# PRECISION AGRICULTURE USING ADVANCED REMOTE SENSING TECHNIQUES IN ARID LANDS

#### By

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B.Sc. Agric. Sc. (Natural Resources & Agric. Engineering), Alex. Univ., 2007
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M.Sc. Agric. Sc. (Soil and Water), Damanhour University, 2013

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### **Approval sheet**

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#### **ABSTRACT**

Mohamed Mortada Ragab El-Sharkawy: Precision Agriculture Using Advanced Remote Sensing Techniques in Arid Lands, Unpublished PhD thesis, Arid Land Agricultural Graduate Studies and Research Institute, Faculty of Agriculture, Ain Shams University, 2018.

One of the most important tools for precision agriculture is remote sensing data and GIS advanced techniques. The main objectives of the current study are to establish a management zones based on various soil and nutrient maps linked to crop productivity. Monitoring plant water consumption and nutrient availability help for increasing the efficient use of water and other inputs. Two sites were selected in El-Salhiya and East of Beni-Suef to represent two regions and different management practices in irrigated desert lands. Soil samples were collected from both study areas. The collected soil samples were analyzed and nutrient availability were measured. Data obtained were used to identify different soil management zones. In order to evaluate the role of remote sensing data, Landsat satellite data (2014-2016) were selected and processed. The advanced resolution merge techniques were used to enhance the spatial resolution from 30 meters in Landsat sensors to 5 meters of Rapideye data which specially designed for precision agriculture service. In this study, image fusion using Principal Component Spectral Sharpening (PCSS) method was applied to integrate NDVI and plant water consumption calculated from Landsat satellite data. Moreover, four hyperspectral vegetation indices were calculated from satellite data.

The obtained results showed that remote sensing data with soil data analyses allowed for the identification of spatial pattern of crop growth variability. Data indicated that peanut yield was mainly affected by soil variability obtained from different spatial maps. Using the soil suitability model and a sufficient number of field observations within each class, an acceptable accuracy and good spatial distribution of the suitability classification was achieved. The empirical regression growth model was able to predict the yield across the field when the correct inputs were used, showing great potential for use in yield map prediction and interpretation in the contest precision agriculture. Results of Tukey's HSD showed that Blue, Red and NIR spectral zones are more sufficient in the monitoring differences between peanut growth stages than green, SWIR-1 and SWIR-2 spectral zones. Also, there are significant correlations between varied classes productivity and spectral similarity measures, indicating that similarity between the samples' spectra decreases as the pigments concentration in the plant leaves increases, which offer as a precision agriculture tool to manage crop variations within fields that can affect crop yield. Moreover, resulting yield predictions showed a high agreement with field data with good significant correlation coefficient. From the obtained results we concluded that remote sensing spectral data with their appropriated derived indices are important source for producing various vegetation-soil models and controlling inputs based on soil and plant requirements in irrigated lands of arid climate.

**Key words:** Precision agriculture, Peanut, Data fusion, Hyperspectral indices; arid land, SMCE, Remote sensing and GIS.

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