

## Development of district-level Agro-meteorological Cotton Yield Models in Punjab

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### Abstract

District-level agromet-cotton yield models were developed using fortnightly meteorological variables like mean maximum temperature (MXT), mean minimum temperature (MNT) and total rainfall (TRF) from first fortnight of June to First fortnight of November of every year for 24 years (1980-2003). The crop condition (CC) term was also incorporated into the yield model to account for yield losses due to pest/disease or drought conditions. This study was conducted in the five major cotton producing districts namely Bathinda, Faridkot, Firozpur, Mansa and Muktasar which account for 93.9 per cent of acreage and 95.5 per cent of cotton production in Punjab State. Technology trend has been separately modelled using time series data of district-level cotton yield data of 24 years. Yield deviations from their respective trends have been regressed against the selected subset of explanatory variables. The weather variable subset for each district was selected by backward elimination procedure based on the strength and significance of association observed from the correlation matrix. The regression coefficients were tested for significance at 95 percent confidence level using two tailed 't' test. Repetitive analysis for different districts has been facilitated by a semi-automated procedure based on macros of different steps in data analysis.

The best sub-set of 3-4 explanatory agro-meteorological variables, which resulted in highest  $R^2$  with minimum SEOE were selected for each district independently. Using these set of variables cotton yields were predicted for each district. The results indicate that the adjusted  $R^2$  for all the five districts range from 0.91 to 0.95. This indicates that these variables explain around 91 to 95 per cent variability in the cotton yields in five districts. The most significant variables in the regression equations of five districts are: Crop Condition (CC), TRFJUN1, MNT OCT1 and MXTNOV1. The agro-

meteorological model predicted and observed cotton yields in five districts indicated that the relative deviations were in the range of -0.5 to 10 per cent in all the districts from 1980 to 2003 period.

**Key words:** Agro-meteorological variables, Step-wise regression, Crop Condition term, Relative Deviations, Mean Absolute Percent Error (MAPE),

## **1. INTRODUCTION**

Cotton is one of the finest natural fibres available to mankind for clothing from time immemorial. Cotton is an important commercial crop of the country and contributing nearly 85% of the total domestic fibre consumption. It is estimated that cotton requirement in India by 2025 will be around 140 lakh bales of lint and the present production is around 123 lakh bales. India is the third largest cotton producer after China and the United States of America. In terms of acreage India has largest cotton acreage, accounting for 25 per cent of world acreage. However, India has the lowest average yield (287 kg/ha against the world average of 567 kg/ha) amongst the major cotton producing countries of the world.

### **1.1 Cotton growing regions in India**

Cotton is mainly concentrated in four agro-ecological regions in India, either as a sole or as an intercrop in cereal or pulse based cropping systems. Four species of cotton grown in India, of which two are diploid (*Gossypium arboreum* and *G. herbaceum*) and the other two tetraploids (*G. hirsutum* and *G. barbadense*). The diploid species referred to as 'Desi' variety and accounts for 25-30 % of the production. These varieties have low productivity and are not responsive to good agronomic practices in terms of yield. In addition, its fibre is rough and short. However, the negligible cost of desi cotton seeds accounts for its popularity amongst the poor farmers. The tetraploid varieties account for the remaining 70% of the cotton production in India. These varieties provide fine quality fibre, which is used by the textile industry. Although hybrid seeds are costly, their yield can be as high as 800 kg/ha, depending on the agronomic conditions. However, this too is lower than the 1200 kg/ha yield obtained in the US.

Cotton growing regions in India can be categorised into three major zones (**Figure-1**) and nearly 99 per cent of cotton production in the country comes from first three zones with Eastern zone contributing merely one per cent. The northern zone typically produces 35 per cent of the total cotton produced in India, the central zone 40 per cent and the southern zone 24 per cent. The northern zone is the primary producer of short and medium staple cotton. These nine states are in first three zones, have distinct characteristics in terms of soil base, varietal composition, irrigation practices, crop rotations and intercrops, sowing and harvest periods and agro-climatic conditions, which reflect on the realised yield levels and yield trends in respective zones.

### **1.2 Problems in Cotton Forecasting**

Peculiarities like indeterminacy in growth habits, longer maturation period, multiple

cultivated species, multiple pickings, severe pest infestation etc. make cotton a challenging crop. Variations in cotton yield mainly arise out of variations in i) initial crop stand, ii) post-flowering growth habits, and iii) the shedding of flowers and bolls, i.e., the bearing capacity of plants. Physiological response of cotton to stresses differs significantly from other crops. It sheds reproductive organs viz. squares, flowers and bolls in response to stresses and subsequently generates new organs under favourable conditions. Information on cotton yield losses due to severity and extent of pest/disease infestation is important for cotton due to its known susceptibility. Modelling on this front is hindered due to lack of basic understanding of pest dynamics in relation to weather conditions. Ground observations of severity and extent are not available readily. Despite these complexities cotton growth and development is fairly regular and it is possible to devise empirical regression models for yield estimation. Thus, regression techniques using a variety of data sources related to major groups of yield affecting parameters are used for cotton yield modelling.

