Assessment of Climate Change Impact on Crop yield and Evaluation of Coping Strategy using Crop Growth Simulation Model

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Abstract: Any change in climate will have implications for climate-sensitive systems such as agriculture, forestry and some other natural resources. Changes in solar radiation, temperature and precipitation will produce changes in crop yields and hence economics of agriculture. It is possible to understand the phenomenon of climate change on crop production and to develop adaptation strategies for sustainability in food production, using a suitable crop simulation model. CERES-Maize model of DSSAT v4.0 was used to simulate the maize yield of the region under climate change scenarios using the historical weather data at Kharagpur (1977-2007), Damdam (1974-2003) and Purulia (1986-2000), West Bengal, India. The model was calibrated using the crop experimental data, climate data and soil data for two years (1996-1997) and was validated by using the data of the year 1998 at Kharagpur. The change in values of weather parameters due to climate change and its effects on the maize crop growth and yield was studied. It was observed that increase in mean temperature and leaf area index have negative impacts on maize yield. When the maximum leaf area index increased, the grain yield was found to be decreased. Increase in CO₂ concentration with each degree incremental temperature decreased the grain yield but increase in CO₂ concentration with fixed temperature increased the maize yield. Adjustments were made in the date of sowing to investigate suitable option for adaptation under the future climate change scenarios. Highest yield was obtained when the sowing date was advanced by a week at Kharagpur and Damdam whereas for Purulia, the experimental date of sowing was found to be beneficial.

Keywords: Climate change, CERES-Maize, DSSAT v4.0, crop simulation, yield.

1. Introduction

Maize is the third most important food crop in the world among global food crops in terms of production, following rice and wheat and is grown in various climates throughout the world. In India, it is grown over an area of 6.4 million hectares with total production of about 11.5 million...
India, it is grown over an area of 6.4 million hectares with total production of about 11.5 million tonnes. Based on climate records, average global temperatures at the earth’s surface are rising. Since global records began in the mid 19th century, the five warmest years have occurred during the 1990s and 10 of the 11 warmest years have occurred since 1980 [30]. Climate is changing mainly due to increasing concentration of greenhouse gases and is affecting many economic sectors, mainly agriculture and forestry. A change in climatic variables is predicted to increase the earth’s mean surface temperature and is likely to be accompanied by increased precipitation [8]. Field crop production is significantly affected by climatic variables because photosynthetically active radiation, air temperature and water are the driving forces for crop growth [34, 33]. It is now well recognized that crop production is very sensitive to climate change with varying effects according to region [27]. In the Indian context, Kumar and Parikh [20] estimated the macro level impacts of climate change using a distinct approach. They showed that under doubled carbon dioxide concentration levels in the later half of twenty-first century, the gross domestic product would decline by 1.4 to 3% points due to climate change. Increasing CO₂ level increases crop production due to higher rates of photosynthesis and increased water use efficiency, especially at low water or high nutrient availability [24]. Mall et al. [26] provided an excellent review of climate change impact studies on Indian agriculture, mainly from the perspective of physical impact.

