

# **Original Research Article**

## **Calibration and validation of DSSAT CROPGRO Peanut model for yield and yield attributing characters of groundnut varieties in Northern Agro-Climatic zone of Tamil Nadu**

### **ABSTRACT**

**Aim:** The research study was conducted to calibrate and validate the DSSAT CROPGRO peanut model for simulating the potential yield of groundnut to deciding the best possible management options at major growing areas of Northern Agro-Climatic zone of Tamil Nadu.

**Study Design:** The experiment was conducted in Split plot Design with four Sowing dates and cultivars.

**Methodology:** The DSSAT model requires layer wise soil data (physical and chemical), including soil texture and other soil properties. Daily weather data, including maximum and minimum air temperature (°C), solar radiation ( $\text{MJ m}^{-2} \text{ day}^{-1}$ ) Relative Humidity (%) and precipitation (mm) were used as inputs. Data describing management practices and information about cultivar-specific genetic coefficients were used to calibrate the model. Validation of model were carried out using observed growth and yield attributes of TMV13 and G7 varieties using RMSE, NRMSE and agreement per cent as test criteria for the evaluation.

**Results:** The performance of DSSAT CROPGRO peanut model for simulated growth attributes were underestimated the growth attributes like days to anthesis, leaf area index, days to first pod and days to maturity than compared to observed growth attributes of TMV13 and G7 varieties. But the model performs better for G7 as compared to TMV13. Whereas, yield and yield attributes of CROPGRO peanut model were overestimated than the observed yield.

**Conclusion:** The simulation model shows the low RMSE, NRMSE and high agreement per cent for growth and yield of groundnut which was more than 90 per cent, it shows the higher level of confidence on model simulation with observed characters.

**Key words:** *DSSAT, CROPGRO, Groundnut, Yield, Simulation model*

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## **1. INTRODUCTION**

Groundnut (*Arachis hypogaea* L.) is one of the major oilseed crops grown in subtropical and tropical regions of the world. India occupied the second position in acreage and production of groundnut with an area of 4.73 million hectares and a production of 6.93 million tonnes during

2018-19 (Directorate of Economics and Statistics, 2020). The average productivity was 1465 kg ha<sup>-1</sup>. Seventy per cent of the groundnut area and production is concentrated in the four states viz., Gujarat, Andhra Pradesh, Tamil Nadu and Karnataka. In Tamil Nadu, it is the major oilseed crop grown under rainfed and irrigated condition accounting for 5.7% of the total cropped area. Groundnut is grown in an area of 3.35 lakh hectares with a production of 9.11 lakh tonnes during 2018-19 in the state. The average productivity was 2725 kg ha<sup>-1</sup> (Directorate of Economics and Statistics, 2020). The districts viz., Tiruvannamalai, Vellore, Villupuram, Namakkal, Erode and Salem constituted 54.9% of the area under groundnut in Tamilnadu (Season and Crop report, 2018). Weather is one of the important factor, which affects all stages of groundnut growth and finally the yield (Bandyopadhyay *et al.*, 2005). Owing to variation in monsoon rains, peanut production fluctuates in major growing areas.

Simulation models are useful tools in deciding the best possible management options for optimum growth and yield of any crop against available climatic variables along with soil and water inputs (Parmer *et al.*, 2013 and Yadav *et al.*, 2012). Use of these models are gaining importance particularly against the present climatic change scenario. Improved production technology at the farm level is the most crucial starting point for improving the productivity of groundnut in future by employing and appropriately adapting suitable crop growth simulation models (Anothai *et al.*, 2008). In addition to this, the use of crop yield simulation models comes handy to the government agencies, trade and industry for planning, distribution, storage; processing and export/import of crop produce besides taking timely policy decisions on fixing levy prices provide a reliably accurate advance estimation of yields. Crop simulation models are recent tools that can facilitate identification of production constraints and assist in agrotechnology transfer (Naab *et al.*, 2004; Dangthaisong *et al.*, 2006 and Putto *et al.*, 2009).

Models are widely used as a management tool to evaluate the effects of climate, soil, hydrologic and agronomic factors on crop yield and its variability (Khan *et al.*, 2009). It is not unlikely that use of crop simulation model will definitely improve agronomic weather information and interpretation in time to come and farmers will be able to reduce production risks and increase crop yield by tailoring management decisions to current and expected weather.

