#### **Research Article**

# Yield Estimation and Sensitivity Analysis Using DSSAT Software in Assosa, Weste

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Abstract

ity, Ethiopia

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cted in the Assosa university f and validating maize cultivar coef cie alibrated and evaluated using 2019 and valid ntivar showed, SHONE is highest in grain filing rate, development due to photo-period sensitivity, and highest i (nRMSE) was Zero for days to anthesis, days to maturity and and dry biomass yield as calculated by DSSAT during calibrat -stat (>85) for most measured crop parameters during v d and simulated values of maize cultivars are simila ohysiological maturity, dry biomass yield, grai rformance of the diferent cultivars can anting when changed, starting from vield, but yield on May 1 wa increase 1-15), while pre e ferditoanAdsisigneds Apr-2024, pre QC No: acst-24 8-Apr-2024, QC N

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nt agroecologies, are the major causes

019 and 2020 on maize cultivars with the SAT software version 4.7.5. Five-cultivar e 2020 crop data. The Genetic coe f cient E, BH545 and MH138 are highest in delay of kernels. Normalized diference RMSE hile it was between 0-20% for leaf number ous crop growth showed higher R2 (>90 e calibration results of 2019 showed that near 1, for at least fve measured value nit grain weight. The sensitivity analys ing on the climatic condition that cou gh to May 10 and May 20 continuou ared to actual time of planting ( tress conditions improving the

Keçikainds Haila BE (2024) his el Californation: avalian and in the materia of Malzeet Zatial pote. Plan Alife Bille & Plan Ring HQ DFIFHVV DUWLFOH GLVW the hard bille bille de Canada and the company of Malzeet Zatial pote. Creative Commons Attribution License, which permits unrestricted (Jone et al., 2003) fo

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edited.

DSSAT is one best crop simulation so ware, which is showing best promises in yield estimation and yield prediction over long time enabling decision support to di erent parties (Abera et al., 2018; Hoogenboom, et al.2012). DSSAT So ware package has gone through di erent progress since its development with only few crops and four crop models (CERES maize, CERES-wheat, SOYGROW, NUTGROW) (Jone et al., 2003; Kaleita et al., 2020) to more than 42 crops and addition of new modules (Kaleita et al., 2008; Sachin et al., 2019; Abayechaw, 2021). Its advancement for use in many operating systems ( orp et al., 2011), data interchange system and its ability to be integrated with di erent so ware (Dzotzi et al., 2013; orp et al., 2008) inclusion of new modules (Kaleita et al., 2020) used in precision agriculture (Paz et al., 2001a, 2003; orp et al., 2008) and with its betterment in the challenges in formatting input and output les (Kaleita et al., 2008) it has become more user friendly, and is well validated for a number of regions and crops ( rop et al., 2008) [1].

In the same line as the progress in DSSAT developers the progress that demand in the users of DSSAT in the agricultural sector should also be in line by evaluating and validating the di erent new interfaces of DSSAT added from time to time and covering most areas of the agricultural ecosystem while, adjusting crop coe cients of cultivars is needed for its e cient utilization (Hoogmboom et al., 2020) [2]. demand continues in evaluation and validation of crop parameters for various crop genes by environment interactions so that we can address the di erent combination of e ects from changes occurring on the environment (Abera et al., 2018) and changes occurring by varying di erent technologies (orp et al., 2008) and at the end it will be possible to nd ways to modify or optimize the models within DSSAT for our local condition and speci c crop (Jing-yi et al., 2012) [3].

Environmental changes that arise due to climate change and variability across agro-ecosystems (Abera et al., 2018) as well as the

(Jone et al., 2003) for uctuations and yield gap problems that are widely seen in most cropping systems and have predicted future impact on the crops productivity (Abera et al., 2018; Mulune et al., 2015) [4]. For example in Ethiopia climate change perdition across time between 2010-2099 showed a decrease in maize yield by more than 24 % at the end of the century. e use of adaptation strategies, such as, the best cultivar of maize and change in the date of planting will have a salvaging e ect up to 12% yield reduction during predicted years of 2012 -2040 (Eulenstein et al., 2017). Being at the start of long term climate change predictions this study is done with the objective of Evaluating and validating di erent cultivars using DSSAT-CSM, while testing the maize cultivar sensitivity based on some anticipated climate change scenarios [5].

# **Materials and Methods**

#### **Description of the Study Site**

e experiment was conducted for two seasons in a warm subhumid lowlands agro-ecology possessing one altitude feature in the region. e major agroecology covering vast area of the region are

the warm moist lowlands and warm sub-humid lowlands having distribution in all the Zones and most districts. Assosa district is selected to represent warm moist lowlands agroecology. e study site is geographically located at 340 31'E longitude and 10004'N latitude with an altitude 1580 meters and it is approximately 660 km west of the capital, Addis Ababa [6].

