T_wT_r + S_wS_r + Dd_wDd_r + Sw_wSw_r \qquad (1)$$

where GWPI is the groundwater potential zone index, C is the hydraulic conductivity, G is the surface lithology (geology), D is the distance from the Lake, Ld donates the lineament density, T refers to topography, S is the slope, Dd is drainage density, Sw is the static water or depth to water, (w) is the weight, and (r) is the rate of each factor.

The weight of each factor (w) was determined using the AHP model through two main steps. The first step is the construction of a pair-wise comparison matrix between all of the (eight) influencing factors [54]. The relative importance of each parameter on groundwater potentiality was graded according to Saaty's 1–9 scale [9]. The rank of 9 was

Water 2022, 14, 882 8 of 18

given to criteria that have an extremely strong influence over the other, while a rank of 1 meant equal significance for the two compared parameters.

The relevance of each factor was determined based on the review of past studies, field experience, and expert's opinions. Accordingly, in the pair-wise comparison matrix, all concerned parameters were graded compared with each other (Table 1). In the next step, the normalized pair-wise comparison matrix was created by dividing each cell by the sum of each column, and normalized weights for each component were calculated using the average of each row (Table 2).

Table 1. Pair-wise comparison matrix between the applied parameters for the AHP mode	l.

	Hydraulic Conductivity	Distance from Lake	Lineament Density	Surface Lithology	Topography	Slope	Drainage Density	Static Water Level
Hydraulic Conductivity	1	2	2	3	4	5	5	6
Distance from Lake	½	1	2	3	3	4	4	5
Lineament Density	1/2	1/2	1	2	2	3	3	4
Surface Lithology	1/3	1/3	1/2	1	2	3	3	4
Topography	1/4	1/3	1/2	1/2	1	2	1	2
Slope	1/5	1/4	1/3	1/3	1/2	1	1	2
Drainage Density	1/5	1/4	1/3	1/3	1	1	1	2
Static Water Level	1/6	1/5	1/4	1/4	1/2	1/2	1/2	1
Sum	3.15	4.87	6.92	10.42	14.00	19.50	18.50	26.00

Table 2. Normalized pair-wise comparison matrix and weights of each factor.

	Hydraulic Conductivity	Distance from Lake	Lineament Density	Surface Lithology	Topography	Slope	Drainage Density	Static Water Level	Weights
Hydraulic Conductivity	0.317	0.411	0.289	0.288	0.286	0.256	0.270	0.231	0.294
Distance from Lake	0.159	0.205	0.289	0.288	0.214	0.205	0.216	0.192	0.221
Lineament Density	0.159	0.103	0.145	0.192	0.143	0.154	0.162	0.154	0.151
Surface Lithology	0.106	0.068	0.072	0.096	0.143	0.154	0.162	0.154	0.119
Topography	0.079	0.068	0.072	0.048	0.071	0.103	0.054	0.077	0.072
Slope	0.063	0.051	0.048	0.032	0.036	0.051	0.054	0.077	0.052
Drainage Density	0.063	0.051	0.048	0.032	0.071	0.051	0.054	0.077	0.056
Static Water Level	0.053	0.041	0.036	0.024	0.036	0.026	0.027	0.038	0.035
Sum	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000

The created matrix was checked for consistency by determining the consistency ratio (CR) using Equations (2) and (3) developed by [55].

$$CR = \frac{CI}{RI} \tag{2}$$

$$CI = \frac{\lambda max - n}{n - 1} \tag{3}$$

where RI is the random index obtained from Saaty's standard (Table 3), determined by the number of criteria, which in this study was equal to 1.41; CI is the consistency index; where λ amax is the principal eigenvalue of the matrix, which is equal to 8.19; and n is the number of parameters applied in the matrix, which was equal to 8 in the present study.

Table 3. Random consistency index (RI) introduced by [9].