The Cropping System Model (CSM)-CROPGRO-Peanut is a process-oriented model that is part of the Decision Support System for Agrotechnology Transfer (DSSAT) (Hooenboom *et al.*, 2010). The model has been evaluated extensively for investigating multiple environmental conditions and evaluating crop yield, cultivars, cropping practices and genetic coefficient (Boote *et al.*, 1985). Accordingly, the validated model can be used to predict growth and yield responses to sowing dates, nutrients, row spacing, and irrigation. The objective of the present

study was to evaluate the performance of the model for peanut crops grown in northern agro-climatic zones of Tamil Nadu and.

## **2. MATERIALS AND METHODS**

The groundnut field experiment for the calibration CROPGRO peanut was conducted in *Kharif* season during July to November 2019 at the farm of Oilseeds Research Station, Tamil Nadu Agricultural University, Tindivanam. The Farm is located in the Northern agro climatic zone of Tamil Nadu at 12.5° N latitude, 79.5° E longitude and at an altitude of 46 m above mean sea level (Figure 1). The growth and yield data collected from the experiments were used to calibrate and validate the DSSAT CROPGRO-Peanut model. In the field experiments, eight treatment combinations, including four planting dates (July 1 and 16, August 1 and 16) and two rulling varieties (TMV 13 and G7) respectively, were used and replicated three times, with plot sizes of 20 square meters. The whole experiment was designed following the split-plot technique with main factor of sowing date and subfactor as groundnut varieties. Groundnut seeds of all the varieties were sown at a depth of 5 cm with 30 cm row-to-row and 10 cm plant-to-plant spacing and all other package of practices were adopted as TNAU crop production guide recommendations.

### **2.1. Model description**

#### **2.1.1. CROPGRO-Peanut model**

The DSSAT CROPGRO-Peanut model v 4.7 was used to study the effect of sowing dates with respect popular cultivars on yield and yield parameters of peanut and to identify a suitable management strategy to cope with possible climate changes in Northern agro climatic zone of Tamil Nadu. The four sowing dates and two varieties combinations were employed to fit the model to changing scenarios. The model requires layer wise soil data (physical and chemical), including soil texture and other soil properties. Daily weather data, including maximum and minimum air temperature (°C), solar radiation ( $\text{MJ m}^{-2} \text{ day}^{-1}$ ) Relative Humidity (%) and precipitation (mm) were used as inputs. Data describing management practices and information about cultivar-specific genetic coefficients were used to calibrate the model. The genotype data file contains genetic coefficient data, namely the genetic coefficient, which describe specific cultivar characteristics of peanut. The CROPGRO-Peanut model uses 18 genetic coefficients to define development and growth characteristics of a peanut cultivar were presented in Table 1.

#### **2.1.2. Model calibration**

The calibration of the CROPGRO-Peanut model was based on yield and yield components viz., pod yield, number of pods and dry matter production data were recorded at the time of harvesting. The genetic coefficients of the peanut cultivars that affect the phenological stages in the CROPGRO models were derived using the “trial-and-error” method of DSSAT v 4.5. Adjustment was performed to match the observed crop phenology and yield with the simulated values and to make the calibrated genetic coefficient lie within the predefined error limits for the cultivar. Following this method, all coefficients were optimized for further simulation (Table 1). For calibration, information for key phenological events (anthesis day, LAI, first pod day, days to maturity), and yield-related data including pod yield, haulm yield and HI were used. The model simulation was accordingly started with the default values available in the model for similar soils of other regions. Details of the experiment, data collection and model calibration are described in Banterng et al. (2003) and Suriharan et al. (2006). Flow chart depicting the methodology of DSSAT CROPGRO Peanut crop simulation model were presented in the Figure 2.

The performance of the CROPGRO-Peanut model was evaluated using the dataset from the field experiments. This approach can be considered as a true validation of the model, as the model parameters were not calibrated on the basis of the field experiment dataset. All the simulated and observed yield and yield components were compared and presented.

## 2.2. Test criteria for evaluation

An analysis of the degree of coincidence between simulated and observed values was carried out by using  $R^2$ , Root Mean Square Error (RMSE), Normalised Root Mean Square Error (NRMSE) and Agreement percent (Jemison *et al.*, 1994).

$$RMSE = \sqrt{1/N \sum (O_i - P_i)^2}$$

$$NRMSE = 100 \times (RMSE / O_i)$$

$$\text{Agreement (\%)} = 100 \times (1 - (RMSE / O_i))$$

Where  $P_i$  and  $O_i$  were the predicted and observed values for the observation, and  $N$  was the number of observations within each treatment. RMSE was measure of the deviation of the simulated from the measured values, and was always positive. A zero value was ideal. The lower the value of RMSE, the higher was the accuracy of the model prediction.

