Use of CERES-wheat model for predicting wheat yields of Ninital District (U.P.) India

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ABSTRACT

CERES-Wheat model was investigated for its ability to simulate weather induced year-to-year yield variability (1980-1994) and to provide wheat yield prediction (1994-1998) over Nainital district (U.P.). Wheat yield of the district has increased from 1180 to 2912 kg hard between 1970 and 1998. Published genetic coefficients, dominant soil hydrological properties and daily weather measurements at Pantnagar were inputs for simulation. Yield deviations (YDOBS) from segmented linear technology trends (YOBS)(Period I: 1970-1987, Period II: 1988-1994) were used along with year-wise deviations in simulated yields (YDSIM) from its fifteen-year trend (YSIM) (1980-1994) for yield prediction. Deviations of yearly-simulated wheat yields during 1980-1994 could explain about 33 percent of variation in observed yield from the technology trend, while segmented technology trends combindly explained about 58% variability in observed wheat yield. Comparison of two approaches of yield prediction for the period 1995-1998 namely direct use of YDSIM and using relation between YDSIM and YDOBS gave RMSE of 163.3 kg har and 212.2 kg ha⁻¹, respectively.

Key words: CERES-Wheat, DSSAT 3.5, Simulation model

Weather plays a dominant role in determining year-to-year yield variability in the crops. A number of statistical techniques such as multiple regression, principal components (Jain et al. 1984) and Markov chain (Ramasubramanian & Jain, 1999) and agro-meteorological models (Walker, 1989) have been used to quantify the response of crops to weather. These techniques are

limited by availability of only short period historical data series; limited weather variables at specific periods are used for yield prediction, while change in variety and cultural conditions, non-linear and interaction effects among weather variables are generally ignored.

Several crop simulation models have been developed based on physical

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relationship between crop growth and weather input interaction (Whisler et al., 1986). These require crop specific parameters, soil characteristics, cultural practices and daily weather inputs to simulate crop growth and yield. These models are generally calibrated and validated using observations from experimental plots and can mimic crop response to various inputs at plot level (Kenneth et al., 1996). Currently simulated models are being investigated for regional yield prediction (Supit, 1997) as well as for projecting crop response to future climate (Lal et al., 1999).

In this study, CERES-Wheat (Ritche and Otter, 1985) model has been used to simulate crop growth and yield over a period of 19 years from 1980 to 1998 for Nainital district. Using an approach for relating simulated yield deviation to regional yield variation (1980-1994), yield forecasts were made for the period of 1995 to 1998.

MATERIALS AND METHODS

Model description

The CERES-Wheat (Crop Estimation through Resource and Environment Synthesis) model was developed in the U.S.A. as a general simulation tool for the wheat crop (Godwin et al. 1990) and has now been included in DSSAT (Decision Supporting System for Agro technology Transfer, Hunt et al., 1994) version-3.5. It employs canopy photosynthesis, water uptake and evapotranspiration modules to compute dry

matter production and its partition along with temperature and photosynthesis based phenology and phylochron development. The model has been successfully tested and calibrated in a multi-year experiment for wheat cultivation at Ludhiana (Hundal & Kaur, 1997) in India.

Inputs

Daily weather data were collected from the Agro-meteorological observatory of Govind Ballabh Pant University of Agriculture and Technology (G.B.P.U.A. & T.), Pantnagar, district Udham Singh Nagar (Uttar Pradesh). Solar radiation was estimated from recorded sunshine hours using Angstrom relationship with coefficients a and b. 0.256 and 0.448, respectively (Mani and Rangarajan 1982). Crop genetic coefficients of the popular wheat variety HD-2329 published by Hundal and Kaur. (1997) were adapted for use. The PHINT (Phylochron interval; the interval in thermal time (degree days) between successive leaf tip appearances) value for the model has been used as 75 (Table 1). The average crop yields (1970-1998) for the Nainital district were obtained from State Agricultural Department.

Soil input data, which included soil water availability, bulk density, total carbon in the soil profile and soil pH were obtained from published information (Murthy et al., 1982 and Arya et al., 1984). The initial water status of the soil in each year was assumed at drained upper limit (0.03 Mpa) in a fallow field.