	Less Im _]	portance						More Importance					
RI	0	0	0.58	0.9	1.12	1.24	1.32	1.41	1.45	1.49	1.51		
n	1	2	3	4	5	6	7	8	9	10	11		

Water 2022, 14, 882 9 of 18

Refs. [54,55] stated that the model prediction can be accepted and reliable if the CR is less than or equal to 0.1%. However, if the CR is higher than (0.1), the prediction model needs to be reassessed.

All factors were classified into subclasses and ranked based on their influence on groundwater potentiality. The lowest grade was given to the class with the most minor influence, and the highest was given to the class with the strongest influence. Finally, the normalized rate I of each subclass was determined by dividing each rank value by the total of all rankings for each factor, as shown in Table 4. In the AHP method, the groundwater potential map was constructed by overlaying the obtained controlling factors (thematic layers), which were calculated from Equation (1), using ArcMap 10.7.1.

Table 4. Assigned normalized weights and rates for all factors and subclasses.

No.	Factors	Subclasses	Rating	Weights	
		0.066–3.5	0.053		
1	Undraulia	3.5–6.567	0.158		
	Hydraulic	6.567-9.318	0.211	0.294	
	conductivity	9.318-11.6	0.263		
		11.6–16.00	0.316		
		0–5	0.316		
	D: 1	5–10	0.263		
2	Distance	10–20	0.211	0.221	
	from Lake	20–30	0.158		
		>30	0.053		
		0-0.1	0.067		
	Lineament	0.1–0.2	0.133		
3		0.2-0.3	0.200	0.151	
	Density	0.3-0.4	0.267		
		>0.4	0.333		
		Igneous and Metamorphic rocks	0.059		
	0 (Clay, claystone, fine sandstone, and sand sheets	0.118		
4	Surface lithology	Limestone of Kurukur and Garra Fms.	0.176	0.119	
		Wadi and Quaternary deposits	0.294		
		Nubian Sandstone and Gravels	0.353		
			<200	0.353	
		200–230	0.294		
5	Topography	230–260	0.176	0.072	
	101	260–300	0.118		
		>300	0.059		
		0–1.5	0.333		
		5–3	0.267		
6	Slope	3–7.5	0.200	0.052	
	•	7.5–20	0.133		
		>20	0.067		
		<0.46	0.333		
	Drainasa	0.46-0.65	0.267		
7	Drainage	0.65-0.78	0.200	0.056	
	Density	0.78-0.93	0.133		
		0.93–1.38	0.067		
		0–50	0.353		
	0	50–100	0.294		
8	Static water	100–150	0.176	0.035	
-	level	150–200	0.118	2.000	
		>200	0.059		

Water 2022, 14, 882 10 of 18

3.3. Frequency Ratio (FR) Method

The frequency ratio (FR) model is a bivariate statistical approach used as a valuable technique for geospatial evaluation to examine the probability connection between independent and dependent data, such as multi-classified maps [56]. It has recently been utilized to map groundwater potential in a specific place based on the link between the observed pumping wells and parameters regulating groundwater potential [18,21,57]. In this method, an FR value should be assigned for each subclass of the groundwater influencing parameter using Equation (4) [56]:

$$FR = \frac{W/G}{M/T} \tag{4}$$

where W is the number of groundwater wells representing each conditioning factor's subclass, G is the total number of groundwater wells in the study area, M is the number of pixels representing the subclass of the factor, and T is the total number of pixels in the study area.

In this study, 44 groundwater pumping wells were used, and the FR value for each subclass of used parameters was calculated and is presented in Table 5. The GWPI map was constructed following the FR method by integrating the FR layers of the controlling factors (thematic layers) (Equation (5)) using ArcMap 10.7.1.

$$GWPI = \sum_{i=1}^{n} FR$$
 (5)

Table 5. The spatial relationship between the factors and wells with the assigned FR for each subclass.

