

Monitoring And Yield Estimation of Sugarcane Using Remote Sensing and GIS

*Abdelrahim Elhag¹, Abdelaziem Abdelhadi² (14)

¹(Department of GIS, College of Engineering/ Sudan University of Science and Technology, Sudan) (10 Italic)

²(White Nile Sugarcane Company, Sudan) (10 Italic)

Corresponding Author: *Abdelrahim Elhag¹

ABSTRACT: Earlier estimation of sugarcane yield is a main requirement for sugar production. Many operations of sugar production are based on yield estimation, such as planning for the season management including manpower, transportation, storage, marketing ...etc. In this study NDVI had been taken as means of the estimation of sugarcane yield in White Nile sugar factory. Therefore, the objectives of this study is to: (1) determine the ability of an in-season estimation of NDVI to predict sugarcane yield potential; and (2) determine optimum timing for predicting sugarcane in-season yield potential. The actual produced sugar quantities, for the same area and the same period of time, in the previous seasons had been compared with the NDVI results. The study, which had been applied in two successive seasons, had shown more effectiveness than the traditional direct field estimation. Due to the wide areas of the sugarcane fields (50-100) hectares, the height of the plants (1.5-2.00) meters, high cost of transportation together with the limitations of human eye limitations, direct ground field estimation has low precision and feasibility. The precision of estimation of the traditional method was 50% while of NDVI was 90%. In conclusion Geographic Information Systems (GIS), coupled with the remote sensing has proven to be successful in estimating crop yield using the NDVI. This estimation will help in the management of the season and decision for sugarcane farms right on time.

Keywords : GIS, Remote Sensing, Sugarcane, Yield Estimation, (10 Italic)

Date of Submission: 02-01-2018

Date of acceptance: 22-01-2018

I. INTRODUCTION

Sugarcane (*Saccharum officinarum*) is a semi-perennial crop, which can be harvested annually up to five years without replanting; the first harvested crop is termed plant cane and stubble cane for each successive harvest which named as ratoon cycles. (R1-R4) Sugarcane growth season is varying from only nine months to more than 12 months in some places. The sugarcane growth season at Sudan consists of planting in October, and harvested during the winter months (Nov-Mach). During this growth season the sugarcane crop undergoes three distinct growth stages consisting of emergence (crop establishment) at first 3 months, tailoring, (vegetative growth) at next 6 months, and maturation at last 3 months prior to replanting. In recent decades, significant yield increases have been attributed to the addition of N fertilizer beyond any other agricultural input [2]. Nitrogen (N) is one of the most important crop growth factors, influencing both productivity and crop quality. Therefore, utilizing methods that can more accurately determine N rate recommendations is essential to maintain agronomic productivity [3]. Above long growth cycles combined with the shorter growth period makes accurate N rate recommendations that optimize yields and minimize environmental impacts difficult. Worldwide N recommendations for sugarcane production are dependent on climate, crop age, length of growth cycle, plant characteristics, and soil characteristics [3]. However, currently for Sudan sugarcane projects N rate recommendations are dependent on crop cycle, either plant cane or stubble cane, and soil type, generalized as light or heavy textured soils and crop N demand [1]. These N rate recommendations are applied in a split application (half dose at the beginning of planting process) then another half at Unbarring period. Historically, soil sampling has been a technique utilized for determining N rate recommendations. Therefore, crop yield monitoring has become an important aspect of many N management schemes. A common method of incorporating crop yield into N rate recommendations is using yield goals specifically in crop production [5]. A

yield goal is defined as yield per unit area we might expect. Many plant indices based on canopy spectral reflectance have shown the ability to accurately estimate crop physiological properties, including plant biomass and crop yield [10–12]. The NDVI value, which is a vegetative index that compares reflectance at the red and near infrared region, has also shown the ability to determine yield potential (YP) [13–15]. Yield potential differs from yield goal because it is a function of the environmental conditions of the current growing season and is defined as achievable yield with no additional N fertilizer [11]. Teal [14] reported that there was a strong relationship between NDVI and Crop yield in Cron using an exponential model. Raun [13] and Lukina [16] showed this relationship was improved when NDVI readings were adjusted using growing degree days (GDD), where NDVI was divided by GDD accumulated from planting to sensing, to create an in-season estimate of yield (INSEY). Raun [13]. Several studies have suggested that growth stage, or time of sensing, were important in the ability to predict yield [13,14,16]. Raun [13] and Lukina. Several reports have shown that an estimate of yield alone is poorly correlated with optimum N rate [17]. However, Raun [11] showed the potential of utilizing a predicted YP as a component of N management scheme. This technology has shown the ability to improve N management decisions in many cropping systems across USA, Canada, Mexico, and other countries [18–20]. These reports suggest the potential of using yield prediction as an integral part of an N management decision tool to improve recommendations in crop production. Previous reports have documented the ability of NDVI to estimate sugarcane yield potential, however, most of these reports have been focused on satellite based platforms or passive sensors with few demonstrating the ability of an active ground-based remote sensor to estimate sugarcane yield [21–25]. Therefore, the objectives of this study were to: (1) determine the ability of an in-season estimation of NDVI to predict sugarcane yield potential; and (2) determine optimum timing for predicting sugarcane in-season yield potential.