The Decision Support System for Agro-technology Transfer (DSSAT), which is a combination of several dynamic crop simulation models, is one such model that can predict accurately the growth, development and yield of crops with the help of soil, daily weather and management inputs, to aid farmers in developing long-term rotational strategies. DSSAT is the major product of the IBSNAT (International Benchmark Site Network for Agrotechnology Transfer) project, initiated in 1982 [37]. Crop simulation models (CSMs) have been used extensively in India to quantify the magnitude of improvement in yield potential at different levels of management and climatic variability and proved that the simulation studies can be used in supplement to field studies [1]. Oosterom et al. [29] used crop simulation models as agronomic tools to assess the growth and development, biomass accumulation and yield of crops. Steiner et al. [36] used CERES crop growth simulation model for maize, sorghum and winter wheat grown at Bushland, Texas to predict evapotranspiration and crop growth parameters like leaf area index and total dry matter accumulation under water stress. The CERES-Maize model includes the capability to simulate the effects of increased atmospheric CO₂ concentrations on photosynthesis and water use by the crop. Daily potential transpiration calculations are modified by the CO₂ concentrations [22, 10, 31]. Wafula [38] evaluated CERES-Maize under different management strategies and tested the yield potential under different sowing dates, planting density and rates of fertilizer N application. Bannayan et al. [5] used CERES-Maize model to check the photothermal impact on maize performance and found that the model is suitable as a modeling tool in simulation studies to examine the effects of the genotype by environment interaction on growth, development and yield. White et al. [40] tested the CERES models in response to temperature and evaluated the models for predicting the growth, biomass partitioning and yield of wheat and maize crops. In order to overcome the predicted limitations for crop production in the future, there is also a need to identify and evaluate the suitable agronomic practices such as altered sowing date and
selection of improved varieties with increased production at high temperature and other useful traits. These adaptation strategies may potentially lessen future yield losses from climate change or may improve yields in regions where beneficial climate changes occur [15]. Keeping this in view, the study was conducted at IIT Kharagpur during 2008/09 with the objectives: (i) to simulate the effect of climate change on maize growth and yield parameters and (ii) to evaluate optimum planting date for maize using CERES-Maize model.

2. Materials and Methods

2.1 Site Description

The current study was carried out at Agricultural and Food Engineering Department, Indian Institute of Technology, Kharagpur, India (22°19' N latitude and 87°19' E longitude and 48 m above the mean sea level). The soil of this region is of lateritic type with sandy loam texture, which is taxonomically grouped under the group ‘Alfisol’. The climate of Kharagpur is classified as sub-humid, subtropical. The site receives an average rainfall of 1200 mm with 70-75% of the total rainfall in the monsoon during June to October. The average temperature varies between 21 °C and 32 ºC.

The study was extended to Dumdum (latitude 22.38° N, longitude 88.38° E) and Purulia (latitude 23.2° N, longitude 88.28 ° E) to assess the impact of climate change on maize growth and yield using the calibrated crop growth model at Kharagpur. Only climatic and soil data were used to run the model at Dumdum and Purulia. Soil properties at Dumdum and Purulia were obtained from National Bureau of Soil Survey and Land Use Planning (NBSSLUP), Kolkata.

2.2 Experimental Crop Data

The experimental crop data required for calibration and validation of the CERES-Maize model were collected from previous experiments conducted for the period of 1996-98 at the experimental farm of the Agricultural and Food Engineering Department, Indian Institute of Technology, Kharagpur, India. The local high yielding maize variety namely “Vijaya Composite” was considered as the experimental crop in the study. The field experiments were designed as per randomized block design (RBD) with irrigation schedules or treatments as the factors. There were four irrigation treatments in all the three experiments. There were three replications for each treatment. A standard seed rate of 17–20 kg/ha was used. The seeds were sown at a row spacing of 60 cm and a plant spacing of 25 cm during all the three experiments. The fertilizer dose of N:P:K was 100:60:40 kg/ha. The irrigation treatments consisted of irrigation scheduling based on maximum allowable depletion (MAD) of the total available soil water (ASW) criteria. Each irrigation treatment was based on a predefined level of MAD, which was a fixed percent of the total ASW. Irrigation water was applied whenever the threshold value of MAD for the particular irrigation treatment was attained. The irrigation treatments considered in the study were: $T_1 = 10\%$ MAD of ASW, $T_2 = 30\%$ MAD of ASW, $T_3 = 45\%$ MAD of ASW, $T_4 = 60\%$ MAD of ASW.
2.3 Weather Data

Daily values of the weather variables such as: solar radiation, maximum and minimum temperature and rainfall for the experimental period were obtained from an automatic weather station installed close to the experimental crop field. Thirty one years weather data (1977-2007) of Kharagpur were collected for use in the CERES-Maize model. The climatic data for Dumdum (1974-2003) and Purulia (1986-2000) were collected from the India Meteorological Department, Pune.