Table-1: Genetic coefficients used in model (Adapted from Hundal and Kaur, 1997)

Parameter	Description of parameter	Coefficient			
PIV	Relative amount that development is slowed for, each day of unfulfilled vernalization, assuming that 50 days of vernalization is sufficient for all cultivars. Relative amount that development is slowed when plants are grown in a photoperiod 1 hour shorter than optimum (which is considered to be 20 hours). Relative grain filling duration based on thermal time (degree days above a base temperature of 10°C), Where each unit increase above zero adds 20 degree days to an initial value of 430 degree days.				
PID					
P5					
G1	Kernel numbers per unit weight of stem (less leaf blades and sheaths) plus spike at anthesis (1g ⁻¹).	2.5			
G2	Kernel filling rate under optimum conditions (mg day-).	2.9			
G3	Non-stressed dry weight of a single stem (excluding leaf blades and sheaths) and spike when elongation ceases (g).	4.0			
PHINT	Phylochron interval; the interval in thermal time between successive leaf tip appearances.	75			

Crop management data related to agronomic mactices such as sowing, fertilizer amount and application date. irrigation quantity and time and harvesting date and method were kept constant at optimal recommended guideline for the region given by G.B.P.U.A. & T. Pantnagar. The year-to-year variability in sowing dates was generated by using auto sowing option providing the sowing limits as Nov 15 to Dec 15, in DSSAT 3.5 modelling environment. Nitrogen at 120 kg har and Phosphorous at 40 kg hard were applied (half of nitrogen and full dose of phosphorous at the time of sowing and remaining amount of nitrogen as top dressing, 50 days after sowing). Three irrigations of 60 mm each at Crown Root Initiation, Jointing and Dough stage were prescribed in simulation.

Yield simulation and historical yield data analysis

A consistent wheat yield series (YOBS) of Nainital for period 1970 to 1998 was generated by combining acreage and production of Udham Singh Nagar and New Nainital for the period 1996-1998 with earlier yield series of erstwhile Nainital district (Year indicates sowing, thus 1970 indicates crop season 1970-71). Wheat yield simulation was done using daily weather observations for the period 1980-1998. Simulated wheat yields (YSIM) were compared against observed district average

yield for the period 1980-1994 for developing yield prediction approach involving six steps as given below:

(a) Computation of trend yield

Observed yields (YOBS) for the period 1970-1994 were regressed with time to obtain contribution of technological improvement in observed yield. A segmented regression for 1970-1987 and 1988-1994 was found suitable for calculating trend yields. The technology trend of later period was used to predict trend yields (ŶOBS) for the years of 1995 to 1998.

For single trend $\hat{Y}OBS = a_1 + b_1 * Year$ (1) For segmented trend $\hat{Y}OBS1 = a_0 + b_0 * year$ (when year = 1987) (2) $\hat{Y}OBS2 = a_1 + b_1 * year$ (when year > 1987 (3)

(b) Estimation of observed yield deviation

Deviation in observed yield (YDOBS) from its technology trend (ŶOBSI, if year is = 1987 and ŶOBS2, if year is > 1987) represents the yield variability due to varying weather conditions in different crop seasons and was computed as

 $YDOBS(\%)=((YOBS - \hat{Y}OBS) / \hat{Y}OBS) * 100$ (4)

(c) Simulation of wheat yields and estimation of simulated trend yield

Wheat yield was simulated (YSIM) from 1980 to 1998, using CERES-Wheat model incorporated in DSSAT-3.5 modelling environment. The simulated wheat yields are expected to capture the deviation in yield arising from year-to-year variability in weather condition. Simulation yields (YSIM) have been regressed with years for calculating simulated trend yield (ŶSIM). ŶSIM= a, + b, * Year (5)

(d) Estimation of simulated yield deviation

Deviation in simulated yield (YDSIM) from its average/trend (ŶSIM) show the likely effect of weather on average wheat yield as identified by CERES-Wheat model and was computed as:

 $YDSIM(\%)=((YSIM-\hat{Y}SIM)/\hat{Y}SIM)*100$(6)