II. MATERIALS AND METHODS

2.1 Study Area

The study had been conducted in the White Nile sugar project which is in Sudan in the White Nile state near Aldoium city, It is about 180 Kilometre southern The Capital (Khartoum) and between the (1580650,428010 15) and (19192 461954) Coordinate system WGS_UTM_ZONE 36N It occupies a space (165000) Fadden (683000) Hectare.

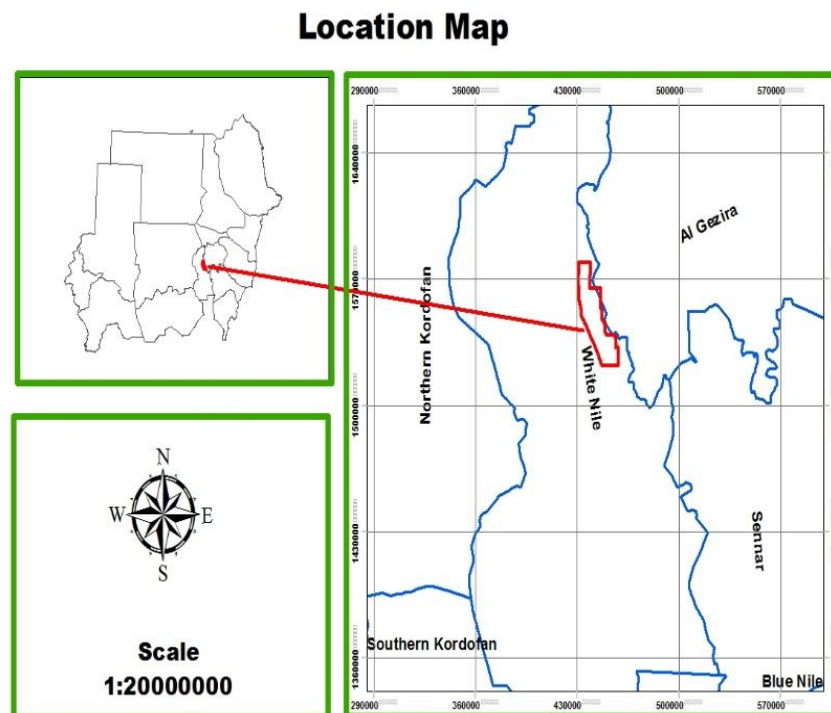


FIG. 1: The Study Area

2.2 Sources of Data

The green areas were obtained from the map of the project 2014, the tables of production for the previous years, and land sat 8 band 4 and 5.

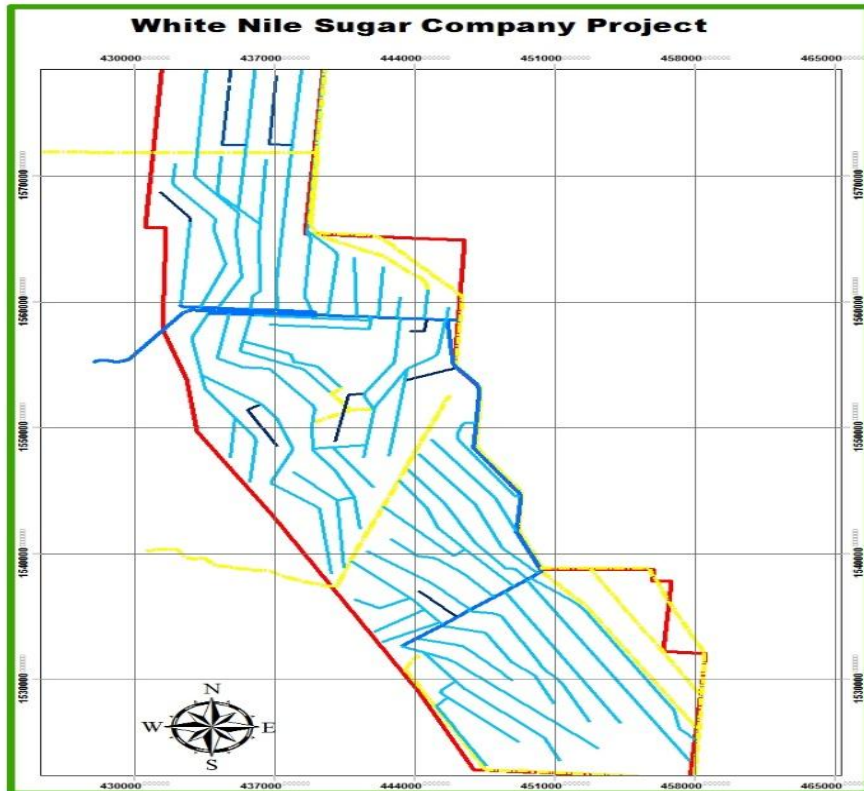


Fig. 2: White Nile Sugar Company Project

Table 1: Season 2014 – 2015

Canal	Field	Crop Cycle	Variety	Area for Harvesting	Area harvested	Ton	TCF
P4SS	13	R2	Co6806	41.3	41.2	1406	34.14
P2SS	15	R1	Co6806	30.8	30.8	1402	45.57
P2SS	16	R1	Co6806	36.7	36.7	1540	41.98
P2SS	17	R1	Co6806	35.6	35.6	1638	46.05
P1SS	7	R1	Co6806	25.2	25.2	1027	40.8
P1SS	8	R1	Co6806	44.7	44.7	1939	43.41