### **1.3 Agro-meteorological Cotton Yield Model Development in India**

Zonal-yield models incorporating a linear time trend and agro-meteorological (agromet) variables each spanning successive fortnights within the growth period of the cotton crop are developed within the framework of multiple linear regression analysis. These models have been used to predict the cotton yields in five cotton growing districts namely; Hisar, Sirsa, Bhiwani, Rohtak and Jind covering more than 90% of cotton production of the Haryana State (Verma et al., 2014). Kalubarme et al. (1997) studied pre-harvest production forecasts for the cotton crop in Hisar and Sirsa districts. Cotton yield models for three main cotton zones viz. North, Central and South zones of the country have been developed using 22-year (1970-92) time series of monthly weather data of 42 stations (Dubey and Sehgal, 1997). The trend and weather variables based models accounted for 82, 65 and 90 per cent yield variability in North, Central and South zones, respectively. State-level cotton models for acreage and yield have been developed for national cotton production forecasting, using fortnightly minimum-maximum temperatures and rainfall data of 15 meteorological subdivisions (Kalubarme and Dubey, 1999). Tehsil-wise technological trend and fortnightly rainfall-based cotton yield models for seven districts in Khandesh, Marathwada and Vidarbha regions of Maharashtra were developed using daily rainfall data aggregated to fortnightly totals and historical yields of cotton for last 23 years (Patil, et al., 1997). The multiple regression models contained one to four rainfall variables. Except for time trend in three talukas, the relations for trend as well as rainfall variables were significant. The total yield variability explained by trend and rainfall regression ranged from 12 to 93 per cent.

In Surendranagar district of Gujarat, cotton yield models have been developed using trend analysis, multi-date remote sensing data and climatic evapo-transpiration (Ray et al., 1994). The actual evapotranspiration (AET) estimated from simple soil moisture balance (Lhomme and Katerji, 1991) correlates well with regional cotton yield, especially in a semi-arid rainfed situation. Cotton yields in Punjab and Andhra Pradesh are estimated by using analysis of historical yield data and inputs on relative

crop condition and varietal improvement (Anonymous, 1996). The models based on remote sensing and meteorological data have been developed and used for cotton yield forecasting in Gujarat and Maharashtra (Ray et al., 1999; Dubey and Chaudhary, 1995). Towards development of simple yield forecasting procedures, the time trend has been used for yield estimation of different crops including cotton (Bhagia and Dubey, 1994). Kharche (1984) has also validated Gossym simulation model for growth and yield estimation of cotton. Role of rainfall and other climatic data for cotton yield forecasting has been visualised and used in cotton yield forecasting (Jahagirdar and Ratnalikar, 1995).

## **2. OBJECTIVES**

Timely and accurate estimates of area and production of cotton with less subjectivity would be highly useful in deciding the exportable surplus. Comprehensive, reliable and timely data collection over extensive areas is the major problem. Therefore, in the present study Subdivision-level weather data like minimum-maximum temperature and rainfall has been analysed to develop agro-meteorological cotton yield models. Agromet-yield models combining weather parameters have been generated for cotton yield prediction in five major cotton growing districts in Punjab State. The broad objective of this study is development of district-level Agro-meteorological yield models in Punjab State.

## **3. MATERIALS AND METHODS**

### **3.1 Study Area and its Location**

The state of Punjab is one of the most intensively cultivated and irrigated areas of the world. Area under agriculture is around 89 per cent of the total geographical area (5.036 m ha) of the Punjab state with cropping intensity of 178 per cent, thus leaving hardly any scope for further increase in area under agriculture. The success of agricultural sector in Punjab state has very few parallels in the world. Scope for horizontal expansion of agricultural land is limited but the scope for vertical expansion for increasing productivity of lands already under cultivation with improved technology is enormous. The Punjab State, with a geographical area of 50,362 sq. km, forms a part of the Indus plain. It lies between 29° 33' to 32° 31' N latitude and 73° 53' to 76° 55' E longitude. Punjab is bounded by Pakistan to its west, Jammu and Kashmir to its north, Himachal Pradesh to its east and north-east, Haryana to its east and south-east and Rajasthan to its west (**Figure-2**). The general gradient of the state is from northeast to southwest. Most of the region lies between 175-300 m above mean sea level and has a general gradient of about 1:2200.