No.	Factors	Subclasses	No. of Pixels	Percentage of Subclass	No. of Wells	Percentage of Wells	FR
		0.066–3.5	210,565	14.70	0	0.00	0.000
	Urrdmanlia	3.5-6.567	227,811	15.91	1	2.27	0.143
1	Hydraulic	6.567-9.318	338,147	23.61	3	6.82	0.289
	conductivity	9.318-11.6	339,749	23.72	11	25.00	1.054
		11.6–16.00	315,848	22.05	29	65.91	2.988
		0–5	281,709	19.63	20	45.45	2.316
	Distance (man	5–10	227,445	15.85	17	38.64	2.438
2	Distance from	10–20	371,659	25.89	5	11.36	0.439
	Lake	20–30	259,197	18.06	2	4.55	0.252
		>30	295,346	20.58	0	0.00	0.000
		0–0.1	346,082	24.11	15	34.09	1.414
	Lineament	0.1-0.2	487,932	34.00	9	20.45	0.602
3		0.2-0.3	329,150	22.93	7	15.91	0.694
	Bensity Density	0.3-0.4	193,369	13.47	7	15.91	1.181
		>0.4	78,730	5.49	7 6	13.64	2.486
		Ign. And Meta. Rocks	52,600	3.66	0	0.00	0.000
	Surface	Clays, fine s.s, and sand sheets	697,457	48.59	17	38.64	0.795
4	lithology	L.S of Kurukur and Garra Fms.	17,444	1.22	3	6.82	5.610
	nulology	Wadi and Qs. Deposits	13,411	0.93	0	0.00	0.000
		Nubian S.S and Gravels	654,459	45.60	24	54.55	1.196
		<200	276,288	19.25	21	47.73	2.479
		200–230	384,425	26.78	19	43.18	1.612
5	Topography	230–260	356,523	24.84	4	9.09	0.366
		260–300	307,777	21.44	0	0.00	0.000
		>300	110,307	7.69	0	0.00	0.000
		0–1.5	729,970	51.13	24	54.55	1.067
		5–3	545,889	38.23	18	40.91	1.070
6	Slope	3–7.5	106,144	7.43	2	4.55	0.611
		7.5–20	32,691	2.29	0	0.00	0.000
		>20	13,106	0.92	0	0.00	0.000

Water 2022, 14, 882 11 of 18

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No.	Factors	Subclasses	No. of Pixels	Percentage of Subclass	No. of Wells	Percentage of Wells	FR
		< 0.46	927,935	6.47	0	0.00	0.000
	Duniman	0.46-0.65	3,121,085	21.75	19	43.18	1.986
7	Drainage Density	0.65-0.78	4,592,402	32.00	7	15.91	0.497
	Density	0.78-0.93	4,046,225	28.19	13	29.55	1.048
		0.93-1.38	1,664,497	11.60	5	11.36	0.980
		0–50	232,236	16.18	15	34.09	2.107
	0	50-100	573,308	39.95	25	56.82	1.422
8	Static water	100-150	454,686	31.68	4	9.09	0.287
	level	150-200	156,139	10.88	0	0.00	0.000
		>200	18,782	1.31	0	0.00	0.000

4. Results

The groundwater potential zones were estimated based on the AHP and FR models, considering the previously mentioned methodology applied to the eight most influential thematic layers.

4.1. Results of AHP Model

In the AHP model, the final weight of each influencing feature is shown in Table 4. From the pair-wise comparison results, the most weight was given to the hydraulic conductivity, distance from the Lake, lineament density, and surface lithology factors.. The CR was 0.0197, reflecting a good level of consistency in the pair-wise comparison phase. The resulting groundwater potential map obtained by the AHP model is shown in Figure 4, and was classified based on the natural breaks classification method in ArcGIS into four classes: high, moderate, low, and very low categories. In this map, most of the study area was classified as moderate potential (31.4%), 29.4% of the total area was classified as high, 25.8% as low, and 13.4% as very low groundwater potentiality.