Table 2: Season 2015 – 2016

Canal	Field	Crop Cycle	Variety	Area for Harvesting	Area harvested	Ton	TCF
Canal 9	1	R6	Co6806	110	68	2255.9	33.18
P17AS	3	R2	Mix	10	10	327.8	32.78
P20AS	27A	R1	Mix	6	6	225.6	37.60
Canal 2S	15	R2	Co6806	30.7723	30.7723	1042.2	33.87
Canal 2S	16	R2	Co6806	36.6723	36.6723	1222.9	33.35
P6WSS	2	P.C	TUC	26.5	26.5	894.25	33.75

Table 3: Season 2016 – 2017

Canal	Field	Crop Cycle	Varity	Area for Harvesting	Area harvested	Ton	TCF
Canal MJ3	Field 13	R1	co6806	47	47	1425.9	30.34
Canal MJ3	Field 12	R1	TUC	46	46	2601.82	56.56
Canal MJ3	Field 11	R1	TUC	64	64	3359.36	52.49
Canal MJ3	Field 10	R1	24 TUC / 19 R579	43	43	2134.53	49.64
Canal MJ3	Field 9	R1	13 TUC /28.5 R579	41.5	41.5	1752.64	42.23
Canal 11S	Field 8	R2	Co6806	128.772	128.772	4338.07	33.69

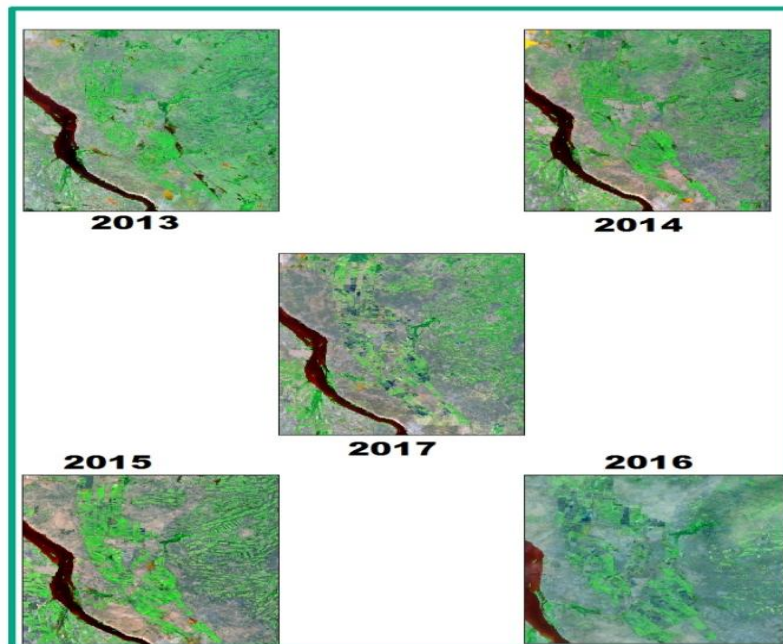


Fig. 3: Image of the area from Land Sat 8

3.3 Data Processing

The land sat 8 band 4 and 5 were used to give the NDVI using the GIS, the maps of the NDVI were converted into quantitative data using the global mapper which convert it into excel sheet. the NDVI was calculated at different periods during the season starting from the planting till the cultivation in order to determine the best time for the estimation of the quantity of the harvest. the produced quantity in the previous seasons were compared with the NDVI for the same period of time. for the farms that are planted for the first time the comparison was made with farms that has the same type of soil and the ratoon, when we used the NDVI of any farm initially regardless of the type of soil and the ratoon the prediction was less precise.

The following equation was used to estimate the production:

$$P = LP * \frac{NDVI}{NDVIp}$$

Equation 1: The equation used to calculate productivity

Where p is the estimated production.
 LP: the production in the previous season
 NDVI: the indicator of the current season
 NDVIp: the indicator of the previous season for the same farm