#### **Description of the study materials**

Five maize varieties adapted to the agroecology, which are high yielding; resistant to disease and recently released varieties, were selected for the study. Based on the selection criteria the ve varieties were, Shone (Pioneer), Melkasa6, MH138, BH545 and local variety. Blended NPS and KCl (60% K) fertilizer was used to supply the three major nutrients, nitrogen, phosphorus and potassium and one micro-nutrient which is de cient in the soil of the study site (Table 1) [7].

#### Soil Sample and sampling methods

Soil samples were taken from the whole eld at 10 points and from 4 depths (0-20, 20-40, 40-60 and 60-80 cm) before treatment application and from each plot at 3 points diagonally a er crop harvest from 30 cm depth of and samples from the similar experimental unit were composited [8].

Soil physical properties like soil texture and soil dry bulk density, accompanied by chemical properties were tested following standard methods, in the Assosa University Soil Lab. Soil texture were determined using density method proposed by Bouyoucos (2003); the dry bulk density were measured by core sampling method of Black (2003); the soil pH (1:2.5) by pH meter (potentiometric analysis) (Jackson, 2003); the percent organic carbon content using wet potassium dichromate oxidation method (Walkley and Black, 2003); while the exchangeable K was measured by ame photometer; total N by kjeldahl digestion method (Jackson, 2003); and available P by Bray No 1 method (Bray et al., 1945) [9].

### Treatments and design of the experiments

e ve cultivars of maize SHONE (Pioneer), MELKASA6, MH138, BH545; and one local cultivar that was planted under two nutrient condition one with (NPS and KCl) and the second one without (NPS and KCl) fertilizers. e four cultivars and the local cultivar with two nutrient situations were planted for two seasons as single factor experiment. e six (6) treatments was, planted on plot size of 4.5 m x 4.5 m with plant spacing of 75 cm and 30 cm between rows and between plants on a row, respectively, for all the cultivars. e experiment was laid in RCBD design, with three replications.

## **Data collected**

#### Crop data

Data on days to emergence, days to anthesis and silking and days to physiological maturity; and crop data on four important stages was measured to calculate the genetic parameter, like P1, P2, P5, G2, G3 and PHINT at each leaf appearance. P1 is the thermal time from seedling emergence to the end of the juvenile phase (expressed in degree days, oC day, above a base temperature of 8 oC) during which the plant is not responsive to changes in photoperiod. P2 is the extent to which development (expressed as days) is delayed for each hour increase in photoperiod above the longest photoperiod at which development proceeds at a maximum rate (which is considered to be 12.5 h). P5 is the thermal time from silking to physiological maturity (expressed in degree days above a base temperature of 8 oC). G2 is Maximum possible number of kernels per plant. G3 kernel lling rate during the linear grain lling stage and under optimum conditions (mg day-1). And PHINT is the Phyllochron interval; the interval in thermal time (degree days) between successive leaf tip appearances, to record the phyllochron the plants were observed every day starting from emergence until owering [10].

Growths of maize such as the leaf area index and plant height were recorded at each full appearance of new leaves until the end of leaf growth and start of owering. Plant samples were selected from the central plant rows for measuring the LAI and the plant height. Yield component data were also measured at physiological maturity, while the grain yield and nal dry biomass yield were taken at time of full maturity. All plant parts (leaves, stalk and the husk) were separated dried and summed up for dry biomass yield [11].

## **Climate and soil data**

A 40 year data between (1980-2020) on ve climate variables, solar radiation, maximum temperature, minimum temperature, relative humidity and rainfall was collected from Ethiopian meteorology agency of Benishangul Gomez region, however because of high numbers of missing data between (1980-2000) only 20 year data between (2000-2020) was used for the calibration and validation purpose as completeness of climate data is more important than the numbers of years (Hognboom, et al., 2012).

During the period of the experiment in the 2019 the amount of rainfall in the growing period was 967 mm which is slightly lower than the rain fall amount in the growing period of 2020 experiment year 987 mm. e min and max temperature during 2019 was slightly higher than 2020 in the growing period the di erent climatic variable in the two growing period (Figure 1,2).

Soil analysis was made by taking soil samples from four depths of 0–20, 20–40, 40–60, and 60–80 and 80-100 cm and soil physical properties like: texture, dry bulk density, and some soil chemical properties, such as pH, Organic carbon, Total N, available P, available K were taken before and a er the experiment [12].