### **3.2 Cotton Producing Areas in Punjab State**

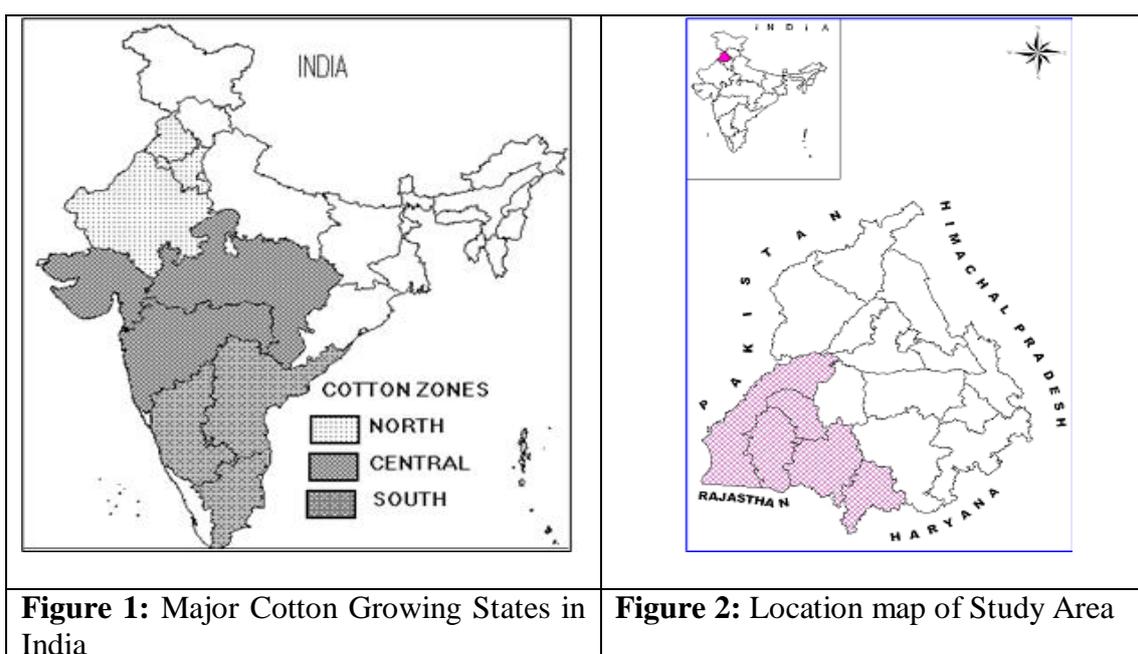
Punjab is one of the major cotton producing states of India, which contributes about 6.4 per cent of cotton acreage and 9.2 per cent of cotton production in the country (average of last five years from 1998-99 to 2002-03). In Punjab State, Bathinda, Faridkot, Firozpur, Mansa and Muktasar are the major cotton producing districts,

accounting for 93.9 per cent of acreage and 95.5 per cent of the cotton production of the state (average of last five years from 1999-2000 to 2003-04). Cotton acreage and production in Punjab State during Kharif 2003-2004 was 0.452 million hectares (mha) and 1.478 million bales.

### 3.3 Data Used

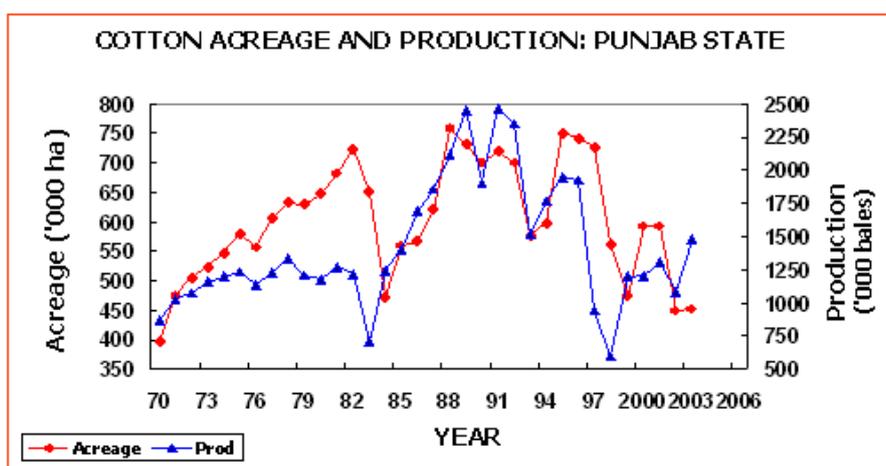
#### 3.3.1 Cotton Acreage, Yield and Production Statistics

Historical cotton crop statistics at district and state-level, published by the Bureau of Economics and statistics (BES) of 25 years (1980-2004) collected for trend analysis (Figure-3).