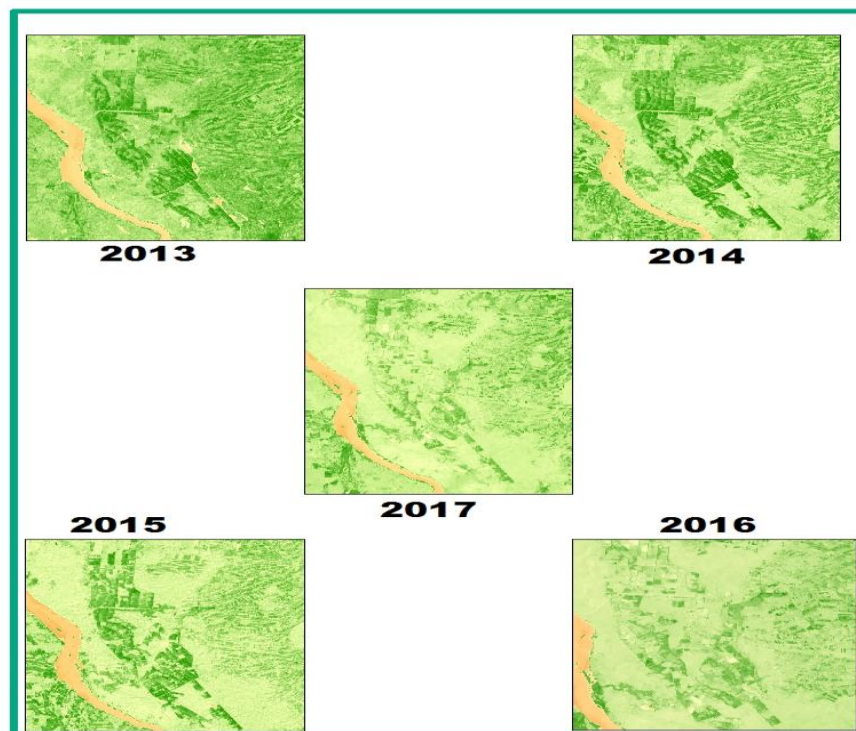


Fig. 4: Land Sat 8 Image NDVI Analysis

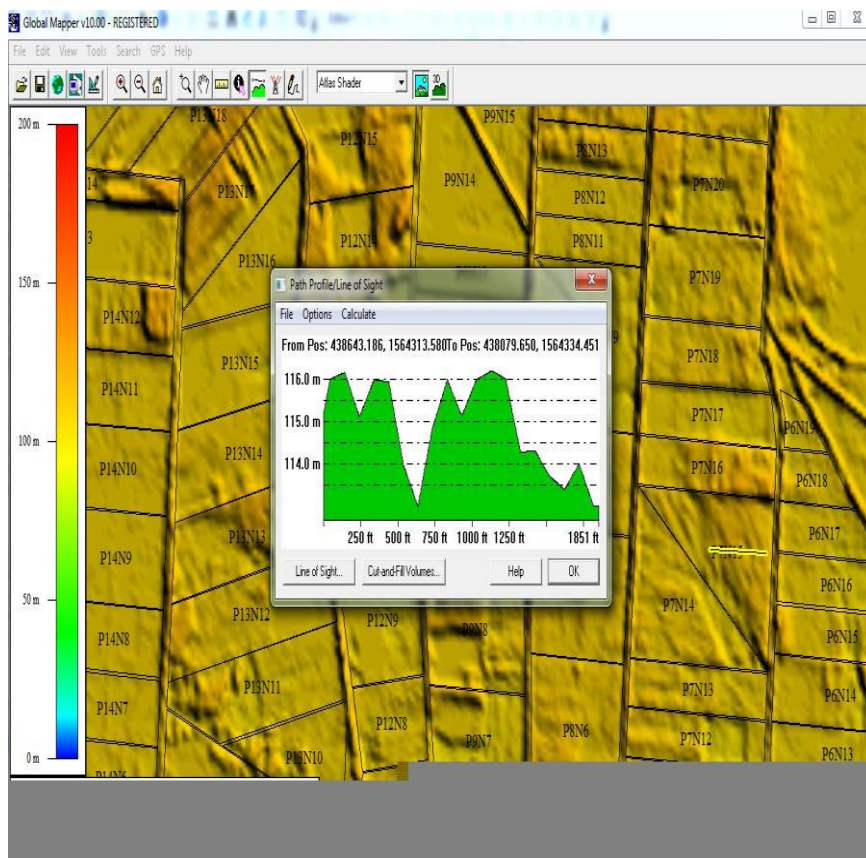


Fig. 5: Calculating NDVI value using Global Mapper Software

III. RESULTS AND DISCUSSION

In the first season the precision for the previously planted farms was 95% and for the new farms was 85%. The estimation for the new farms