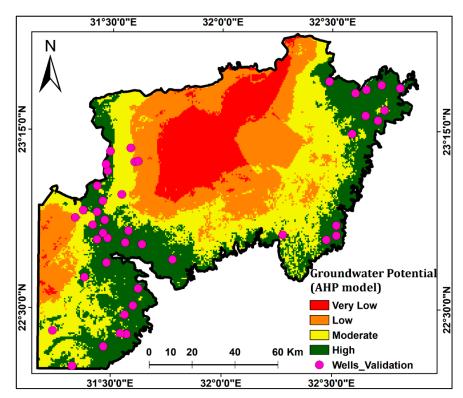


Figure 4. The groundwater potential zones map based on the AHP model.

Water 2022, 14, 882 12 of 18

4.2. Results of FR Model

According to relationships between the pumping well and each conditioning factor, the FR model was applied, and the results are shown in Table 5. An FR value of 1 implies an average geographical correlation between the location of a groundwater well and its conditioning parameter. There is a low correlation if the value is less than 1 and a stronger correlation if the value is greater than 1 [58].

The spatial relationships between the groundwater wells and hydraulic conductivity values revealed a low possibility of groundwater potential in the range of hydraulic conductivity below 9.32 m/day. Areas with distances less than 10 km from the lake had a high probability of groundwater potential, with FR values of >2. The calculations showed that as the distance from the lake increased, the ratio decreased. Additionally, the assessment of the lineament density indicated that density classes of >0.4 km/km² had the highest value of FR (2.49), followed by 0–0.1 classes (1.414). The lowest FR (0.6) value was defined for 0.1-0.2 classes. Limestone of the Kurkur and Garra Formations class had the highest value of FR (5.6) due to the limitation of that type of surface rock in the region with the presence of three pumping wells in that class, followed by the sandstone and gravel deposits class (FR = 1.2), in which most of the pumping wells were drilled in the region. The igneous and metamorphic rocks acting as an aquifuge had an FR value of zero. In case of topography, the FR values reflected a decrease in groundwater potential as the altitude increased. Similarly, the FR values were inversely proportional to the slope angle (Table 5). The locations with drainage densities of 0.46–0.65 and 0.78–0.93 km/km² possessed the highest frequency ratio (FR = 1.99, 1.05, respectively), reflecting the strong relationship with groundwater potentiality. The static water level factor showed that the regions with the lowest static water levels (0–50 m) had the maximum frequency ratio (2.11). The groundwater potential map produced by the FR model is shown in Figure 5, which was reclassified based on the natural breaks classification method into four classes: high potential (23.4%), moderate potential (28.2%), low potential (25.3%), and very low potential (23.1%).

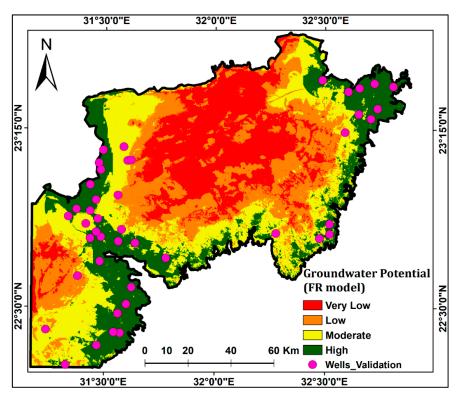


Figure 5. The groundwater potential zone map based on the FR model.

Water 2022, 14, 882 13 of 18

4.3. Validation

The validation of the results is necessary to acquire trust in the spatial approaches developed in this work and to identify their physical relevance. In the present study, the groundwater potential zone maps were validated by the overlying number of productive wells on the map (Figures 4 and 5), considering that most of these wells were pumped with a high yield.

The overlying and extraction processes within ArcGIS were performed to assume the potential zone of each well. A total of 44 existing pumping wells were identified in the study area. In the AHP method, no wells were found to be in the very low groundwater potential zone (GWPZ), only two wells were found to exist within the poor GWPZ, and 36 wells were found within the high GWPZ (Table 6), indicating a high level of validation. Likewise, only 2 of these 44 wells existed within the very poor and poor GWPZ in the FR method, and 33 wells were located within the high GWPZ. Therefore, the model was valid, as it indicated that 82% and 75% of wells existed in the high GWPZ for the AHP and FR models, respectively.