### Data analysis

Crop data in the rst season, was used to calibrate the CERES maize model of DSSAT so ware version 4.7.5, while in the second season the





Figure 1: The weather condition in the growing period of 2019.



Figure 2: The weather condition in the growing period during 2020.

crop data were used to validate using the statistic root means squared error (RMSE), Normalized di erence RMSE (nRMSE), mean absolute error (MAE), index of agreement (d) of DSSAT (Yang et al., 2014).

en crop growth and yield of maize scenarios were estimated by changing climatic variables that approximate the El-nino periods of historic El nino years, and di erent adaptation strategies were tested based on changes in the length of growing period and maize varieties using the DSSAT so ware.

# **Result and Discussion**

# **Soil Test Results**

e soil analysis across ve depth showed that the dry bulk density increased downward in the range between 2.00 - 2.20, while the soil pH; organic carbon, total nitrogen and available phosphorus decreased

across depth within range between, 5.50-5.30, 2.84-0.62, 0.49-0.12, and 27.0-25.2 respectively, while the available potassium increased downward from 0.5-0.8 (Table 2).

#### **Evaluation of calibrated result**

Calibration of ve maize varieties using the DSSAT So ware: the days to anthesis, the days to physiological maturity, grain yield, harvest index, unit grain weight, and kernels number are simulated with acceptable RSME and d-stat values, higher R2 values and within the ranges of crop coe cient limits.

e minimum and maximum DSSAT-CSM crop coe cients and the new calibrated coe cients of the ve cultivars of maize are shown. Comparison of the cultivars from the genetic coe cient may show that variety SHONE is the highest in grain lling rate, while it

was the second in number of kernels compared to BH545 and MH138,

calibration year. In both calibration and validation year all parameter showed less than 10% di erent except for harvest index in the rst year and LAI and harvest index in the second year.

e nRMSE comparison between Calibrated vs. Validated show that values for validated year are higher in six measured parameters among nine. is can be an indication that better estimation was done during the calibration than validation year hence. Improving high values seen in the harvest index during the calibration year by improving the genetic coe cient could improve the simulation during validation. e R2 and d-stat values are greater than 0.70 for at least four parameter in the calibration year (2019) compared to the validation year (2020) which showed less values for most parameters, exhibiting variability in yield estimation.

# Sensitivity analysis on the planting date and plant spacing

# **Date of Planting**

Changing time of planting known in the area from June 1-15 to a di erent date of planting between May 1 to June 30 showed a change both in phenology and yield of maize. When the planting date is shi ed 1 month before the known time sowing in the site June 1-15 the phenology as well as the yield of maize decreased. Increasing the panting date between May 10 to May 20 the phenology and the yield increased compared to May 1 planting, but without di erence between the two days of planting (May 10 and May 20), however, planting date when shi ed to June 24 the phenology are approximately similar to the known planting date of the site, but the grain yield and dry biomass decreased for both calibrated and validated simulations in 2019 and 2020, data presented here is only validation year. erefore, the time of planting of the area could still be as appropriate using the DSSAT model, but with changes in the climate changes or uctuation the panting date between May 10 to end of May could be bene cial during the occurrence of short rains due to El Nion (Table 8,9).

# **Plant population (plant spacing)**

e plant spacing was made for the corresponding plant population without changing on row spacing, but making change only on plant spacing (Table 10).

Changing the plant spacing (plant desity), from 4.4 plants m-2 to 5.3 and up to 12 plants m-2 consciously increased the grain yield and biomass yield without change on the phenology of maize (days to anthesis, days to physiological maturity). However, increasing the plant density to 14 plants m-2 decreased the grain, but kept increasing the dry biomass yield of maize cultivars, during both evaluation years (2019) and validation year (2020) data presented here is only validation year (Table 11,12) [14].

### Conclutions

## Table 10: Plant population per meter square and its corresponding plant spacing.

Plant population in 1 m <sup>2</sup>	Area in m <sup>2</sup> used per plant	Pant spacing (m x m)			
5.3	0.188	0.75 x 0.25			
6.2	0.161	0.75 x 0.21			
7.4	0.135	0.75 x 0.18			
8	0.125	0.75 x 0.16			
9	0.111	0.75 x 0.15			
12	0.083	0.75 x 0.11			
14	0.071	0.75 x 0.095			

	Population Density (5.3, 6.2, 7.2, 8, 9, 12, 14 plant m <sup>-2</sup> ) maize cultivars										
Cultivar	5.3	9	12	14	5.3	9	12	14			