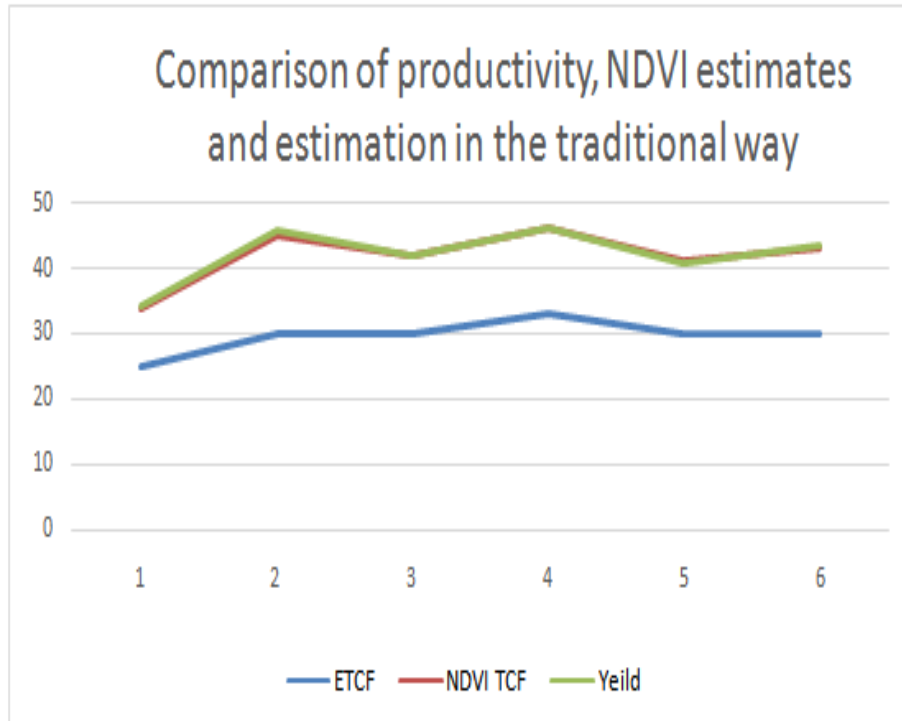


Fig. 5: Results of the 2014-2015 season

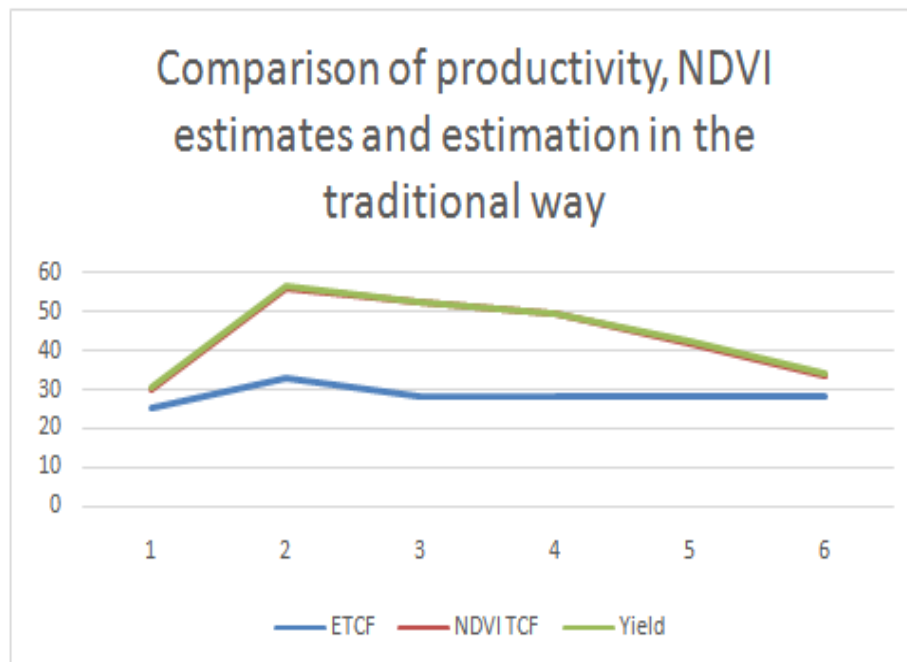


Fig. 6: Results of the 2015-2016 season

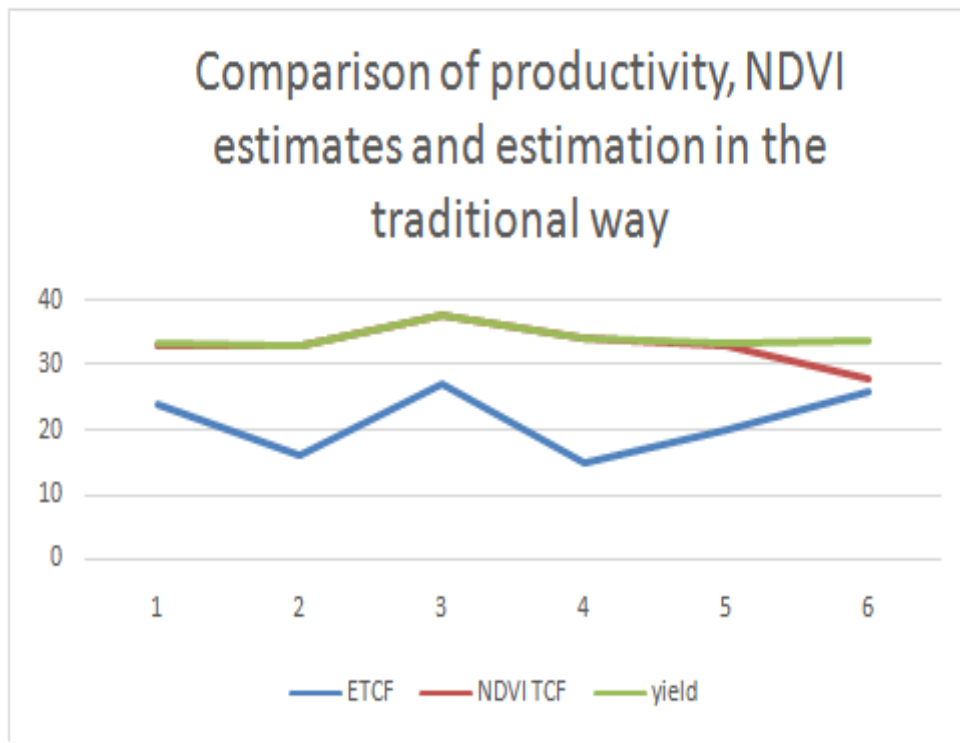


Fig. 7: Results of the 2016-2017 season

And it was improved in the following season by comparing the farm to a farm with the same type of soil and the same ratoon to reach 90%.