## 3. RESULTS AND DISCUSSION

### 3.1. Growth attributes

### **3.1.1. Days to Anthesis**

The observed days to anthesis for two cultivars like TMV13 and G7 was 30 and 27 days, respectively, whereas model simulated days to anthesis was 27 and 26 days. The test criteria computed by RMSE, NRMSE and Agreement percent for two cultivars are 3.1; 10, 3.70 and 90.00, 96.30 per cent, respectively and presented in Table 2. It suggested that, the model performance was better for G7 variety as compared to TMV7 for simulation of days to anthesis. The findings were conformity with the results of Ujinwal and Patel (2008) and Akula (2003) days to maturity and the model were under estimate the days to anthesis than observed days.

### **3.1.2. Leaf Area Index (LAI)**

The observed LAI of the two varieties like TMVB and G7 was 5.72 and 6.10, respectively whereas DSSAT CROPGRO peanut model simulated LAI was 5.32 and 5.70, respectively for the two varieties. In terms of LAI, G7 variety performed better is compared to TMV13 with respective to the RMSE, NRMSE and Agreement percent (Table 2). Thus, the model under estimate the LAI than compared to observed data. The results of Leaf Area Index of groundnut simulated by CROPGRO peanut model was in Conformity with the finding of Parmer et al. (2013) and Babu (2006).

### **3.1.3. Days to First Pod and maturity**

The days to first pod formation and days to maturity for the two varieties like TMV13 and G7 was underestimated the CROPGRO peanut model than the observed data. The test criteria for two cultivars TMV13 and G7 (Table 2) suggested that model performance was better for G7 as compared to TMV13 variety for days to pod development whereas TMV 13 performed better than G7 for days to maturity. The similar results were found by Pandey et al. (2001) and Deiveegan and Pazhanivelan (2016).

## **3.2. Groundnut pod and haulm yield**

The performance of the model in simulating the pod yield was overestimated than the observed groundnut yield. The pod yield of groundnut for G7 was better performed than TMV13 in CROPGRO peanut model and test criteria of the model was in good confidence level. The test criteria like RMSE, NRMSE and agreement per cent for evaluating pod yield of TMV13 was 264, 8.44 and 90.20, respectively whereas, 121, 3.85 and 96.15, respectively for G7 variety (Table 2). The findings were conformity with the results of Parmer et al. (2013); Soler et al. (2007) and Yadav et al. (2012).

The observed haulm yield of the two varieties like TMVB and G7 was 5892 and 6536, respectively whereas DSSAT CROPGRO peanut model simulated haulm yield was 6389 and 6874 kg/ha, respectively for the two varieties. In terms of haulm yield, G7 variety performed better is compared to TMV13 with respect to the RMSE, NRMSE and Agreement percent (Table 2). Thus, the model overestimates the haulm yield than compared to observed data. The results of haulm yield of groundnut simulated by CROPGRO peanut model was in Conformity with the finding of Babu (2006) and Sabarinathan et al. (2020).

Results showed that the values of simulated harvest index were found to have been underestimated when compared with corresponding observed values. The average errors as computed by RMSE, NRMSE and agreement percent were 0.014, 4.39 and 95.61, respectively for TMV13 whereas 0.018, 5.27 and 94.73, respectively.

#### **4. CONCLUSION**

In the simulation model, Days to anthesis, LAI, days to fist pod development, days to maturity, pod yield, haulm yield and harvest index were satisfactorily simulated by CROPGRO peanut model, pod and haulm yield were overestimated and rest of the parameters under estimated by the model with higher agreement percent of more than 90 with low RMSE and NRMSE values. DSSAT CROPGRO peanut model has proved to be valuable tool and effective model to predicting groundnut yield. Therefore, the validated DSSAT model can be further used for applications such as prediction of crop growth, phenology, water management, potential and actual yields, performance of groundnut under climate variability and change scenarios etc. The model may also be used to improve and evaluate the current practices of groundnut growth management to achieve enhanced groundnut production.

#### **COMPETING INTERESTS DISCLAIMER:**

Authors have declared that no competing interests exist. The products used for this research are commonly and predominantly use products in our area of research and country. There is absolutely no conflict of interest between the authors and producers of the products because we do not intend to use these products as an avenue for any litigation but for the advancement of knowledge. Also, the research was not funded by the producing company rather it was funded by personal efforts of the authors.