(e) Prediction of yield deviation

To predict the yield deviation (ŶD), observed yield deviations (YDOBS) were regressed with simulated yield deviations (YDSIM) and estimated regression coefficients a³ and b³ were used for prediction of yield deviation as:

$$\hat{Y}D(\%) = a_3 + b_3 * YDSIM(\%) + (7)$$

(f) Yield forecast

Likely increase in crop yield due to technological advancement was obtained using step (a), and likely impact of current year weather on crop performance have been judged using step (d)(method I) or step (e)(method II). The yield prediction for the period 1995-1998 were generated by these two methods as:

Method I:

$$\hat{\mathbf{Y}}\mathbf{P}\mathbf{1} = \hat{\mathbf{Y}}\mathbf{O}\mathbf{B}\mathbf{S} + (\mathbf{Y}\mathbf{D}\mathbf{S}\mathbf{I}\mathbf{M}(\%) * \hat{\mathbf{Y}}\mathbf{O}\mathbf{B}\mathbf{S})/100$$

.... (8)

Method II:

$$\hat{Y}P2 = \hat{Y}OBS + (\hat{Y}D(\%) * \hat{Y}OBS)/100$$
..... (9)

RESULTS AND DISSCUSSION

Analysis of wheat yield

The historical wheat yield series for Nainital district for the period 1970-1998 is given in Figure- 1. The wheat yields increased from 1180kg hall in 1970 to 2912 kg hall in 1998. A single linear trend indicates an increase in wheat yield by 62.9 kg halycarl, with RMSE (Root Mean Square Error) 275.25, and R2 0.87 (F = 192.3, P<0.0001). The plotted data (Figure 1) indicates higher variability and decreased rate of technology trend since late eighties. In order to capture weather caused yield variability, yield series was split in two parts from 1970-1987 and 1988 to 1994 (Figure 2). Use of segmented regression with two linear technology trends, decreased RMSE to 248.4. In the period I (1970-1987) wheat vield increased at 64.06 kg ha-lyear-1 (R2 0.657)(F = 208, P<0.0001), while in period II (1988-1998) yearly technological trend decreased to 43.39 kg ha 'year' (R2 0.087) (F = 1.021, P < 0.33). Low R² in period II indicates decreased technology trend and an increase in year-to-year yield variability in the last decade. Trend yields have been calculated from 1994 to 1998 (Figure 2) using following equation:

Y = 43.393 * X + 1679.5 (where X is year) (10)

Analysis of simulated wheat yields and yield deviation

Simulation of wheat yield over the 19 years period from 1980 to 1998 gave yield range from 3877 kg ha⁻¹ (1983) to 5289 kg ha⁻¹ (1996) and mean yield of 4445 kg ha⁻¹. A plot of simulated yield against time indicated a small trend of ŶSIM = 17.344 * X + 4259.6 (X=year) and R¹ 0.081 (F = 1.49, P<0.25). The yearly simulated yields (YSIM) were transformed to simulated yield deviation (ŶDSIM) from the simulated trend (ŶSIM) using eqs 6.

Deviations of simulated yields (YDSIM) from simulated trend were regressed with deviation of observed yield (YDOBS) from technology trend (Figure 3). The simulated and actual deviations were positively correlated (r = 0.577), and a linear regression equation ŶD(%) = 0.6405 * YDSIM(%) + 2.6061, with $R^2 0.3324$ (F = 6.5, P<5%) could be fitted. This result reveals that about 33 % deviation in wheat vield from trend could be captured by simulated yield deviation with a common set of cultivar and cultural inputs and information on year wise variable inputs of management practices from field surveys, the simulation would be needed to capture higher variability.

Forecast of wheat yields (1995-1998)

The forecasts of wheat yield for the period 1995 to 1998 were generated using three steps approach, comprising (a) prediction of trend yields (ŶOBS) from equation 3, (b) simulation of wheat yields (YSIM) and estimation of simulated wheat