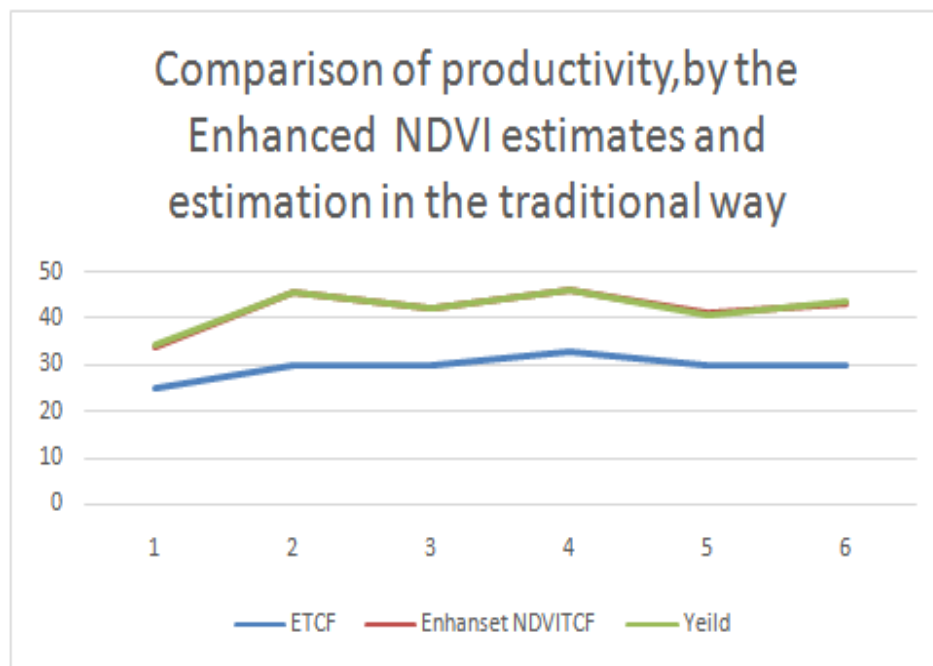


Fig. 8: Enhanced Results of the 2014-2015 season

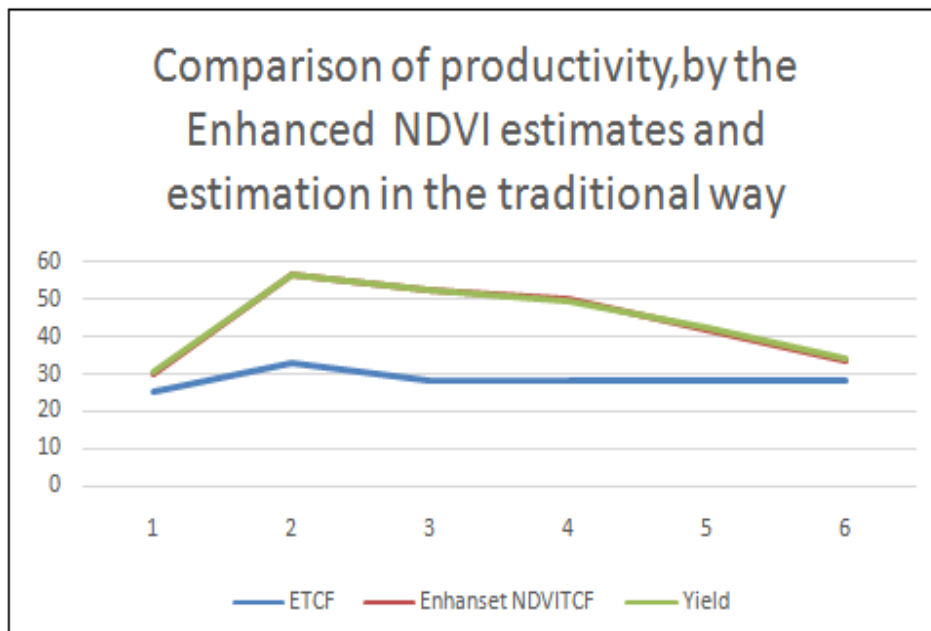


Fig. 9: Enhanced Results of the 2015-2016 season

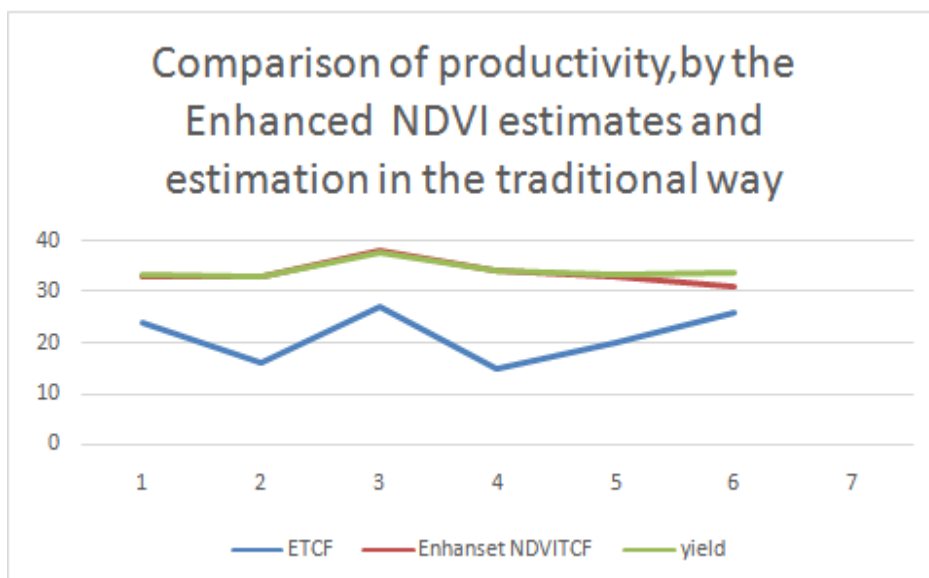


Fig. 10: Enhanced Results of the 2016-2017 season

The new method was tried in two seasons and proved to be more effective than the traditional method in which they calculate the quantity produced by certain area (e.g. 10 m²) and then estimate the production for the rest of the field based on it. The precision of the traditional method was 50%.

Table 4: Season 2014 – 2015 Estimations Results

Canal	Field	Crop Cycle	Variety	ETCF	NDVI TCF	Enhanced NDVI TCF	Yield
P4SS	13	R2	Co6806	25	34	34	34.14
P2SS	15	R1	Co6806	30	45	45.5	45.57
P2SS	16	R1	Co6806	30	42	42	41.98

P2SS	17	R1	Co6806	33	46	46	46.05
PISS	7	R1	Co6806	30	41	41	40.8
PISS	8	R1	Co6806	30	43	43	43.41

Table 5: Season 2015 – 2016 Estimations Results