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**Table 1. Genetic Co-efficient of TMV13 and G7 for CROPGRO Peanut model**

Sl. No.	Code	Description	TMV13	G7
1	CSDL	Critical Short Day Length below which reproductive development progresses with no daylength effect (for shortday plants) (hour)	11.84	11.84
2	PPSEN	Slope of the relative response of development to photoperiod with time (positive for shortday plants) (1/hour)	0.00	0.00
3	EM-FL	Time between plant emergence and flower appearance (R1) (photothermal days)	18.5	19.3
4	FL-SH	Time between first flower and first pod (R3) (photothermal days)	7.3	7.9
5	FL-SD	Time between first flower and first seed (R5) (photothermal days)	17.5	18.7
6	SD-PM	Time between first seed (R5) and physiological maturity (R7) (photothermal days)	62.20	61.30
7	FL-LF	Time between first flower (R1) and end of leaf expansion (photothermal days)	70.50	69.50
8	LFMAX	Maximum leaf photosynthesis rate at 30 C, 350 vpm CO <sub>2</sub> , and high light (mg CO <sub>2</sub> /m <sup>2</sup> -s)	1.15	1.25
9	SLAVR	Specific leaf area of cultivar under standard growth conditions (cm <sup>2</sup> /g)	250.00	245.00
10	SIZLF	Maximum size of full leaf (three leaflets) (cm <sup>2</sup> )	17.30	16.90
11	XFRT	Maximum fraction of daily growth that is partitioned to seed + shell	0.75	0.70
12	WTPSD	Maximum weight per seed (g)	0.371	0.410
13	SFDUR	Seed filling duration for pod cohort at standard growth conditions (photothermal days)	33.00	33.00
14	SDPDV	Average seed per pod under standard growing conditions (#/pod)	1.65	1.65
15	PODUR	Time required for cultivar to reach final pod load under optimal conditions (photothermal days)	19.20	19.00
16	THRSH	The maximum ratio of (seed/(seed+shell)) at maturity.	74.00	74.00
17	SDPRO	Fraction protein in seeds (g(protein)/g(seed))	0.270	0.270
18	SDLIP	Fraction oil in seeds (g(oil)/g(seed))	0.514	0.480