2.4 Analysis of Trend

Trend is a long term variation in a time series. It tells whether a particular data set is increasing or decreasing over the period of time. A time series is a sequence of observations which are ordered in time or space. The Mann-Kendall non-parametric test was used to analyze the trend of variation in climatic parameters such as: solar radiation, maximum temperature and minimum temperature at Kharagpur, Dumdum and Purulia. The Mann-Kendall test is considered as the most appropriate approach for analysis of trend in climatological time series data [12, 11, 7]. Brief description about the Mann-Kendall test is given below.

Let \( x_1, x_2, \ldots, x_n \) represents \( n \) data points where \( x_j \) represents the data points at time \( j \). Then the Mann-Kendall statistic (S) is given by

\[
S = \sum_{k=1}^{n-1} \sum_{j=k+1}^{n} \text{sign}(x_i - x_j)
\]  

(1)

where

\[
\text{sign}(x_i - x_j) = \begin{cases} 
1 & \text{if } (x_i - x_j) > 0 \\
0 & \text{if } (x_i - x_j) = 0 \\
-1 & \text{if } (x_i - x_j) < 0 
\end{cases}
\]  

(2)

Variance of S, \( \text{VAR}(S) \) is calculated by the following equation,

\[
\text{VAR}(S) = \frac{1}{18} \left[ n(n-1)(2n+5) - \sum_{p=1}^{g} t_p(t_p-1)(2t_p+5) \right]
\]  

(3)

where \( n \) is the number of data points, \( g \) is the number of tied group and \( t_p \) is the number of data points in the \( p^{th} \) group.
Normalized test statistic (Z) is calculated as follows,

\[ Z = \begin{cases} \frac{S - 1}{\text{VAR}(S)} & \text{if } (S > 0) \\ 0 & \text{if } (S = 0) \\ \frac{S + 1}{\text{VAR}(S)} & \text{if } (S < 0) \end{cases} \]  

(4)

The probability associated with this normalized test statistic is calculated as follows,

\[ f(z) = \frac{1}{\sqrt{2\pi}} e^{-\frac{z^2}{2}} \]  

(5)

2.5 CERES-Maize Model

DSSAT (Decision Support System for Agrotechnology Transfer) was used for simulating crop growth parameters [13]. The model was developed under the International Benchmark Sites Network for Agrotechnology Transfer (IBSNAT) project. DSSAT is integrated software of different computer programmes, which can facilitate the application of crop simulation models in research and decision-making. For the present study CERES model from DSSAT v4.0 was used for maize crop. In CERES model, the input and output files carry data related to soil, weather, genotype and crop. The file structure can be described with three different categories: input files, output files and experimental performance data files. Input files are further divided into those dealing with the experiment, weather and soil, and the characteristics of different genotypes (crop and cultivar). The files can be recognized by the extension used e.g. ‘.wth’ for weather file, ‘.sol’ for soil file, ‘.exp’ for experimental file and ‘.cul’ for cultivar file. The input and output file related to soil, weather and cultivar were developed and used in this study. Crop growth is simulated by employing a carbon balance approach in a source-sink system [32]. Daily crop growth rate is calculated as:

\[ PCARB = \frac{\text{RUE} \times \text{PAR} \times (1 - e^{-K \times \text{LAI}})}{\text{PLTOP}} \times \text{CO}_2 \]  

(6)

where PCARB = Potential growth rate, g/plant, RUE = Radiation use efficiency (g dry matter/MJ PAR), PAR = Photo synthetically active radiation (MJ/m²), PLTOP = Plant population, plant/m², K= Light extinction factor, LAI= Green leaf area index, CO₂= CO₂ modification factor.