		AHP Mo		FR Model				
	Range	Area (km²)	Area%	Pumping Wells	Range	Area (km²)	Area%	Pumping Wells
Very Low	0.0787-0.142	1906.24	13.43	0	0.891-4.701	3275.38	23.07	0
Low	0.142 - 0.196	3662.00	25.80	2	4.701 - 7.38	3592.61	25.31	2
Moderate	0.196 - 0.242	4452.89	31.37	6	7.38-10.65	4004.96	28.21	9
High	0.242-0.3167	4174.97	29.41	36	10.65-20.09	3323.26	23.41	33

Table 6. The distribution of groundwater potential classes based on the AHP and FR models.

The yield of the groundwater wells was compared to the groundwater potential index calculated using the AHP and FR models. The groundwater yield values varied from 15 to 28, 30–48, and 38–90 m³/h in the low, moderate, and high-potential zones using the AHP model, respectively; while, for the FR model, they varied from 15–30, 28–70, and 32–90 m³/h in the low, moderate, and high-potential zones, respectively. The scatter plot between the groundwater potential index and well yield values showed a reasonable correlation, with coefficient of determination (R²) values equal to 0.57 and 0.47 for the AHP and FR models (Figure 6a,b), respectively. This demonstrates that the AHP model was a good model for describing zones of good groundwater yield, since it described borehole yield variability around its mean.

Moreover, the ROC curve was prepared by comparing the existing groundwater well locations in the validation datasets with the groundwater potential map [1,59].

The ROC curve of the GPMZs generated using the AHP and FR models is shown in Figure 7. These graphs showed that the AHP model (AUC = 82.6%) outperformed the FR model (AUC = 80.9%).

Based on [21], the AUC values corresponding to the prediction accuracy were divided into poor (0.5–0.6), average (0.6–0.7), good (0.7–0.8), very good (0.8, 0.9), and excellent (0.9–1). As a consequence of these validation results, the AHP and FR models used in this work both exhibited reasonable accuracy in spatial groundwater potential prediction, with the AHP technique being more efficient than the FR model in this study.

Water 2022, 14, 882 14 of 18

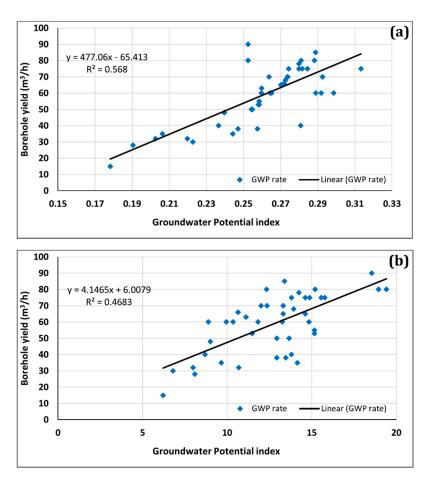


Figure 6. Scatter plots (**a**,**b**) between the borehole yield and groundwater potential index from the (**a**) AHP model and (**b**) FR model.

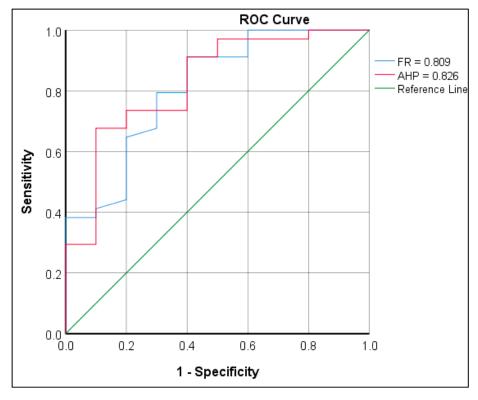


Figure 7. ROC curve and AUC for the groundwater potential zone map in the AHP and FR models.