Canal	Field	Crop Cycle	Variety	ETCF	NDVI TCF	Enhanced NDVITCF	Yield
Canal MJ3	Field 13	R1	co6806	25	30	30	30.34
Canal MJ3	Field 12	R1	TUC	33	56	56.5	56.56
Canal MJ3	Field 11	R1	TUC	28	52.5	52.5	52.49
Canal MJ3	Field 10	R1	24 TUC / 19 R579	28	49.5	50	49.64
Canal MJ3	Field 9	R1	13 TUC / 28.5 R579	28	42	42	42.23
Canal 11S	Field 8	R2	Co6806	28	33.5	33.5	33.69

Table 6: Season 2016 – 2017 Estimations Results

	Field	Crop Cycle	Variety	ETCF	NDVI TCF	Enhanced NDVITCF	Yield
Canal 9	1	R6	Co6806	24	33	33	33.18
P17AS	3	R2	Mix	16	33	33	32.78
P20AS	27A	R1	Mix	27	37.5	38	37.6
Canal 2S	15	R2	Co6806	15	34	34	33.87
Canal 2S	16	R2	Co6806	20	33	33	33.35
P6WSS	2	P.C	TUC	26	28	31	33.75

IV. CONCLUSION

Geographic Information Systems (GIS), coupled with the remote sensing has proven to be successful in the estimation of crop yield using the NDVI (90%). The achieved estimation percentage is appropriate for planning and managing the sugarcane production season properly.