2.6 Simulating the Impact of Temperature and CO₂

Simulation of the duration of each phenological stage uses the concept of thermal time [32]. Because the time scale of plants is closely coupled with its thermal environment, thinking of thermal time as a plant’s view of time is appropriate. Thermal time has units of °C day. The simplest and most useful definition of thermal time \( t_d \) is
where $T_a$ is daily mean air temperature, $T_b$ is the base temperature at which development stops and $n$ is the number of days of temperature observations used in the simulation. The calculation of $T_a$ is accomplished in the CERES models by averaging the daily maximum and minimum temperatures under most circumstances.

### 2.7 Fixed Climate Change Scenarios

Changes in Earth’s climate have been projected by the end of this century because some atmospheric ‘greenhouse’ gases, among them carbon dioxide (CO$_2$), are increasing [2, 14]. It is expected that atmospheric CO$_2$ concentration will double sometime during this century if fossil fuels burning continues and air temperature is predicted to rise 1.5 to 5 °C with more than 90% likelihood by 2100 [3, 25]. As a consequence of a possible increase in atmospheric CO$_2$ concentration and associated climate changes, several studies have been conducted in order to predict the effects of climate change on crop growth, development, and yield. A hypothetical study was done to determine the potential yield of maize under 30 different combinations of CO$_2$ and temperature, including the fixed increment in CO$_2$ (380, 400, 500, 600, 700ppm) and temperature (ambient, +1, +2, +3, +4 and +5°C) individually and with all combinations of these levels of CO$_2$ and temperature. For simulating effect of temperature change the daily maximum and minimum temperatures were increased by 1-5 °C individually in CERES-Maize model and yield was calculated accordingly. Similarly climate change scenarios under different levels of CO$_2$ were applied by changing ambient CO$_2$ in CERES-Maize.

### 2.8 Sowing Date Evaluation

The various possible agro adaptation measures include alternate sowing dates, water management, different tillage depths, nutrient management, improved heat resistant varieties etc. Among these alternate sowing date was applied to the calibrated CERES-Maize model and their impact on the crop yield was determined. The maize yield was simulated for 4 weeks before and 4 weeks after the actual date of sowing (6 February) at the interval of one week. The percentage change in the yield compared with yield at conditions similar during base period was considered to determine the adaptation measures which will mitigate the negative effects of climate change.

### 3. Results and Discussion

The model was calibrated using the daily weather data and experimental-crop data on grain yield, above ground dry matter and maximum leaf area index. The well-watered treatment (10% Maximum Allowable Depletion) of each experiment was selected for calibration. The values of genetic coefficients were estimated using the best-fit method. Model was calibrated for each experiment separately and an average value of each genetic coefficient was considered for simulation. The model was calibrated using two years of field experiment in 1996 and 1997. Then the model was validated using the crop data of the year 1998. The genetic coefficients
determined by the process of calibration were used for validation. Statistical analysis was performed to evaluate the performance of CERES-Maize model in simulating the crop variables. The values of statistical test parameters such as: mean error (ME), root mean square error (RMSE) and model efficiency (EF) revealed that the CERES-Maize Model simulated the crop variables with considerable accuracy and thereby can be recommended for further use under similar agro-climatic conditions (Table 1).

**Table 1:** Statistics for comparison amongst simulated and measured values of crop parameters

| Crop parameters                      | $R^2$ | RMSE | ME  | EF   |
|--------------------------------------|-------|------|-----|------|
| Grain yield (kg/ha)                  | 0.82  | 510  | 233 | 0.807|
| Above ground dry matter (kg/ha)      | 0.85  | 530  | 279 | 0.817|
| Leaf area index                      | 0.86  | 0.138| 0.074| 0.84 |

3.1 Analysis of Trend

Trend of variation in climatic data was analyzed using the Mann-Kendall method, which is a non-parametric test for identifying trends in time series data. The test compares the relative magnitude of sample data. Solar radiation, maximum temperature and minimum temperature during the crop growth period (February to May) at Kharagpur (1977-2007), Dumdum (1974-2003) and Purulia (1986-2000) were analyzed to check their trend. The data values were evaluated as an order time series and each data value was compared with all subsequent data values.