Water 2022, 14, 882 15 of 18

5. Discussion

The groundwater potentiality in a semi-arid area surrounding Nasser Lake was mapped using the integration of GIS, remote sensing, and multi-criteria decision analysis AHP and FR. In such areas characterized by rain scarcity, factors controlling subsurface recharge are predominant, and the influence of the factors controlling the vertical seepage from the surface is lower [60]. In the area surrounding Lake Nasser, where lake water feeds the groundwater aquifer, the factors controlling groundwater flow, such as hydraulic conductivity, proximity to the lake, and lineament density have the greatest impact. The normalized matrix (Table 4) revealed that hydraulic conductivity was the most significant factor, accounting for 29.4%, followed by the distance from Lake Nasser, at 22.1%, while the slope and static water level were the least influential at 5.2% and 3.5%, respectively.

On the contrary, [13,60,61] concluded that lithology (geology), slope, and rainfall are the main influential factors of groundwater potential using the AHP model in the Vaitarna Basin of India, Motloutse Watershed of Botswana, and Tirnavos area of Greece, respectively. This is because groundwater recharge in these areas depends mainly on rainfall and recharge from flood water, while in the area west of Lake Nasser, where there is very little rainfall, recharge from Lake water prevails.

The resultant maps constructed using the AHP and FR techniques showed that the high-potential zones were located close to Lake Nasser due to the high hydraulic conductivity of the subsurface rocks, proximity to the lake, and permeable surface lithology in these areas. Furthermore, due to the low permeability, high altitudes, and steep slopes, the central part of the research site fell within very low groundwater potential zones.

Despite the high matching between the maps resulting from both AHP and FR, the spatial pattern of some zones differed somewhat. This difference may be related to the location of the pumping wells drilled in the observed low and moderate-potential zones, which increased the calculated potentiality of these areas in the FR technique.

One of the most important limitations in this study was the lack of data. Interpolation methods were used to compensate for the lack of that data to cover the entire study area. Future studies should be carried out to fill this gap in data, especially the lineament characteristics, hydraulic conductivity, and groundwater levels. Overall, the produced groundwater potential maps indicated reliable potentiality for groundwater occurrence. Accordingly, the potential areas can be easily identified on the produced maps and they can be used by the decision-makers to locate drilling sites in the ongoing reclamation projects during that period. They will be useful to avoid the wasted cost of drilling unproductive and unsustainable wells.

6. Conclusions

The southern part of Egypt is experiencing arid conditions and a significant increase in water demands due to reclamation projects. Consequently, assessing and evaluating the groundwater potentiality west of Lake Nasser can provide highly beneficial results for water resource management. There have been no studies on the southern parts of Egypt using approaches such as MCDA. Therefore, AHP and FR were used to carry out the present work to assess groundwater potential zones. In this study, an attempt was made using eight groundwater controlling factors to develop a groundwater potential zone map. The study showed that the high and moderate-potential zones were 61% to 52% of the total area for AHP and FR, respectively. Therefore, the results indicated that 82% and 75% of wells existed in the high GWPZ for the AHP and FR models, respectively. The performance and validation of the employed models were conducted using the ROC curve. The ROC curve accuracy for both employed approaches, i.e., AHP and FR, revealed accuracy values of 83% and 81%, indicating a better predictive performance rate for the AHP approach.

The AHP and FR-based approaches for delineating groundwater potential zones used in this study were effective due to their ability to generate reliable results that could be applied in the development of semi-arid regions, especially in developing and low-income countries. It can be argued that the complexity and highly dynamic nature of the

Water 2022, 14, 882 16 of 18

relationship between the lake and adjacent aquifer, the lineament characteristics, aquifer hydraulic parameters, and fluctuation of water levels in both systems, making it is necessary to conduct monitoring and integrating water levels in future studies.

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