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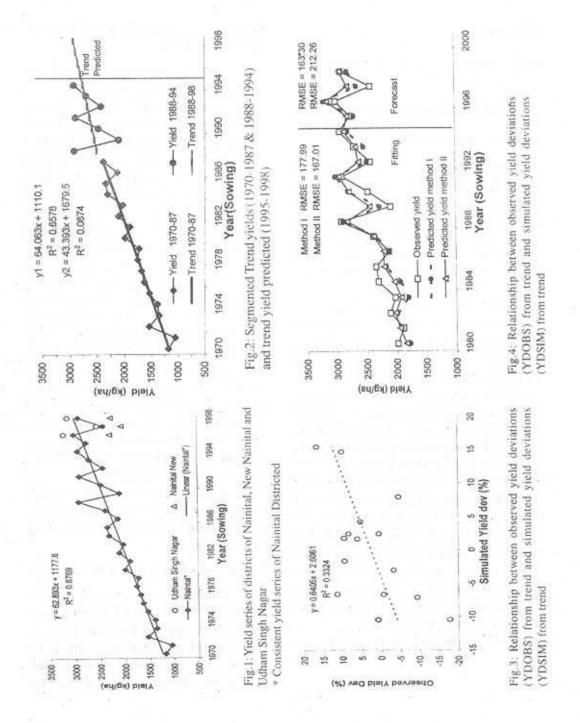


Table 2: Observed and predicted yield during 1980 to 1998.

Year of sowing	Observed yield(kg ha ⁻¹)	Trend yield(kg ha ⁻¹)	Predicted yield			
			Method I		Method II	
			kg ha-	Dev.%	kg ha	Dev.%
		Fitting	g period			
1980	1986	1815	1785	-10.12	1843	-7.19
1981	1889	1879	1924	1.86	1957	3.59
1982	2105	1943	1990	-5.46	2024	-3.86
1983	2025	2007	1797	-11.24	1925	-4.93
1984	2306	2071	1935	-16,11	2038	-11.64
1985 -	2339	2135	2172	-7.12	2215	-5.31
1986	2127	2199	2133	0.30	2214	4.10
1987	2381	2263	2358	-0.95	2383	0.08
1988	2917	2504	2891	-0.87	2817	-3.41
1989	2095	2547	2282	8.92	2444	16.64
1990	2470	2591	2795	13.18	2789	12.93
1991	2897	2634	3021	4.28	2951	1.85
1992	2425	2678	2487	2.56	2625	- 8.26
1993	2704	2721	2540	-6.06	2676	-1.03
1994	2931	2764	2810	-4.13	2865	-2.23
RMSE		219.5	177.99		167.01	
		Forecast	ing perio	d s		
1995	2764	2808	2699	-2.36	2811	1.70
1996	3002	2851	3199	6.56	3148	4.87
1997	2415	2895	2645	9.53	2810	16.40
1998	2912	2938	2809	-3.53	2932	0.69
RMSE		252.6	163.3		212.26	

For method I, and method II see equations 8 and 9, respectively, in text.

yield deviations (YDSIM) and predicted yield deviations (ŶD), and (c) forecast of final yields using equations 8 and 9. In the calibration period (1980-1994) RMSE between observed and predicted yields for method I & II were 177.99 and 167.01. respectively, however in the forecasting period (1995-1998) RMSE were 163,25 and 212.26, respectively. This suggests that method II performs better while fitting, but for the purpose of forecasting it could not reduce the biases. Year-wise comparison of the forecast by the two procedures indicate (Table 2) that in three out of four years, method II have higher accuracy. The performance of the approaches in fitting and forecast periods is summarized in Figure 4.

The study demonstrates that CERES-Wheat model could simulate the pattern of weather induced yield variability in district yield series over the long period of 19 years (1980-1998). Use of simple linear technology trends and simulated yield deviation (YDSIM) can be adopted for yield prediction. In this approach daily weather data of the entire season determine final vield. The cultural practice and technology inputs were kept constant in the period. However, using appropriate field survey data, these inputs could be specified as applied by farmers, thereby eliminating the need for linear modelling of technology trends.

The important issues in use of crop simulation model for regional yield prediction which have not been considered in this study, are (a) Spatial variability in

cultural practice, (b) Spatial variability in soil properties, and (c) Spatial variability of weather. Current research efforts are addressing these issues by adopting use of GIS and remote sensing data. Spatial variability of weather and soil were included for sorghum yield prediction by linking simulation data to GIS (Rosenthal et al., 1998). Use of direct observations on crop growth from remote sensing data at field to regional scales as inputs to simulate models has been reviewed by Moulin et al., (1998). Thornton et al., (1997) used remote sensing derived rainfall and GIS with crop simulation model to predict pearl millet vield in Burkina Faso. A similar extension of the current study for use of GIS in regional wheat yield prediction is in progress.

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