**Table 2:** Mann-Kendall trend results for solar radiation, maximum temperature and minimum temperature at Kharagpur, Dumdum and Purulia

| Weather Parameters | Mann-Kendall statistic (S) | Normalized test statistic (Z) | Probability | Trend* |
|--------------------|----------------------------|------------------------------|-------------|--------|
|                    | Kharagpur                  |                              |             |        |
| Solar radiation    | 85                         | 1.43                         | 0.86        | No Trend |
| Maximum temperature| 58                         | 0.97                         | 0.78        | No Trend |
| Minimum temperature | -45                        | -0.75                        | 0.70        | No Trend |
|                    | Dumdum                     |                              |             |        |
| Solar radiation    | 142                        | 2.56                         | 1           | Increasing |
| Maximum temperature| -38                        | -0.67                        | 0.68        | No Trend |
| Minimum temperature | -126                       | -2.25                        | 0.99        | Decreasing |
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|         | Minimum temperature | -126                       | -2.25                        | 0.99        | Decreasing |
| Purulia | Solar radiation | -5                         | -0.19                        | 0.65        | No Trend |
|         | Maximum temperature | 24                         | 1.13                         | 0.77        | No Trend |
|         | Minimum temperature | -27                        | -1.28                        | 0.81        | No Trend |

* At 5% level of significance.

The Mann-Kendall statistic (S), variance (VAR(S)), normalized test statistic (Z) and probability associated with this normalized test statistic were calculated by using the method described in section 2.3. Microsoft Excel function NORMSDIST, was used to calculate the probability associated with normalized test statistics. At a particular probability level of significance (95%), the trend is said to be decreasing if normalized test statistic (Z) is negative and the computed probability is greater than the significance level. The trend is said to be increasing if the normalized test statistic (Z) is positive and the computed probability is greater than level of significance and there is no trend if the computed probability is less than the level of significance [16]. The results of Mann-Kendall test on solar radiation, maximum temperature and minimum temperature are summarized in Table 2 for Kharagpur, Dumdum and Purulia. The trend analysis at 5% of level of significance showed that there is no trend for solar radiation, maximum temperature and minimum temperature at Kharagpur and Purulia. There is increasing trend for solar radiation, decreasing trend for minimum temperature and no trend for maximum temperature at Dumdum.

#### 3.2 Effect of Temperature on Yield of Maize

Temperature exerts a major effect on the rate at which plants develop and growth can be retarded when the temperature is either too low or too high [28]. For a given minimum temperature, increasing maximum temperature up to 35 °C accelerate both development towards anthesis and maturity. An increase in temperature above the optimum temperature results in a reduction of the developmental rate [6]. There is a strong relationship between crop development and the mean daily temperature [39]. In general, higher temperatures during the growing season will be associated with higher radiation and a demand for more water, which along with elevated CO$_2$ concentration are major factors that have to be considered in any impact assessment. As shown in Fig. 1, rising mean temperatures can lead to decrease in maize yield as revealed during the years 1979, 1986, 1991 and 1999 which supports the earlier findings of Schlenker and Roberts [35]. Similar trends of decreased yield with increased mean temperature were found at Dumdum (Fig. 2) and Purulia (Fig. 3).
Fig. 1: Variation in maize yield with mean temperature at Kharagpur for period 1977-2007

Fig. 2: Variation in maize yield with mean temperature at Dumdum for period 1974-2003

A negative yield response to diurnal temperature range (DTR; the difference between daily maximum and minimum temperature), coupled with a negative yield response to average temperature in most regions, indicates that temperature increase is more harmful during day than at night.

Fig. 3: Variation in maize yield with mean temperature at Purulia for period 1986-2000
In studies with the EPIC crop simulation model, Dhakhwa and Campbell [9] concluded that DTR increase resulted in lower maize yields in the US because of greater evapotranspiration losses, and consequent water stress. A recent study of US maize yields that utilized daily minimum and maximum temperature data provided strong empirical evidence that yield decreased non-linearly with temperatures above 25°C, with even short periods above 30°C resulting in significant yield losses [35]. In general, higher temperatures during growing season will be associated with higher radiation and a demand for more water, which along with elevated CO₂ concentration are major factors that have to be considered in any impact assessment. Yield–temperature response curves (Fig. 4) show that there is a decreasing trend in grain yield of maize per degree rise in seasonal mean temperature.