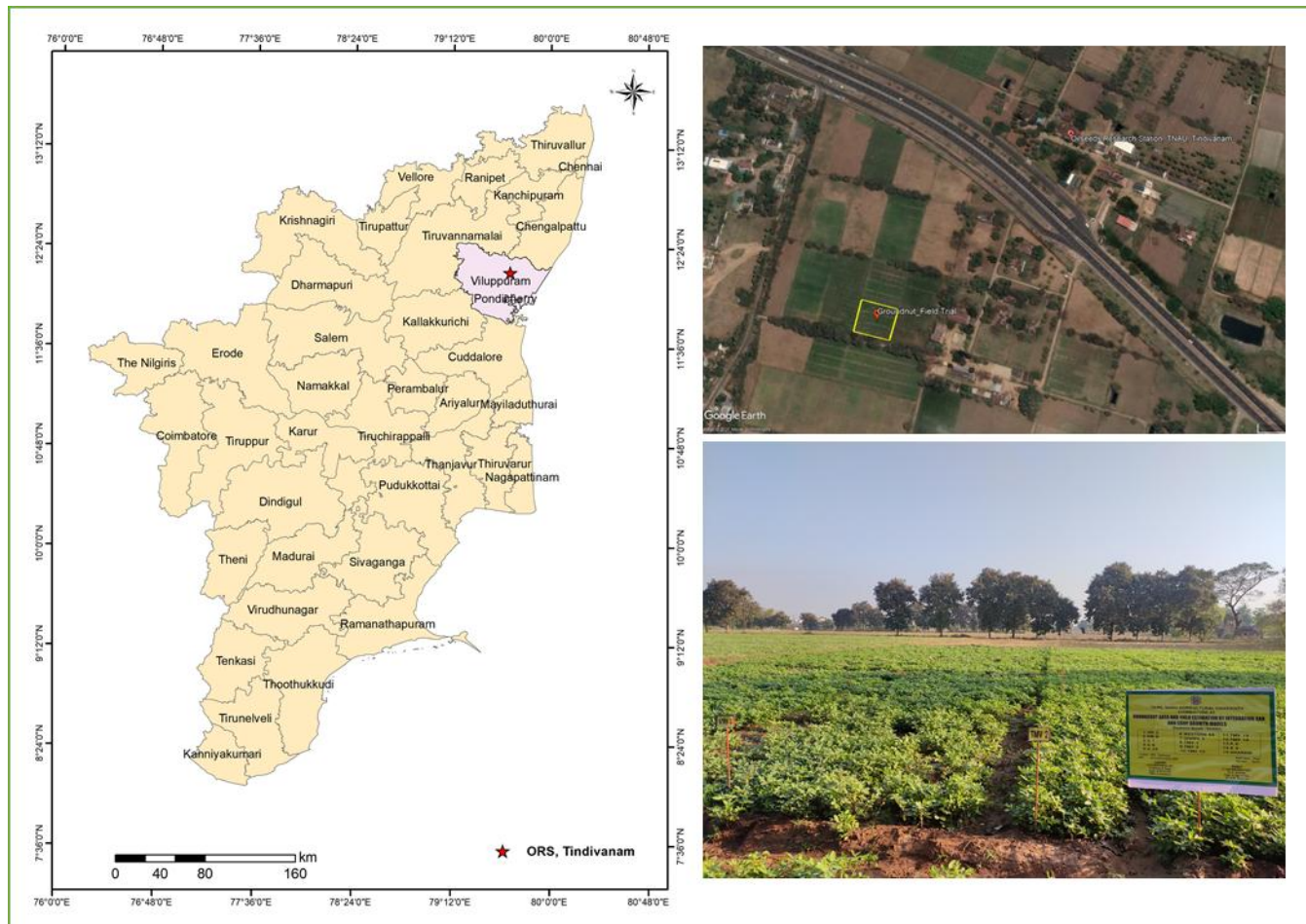


Figure 1. Location Map of field experiment

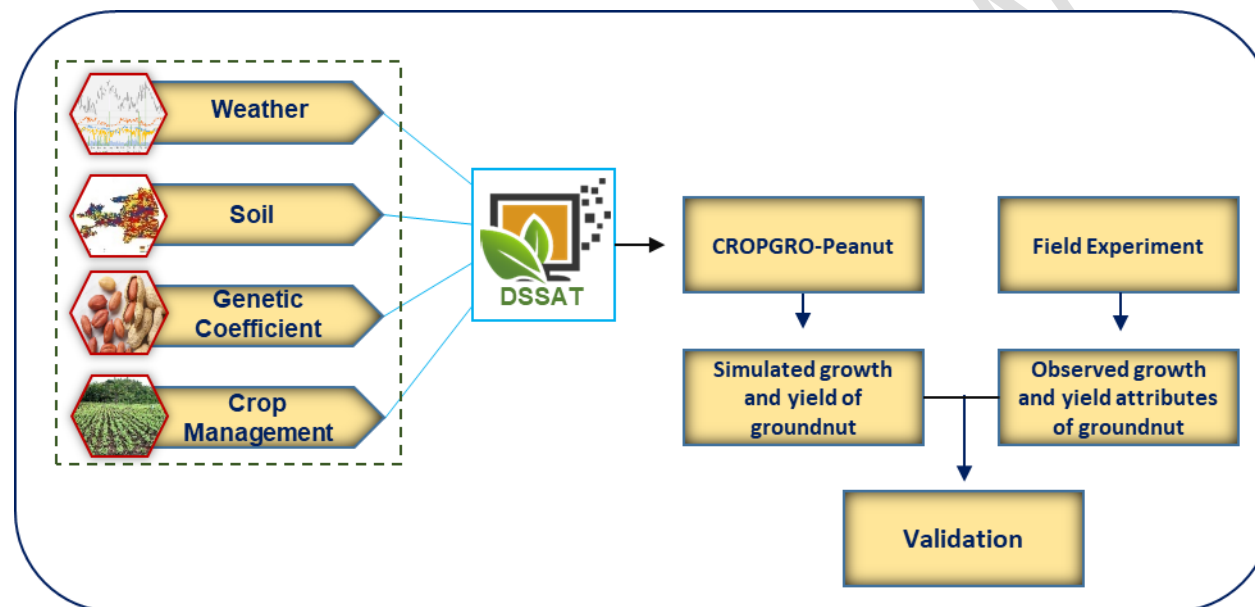


Figure 2. Flow chart depicting the methodology of DSSAT CROPGRO Peanut crop simulation model