REFERENCES

- [1]. LEGENDRE B.L., SANDERS F.S., GRAVOIS K.A, SUGARCANE PRODUCTION BEST MANAGEMENT PRACTICES. (LOUISIANA STATE UNIVERSITY AGRICULTURE CENTER, BATON ROUGE, LA, USA, 2000).
- [2]. Johnson A.E. Efficient Use Of Nutrients In Agriculture Production Systems, Comm. Soil Sci (Plant Anal, 2000).
- [3]. Wiedenfeld B, Effects Of Irrigation And N Fertilizer Application On Sugarcane Yield And Quality, Field Crop Res, Volume 43, Issues 2–3, October 1995, 101-108 .
- [4]. Wilson W.S, Advances In Soil Organic Matter Research: The Impact Of Agriculture And The Environment (Royal Society Of Chemistry, Cambridge, Uk,1991).
- [5]. Tucker C.J. Red And Photographic Infrared Linear Combinations For Monitoring Vegetation, Remote Sens. Environ. Volume 8, Issue 2, May 1979, 127-150.
- [6]. Raun W.R., Solie J.B., Johnson G.V., Stone M.L., Mullens R.W., Freeman K.W., Thomason W.E., Lukina E.V. Improving Nitrogen Use Efficiency In Cereal Grain Production With Sensing And Variable Rate Applications. Agron. J, Volume 94, Issue 4, 2001, 815-820.
- [7]. Zhao D., Reddy K.R., Kakani K.G., Read J.J., Carter G.A. Corn Growth, Leaf Pigment Concentration, Photosynthesis, And Leaf Hyper-Spectral Reflectance Properties As Affected By Nitrogen Supply. Plant Soil. Volume 275, Issue 1, November 2003, 205-218 .
- [8]. Raun W.R., Solie J.B., Johnson G.V., Stone M.L., Lukina E.V., Thomason W.E., Schepers J.S. In-Season Prediction Of Potential Grain Yield In Winter Wheat Using Canopy Reflectance. Agron. J. Volume 93, Issue 1, 19 Jan 2000, 131 - 138.
- [9]. Teal R.K., Tubana B.S., Girma K., Freeman K.W., Arnall D.B., Walsh O., Raun W.R. In-Season Prediction Of Corn Grain Yield Potential Using Normalized Difference Vegetation Index. Agron. J. Volume 98, Issue 6, 4 Apr 2006, 1488 - 1498.
- [10]. Harrell D.L., Tubana B.S., Walker T.S., Phillips S.B. Estimating Rice Grain Yield Potential Using Normalized Difference Vegetation Index. Agron. J. Volume 103, Issue 6, 27 June 2011, 1717 - 1727.
- [11]. Lukina E.V., Raun W.R., Stone M.L., Solie J.B., Johnson G.V., Lees H.L., Laruffa J.M., Phillips S.B. Effect Of Row Spacing, Growth Stages, And Nitrogen Rate On Spectral Irradiance In Winter Wheat. J. Plant Nutr. Volume 23,2000, Issue 1, 21 Nov 2011, 103 - 122.
- [12]. Kachansoki R.G., O'halloran I.P., Aspinall D., Von Bertoldi P. Delta Yield: Mapping Fertilizer Nitrogen Requirement For Crops. Better Crops. Volume 90, Issue 3, 1996, 20 - 23.
- [13]. Olf H.W., Blankenau K., Brenttrupm F., Jasper J., Link A., Lammel J. Soil- And Plant-Based Nitrogen Fertilizer Recommendations In Arable Farming. J. Plant Nutr. Soil Sci. Volume 168, Issue 4, Aug 2005, 414 - 431 .
- [14]. Tremblay N., Belec C. Adopting Nitrogen Fertilization To Unpredictable Season Conditions With The Least Impact On The Environment, Hort. Technol. Volume 16, Issue 3, 2006, 408 - 412.
- [15]. Zillman E., Graefe S., Link J., Batchelor W.D., Claupein W., Assessment Of Cereal Nitrogen Requirement Derived By Optical On-The-Go Sensors On Heterogeneous Soils, Agron. J. Volume 98, Issue3, 2 Sep 2005, 682 - 690 .

- [16]. Abdel-Rahman E.M., Ahmed F.B., The Application Of Remote Sensing Techniques To Sugarcane (Saccarum Spp. Hybrid) Production: A Review Of The Literature, *Int. J. Remote Sens.* Volume 29, Issue13, 2008, 3753 - 3767.
- [17]. Begue A., Lebourgeois V., Bappel E., Todoroff P., Pellegrino A., Baillarin F., Siegmond B. Spatio-Temporal Variability Of Sugarcane Fields And Recommendations For Yield Forecasting Using Ndvi. *Int. J. Remote. Sens.* Volume 31, Issue20, 2010, 5391 - 5407.
- [18]. Begue A., Todoroff P., Pater J., Multi-Time Scale Analysis Of Sugarcane Within-Field Variability: Improved Crop Diagnosis Using Satellite Time Series, *Precis. Agric.* Volume9, Issue 3, Jun 2008, 161 -171.
- [19]. Simoes M.D.S., Rocha J.V., Lamparelli R.A.C., Spectral Variables, Growth Analysis And Yield Of Sugarcane. *Sci. Agric.* Volume 62, Issue 3, Jun 2005 .
- [20]. Rao P.V.K., Rao V.V., Venkataratnam L., Remote Sensing: A Technology For Assessment Of Sugarcane Crop Acreage And Yield, *Sugar Tech.* Volume 4, Issue3-4, Sep 2002, 97 - 101.

Abdelrahim Elhag."Monitoring And Yield Estimation of Sugarcane Using Remote Sensing."
American Journal of Engineering Research (AJER), vol 7, no.1, 2018, pp. 170-179.