### 3.3 Effect of Leaf Area Index on Growth and Yield of Maize

Plant leaf area has an important influence on light interception and dry matter production. The rate of leaf area expansion is an important component of plant growth that is quite sensitive to environmental stress. If the maximum LAI increases, biomass production increases due to increased rate of photosynthesis leading to less accumulation in the form of grain yield due to low partitioning. Historical weather data was used for simulation of crop growth and yield at Kharagpur, Dumdum and Purulia using the crop growth simulation model. It was found from the simulation that in most of the years, when maximum leaf area index (LAI) increased, crop yield decreased. On the other hand, crop yield increased with the decrease in maximum leaf area index. At Kharagpur, in the year 1979 (Fig. 5), maximum LAI was found to be 3.28 and crop yield was 1852 kg/ha, where as in 1981, maximum LAI was found to be 2.5 with crop yield as 3940 kg/ha. This shows that the LAI has significant influence on maize yield at Kharagpur. The effect of leaf area index on the yield of maize at Dumdum (Fig. 6) and Purulia (Fig. 7) was found to follow a trend that when the leaf area index was higher the maize yield was lower and vice versa as shown in Fig. 8.
Leaf growth is more sensitive to plant water deficits than photosynthesis. Lower temperature or moderate drought stresses reduce the expansion growth more than that if photosynthesis is reduced; causing increase in specific leaf weight and increasing the proportion of assimilate partitioned to the roots. Maximum foliar development retards grain development, as a result, crop yield decreases with the increase in maximum LAI.
3.4 Effect of Temperature and CO\textsubscript{2} Levels at Fixed Increments on Maize Yield

Carbon dioxide is considered a greenhouse gas due to its high absorptance in several wavelengths of the thermal infrared radiation emitted by earth’s surface. The greater the content of gases in the atmosphere that absorb thermal infrared radiation emitted from the earth surface, the greater the thermal infrared radiation emitted by the atmosphere towards the earth surface. Consequently, the long wave balance of the surface will be less negative and more energy will be available for latent and sensible heat fluxes at the earth’s surface. As more energy is available for sensible heat flux, air temperature is expected to rise. If the increase in atmospheric CO\textsubscript{2} concentration is accompanied by an increase in air temperature, crops may shorten their growth cycle, which may offset the advantages of an increasing CO\textsubscript{2} concentration. Therefore, the interacting effects of CO\textsubscript{2} concentration and temperature on plant growth are complicated. CO\textsubscript{2} is a component of plant photosynthesis and therefore influences biomass production. It also regulates the opening of plant stomata and therefore affects plant transpiration. As a result, plants growing in increased CO\textsubscript{2} conditions will produce more biomass and will consume less water.
The potential yield of maize was simulated at Kharagpur using the crop growth model under the combinations of CO\textsubscript{2} and temperature with fixed increments in CO\textsubscript{2} concentrations (380, 400, 500, 600 and 700 ppm) and temperatures (ambient, +1, +2, +3, +4, +5 °C) individually and with all combinations of CO\textsubscript{2} and temperature for the year 1996. At all the CO\textsubscript{2} levels tested, the simulated maize yield decreased due to an increase in temperature. On the other hand for an increase in CO\textsubscript{2} level at any particular temperature, maize yields found to be increased (Fig. 9). Similar trends of crop yield were obtained by [17, 18, 19]. He estimated that a doubling of CO\textsubscript{2} concentration, holding other factors constant, could lead to a 34±6% increase in agricultural yields of C3 plants. Similarly, Lawlor and Mitchell [23] found that under availability of adequate water, nutrients and pest control, yield of C3 crops growing in about 700 ppm CO\textsubscript{2} would be about 30 to 40% greater than the present yield.

### 3.5 Effect of Change in Sowing Date on Maize Yield

Different studies have suggested that adjusting the sowing dates and determining the optimum dates for sowing will be helpful in reducing the effect of climate change [4, 21]. Adjustment of management practices may help to offset any detrimental effects of climate change on maize production. Probably the easiest adaptation strategy to cope with climate change is to adjust the sowing dates. The selection of an earlier sowing date for maize will probably be the appropriate response to offset the negative effect of a potential increase in temperature. This change in planting date will allow for the crop to develop during a period of the year with lower temperatures, thereby decreasing developmental rates and increasing the growth duration, especially the grain filling period. Adjustment of sowing date and simulation of its effect was done using the model to investigate a suitable agronomic option for adaptation under the future climate change scenarios.

The potential outcome of adjusting the sowing dates at Kharagpur, Dumdum and Purulia was simulated using the CERES-maize model for the year 1996 to assess the effect of climate change, as shown in Fig. 10. To determine the optimum sowing dates the potential outcomes were studied by shifting the sowing dates 3 weeks before and 4 weeks after the actual date of sowing (6\textsuperscript{th} February) at an interval of one week. The maize yield was simulated for each date and the percentage change in yield corresponding to the experimental yield for actual sowing date was calculated. Among the different sowing dates considered, the sowing date of 30\textsuperscript{th} January, that is one week before the actual sowing date was considered to be the most beneficial for Kharagpur and Dumdum in terms of increase in yield. For Purulia, 6\textsuperscript{th} February was found to be the most appropriate date of sowing. When the date of sowing was shifted a week ahead with respect to actual, an increase of 25% and 12% in yield was observed for Dumdum and Kharagpur respectively. It should be noted, however, that although changes in sowing date are a no-cost decision that can be taken at the farm-level, a large shift in sowing dates probably would interfere with the agrotechnological management of other crops, grown during the remainder of the year.
Climate change is one of the major challenges being faced by the world in coming decades. It is affecting almost every sector including food and water and is becoming a threat to global food security. Maize being the third most important food crop in the world requires special attention and immediate mitigation measures to achieve food security under changed climate scenarios. The objectives of this study were to simulate the maize yield under various climate change scenarios and to identify the possible adaptation measures to reduce the negative impact of climate change on yield using CERES-Maize model. The study was conducted using secondary data collected on crop management, climate and soil. The CERES-Maize model was calibrated and validated using the experimental data for maize at Kharagpur during the three consecutive years from 1996 to 1998. The maize yield was simulated using the historical climate data at Kharagpur (1977-2007), Dumdum (1974-2003) and Purulia (1986-2000) with the help of calibrated model. The model was used to study the impacts of temperature, solar radiation and leaf area index on growth and yield of maize at the three locations. The effect of elevated temperature and CO$_2$ on maize yield was also studied. A study was done to assess the effect of developed scenarios on maize yield. The model was also used to assess the suitable date of sowing, suitable plant density and appropriate nitrogen dose in the climate change scenarios. Based on the results of the study, the conclusions are: (i) There is no trend for solar radiation, maximum temperature and minimum temperature at Kharagpur and Purulia but there is increasing trend for solar radiation, no trend for maximum temperature and decreasing trend for minimum temperature at Dumdum, (ii) Calibrated CERES-Maize model can be used for the simulation of maize yield in this subtropical climate region, (iii) Increase in mean temperature and maximum leaf area index have a negative impact on maize yield, (iv) Increase in temperature affects maize yield negatively while increase in CO$_2$ concentration has positive effect on maize yield and (v) Selection of alternate sowing dates may help to mitigate the negative effects of climate change. Shifting the sowing date a week ahead than the experimental date (6$^{th}$ Feb) is found to be beneficial for Kharagpur and Dumdum whereas for Purulia, the experimental date of sowing is beneficial.
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