**Figure 1:** Major Cotton Growing States in India

**Figure 2:** Location map of Study Area



**Figure 3:** Cotton Acreage and Production in Punjab State

### **3.3.2 Meteorological Data**

Meteorological data like minimum temperature, maximum temperature, and rainfall was collected from Bathinda meteorological observatory in the study area. The daily meteorological data was compiled as fortnightly average values for the period of June to December for developing district-level agrometeorological cotton yield models. In case of rainfall data, instead of average values total rainfall during the fortnight was computed.

### **3.3.3 Ground Truth Data**

Field data at selected sites in the study districts was collected during the cotton growing seasons. The field data collected includes variety, crop growth stage, crop vigour, disease/insect-pest attack, soils, irrigation, fertilizer application, etc.

## **3.4 Data Analysis Procedure**

### **3.4.1 Data Aggregation and Data-base Generation**

The daily meteorological data like minimum-maximum temperatures (MNT-MXT), and total rainfall (TRF) were aggregated over fortnightly values as independent response variables. The first fortnight of May refers to average values of daily data over May 1-15, while second fortnight of May (for example) refers to average values of daily data over May 16-31; and likewise for other months from June to November. The crop condition term computed from the historical yield data was also included in the data series. This term takes into account the impact of yield reducing factors like crop stresses (moisture stress, disease/pest stress etc.). Chakraborty, 1991 also incorporated this crop condition term in the multiple regression for district-level cotton yield forecast in Punjab State. The data was arranged in a matrix form where each row refers to the data set of one year with columns as year, yield and fortnightly meteorological values. Such data matrix was prepared for each district.

### **3.4.2 Semi-automated Procedure for Agro-meteorological Model Development**

A Semi-automated procedure for development of cotton yield models based on trend and fortnightly rainfall, maximum temperature (Tmax), and minimum temperature (Tmin) was designed and developed. In this procedure, a set of command language programs (macros) were written for semi-automated execution of model development steps, in order to improve the efficiency of model development. A sequence of five macros were written using development tools of spreadsheet (Quattro Pro) software for following distinct steps of this procedure (Kalubarme, and Dubey, 1999), they are as follows:

- a) **Data matrix generation:** The data is arranged in a matrix form where each row refers to the data set of one year with columns as year, fortnightly Tmax, Tmin and rainfall and cotton yield.
- b) **Scatter plot creation for time trend:** Year versus cotton yield line graph is generated from the data matrix.
- c) **Computation of trend line and normalised yield deviations:** Based on observations of scatter plot, trend analysis and normalised deviations of yields (NDY) are computed for one of the three options like Average, Single trend

- and Two trends.
- d) **Correlation analysis and regression with backward elimination:** A correlation matrix of NDY and Tmax, Tmin and rainfall is generated and subset of six best variables is selected on the basis of magnitude of correlation coefficient.
- e) **Cotton yield prediction and its graphical presentation:** Forecast of cotton yield is generated using current season data and predicted values for current as well as previous years are plotted on the original scatter plot.

### 3.4.3 Trend Analysis

For the purpose of trend analysis, 1970 was taken as reference year and the subsequent years were referred to as modified years (Tm) where,  $T_m = (\text{YEAR} - 1970)$  and YEAR is 1971 for 1971-72 crop season. The scatter plots of year vs. yield have been created for evaluating the time trend. Based on the assessment of scatter plots, single or two-trend linear analysis is performed and appropriate regression equations and their significance is obtained. However, if the regression line showed the significant slope and the data points were uniformly distributed around this line, then a single trend line was assumed to apply to the entire yield time series. The significance of the slope of the equation was tested by using two-tailed t-test at 95% confidence interval. The general form of the equation fitted is as follows (Kalubarme, and Dubey, 1999):

$$Y = a + b * T_m$$

Where,

Y = seed cotton yield (kg/ha)

Tm = modified year i.e. (Year-1970)

a = intercept,

b = slope

Depending on the applicability of one of the above situations, i.e. no trend, single linear trend and two trends, the further analysis have been done to compute the trend-predicted yields, using the following equations:

i) No Trend: 
$$Y_{ti} = \left( \sum_{i=1}^n Y_i \right) / n$$

ii) Single Trend: 
$$Y_{ti} = a + b * T_{mi}$$

iii) Two Trends: 
$$Y_{ti} = \left( \sum_{i=1}^p Y_i \right) / p$$

Where,

$Y_{ti}$  = predicted yield for the ith year

$Y_i$  = seed cotton yield of ith year

$T_{mi}$  = modified year for ith year

- n, p = number of years used in model  
 a = intercept of trend line  
 b = coefficient of regression of trend line

#### 3.4.4 Normalized Yield Deviations

Highly varying component of yield time series appears as transient around the trend line. Normalised yield deviations representing these transients were computed and related to the weather variables. The normalised yield deviation is absolute difference of observed yield and the predicted yield from trend equation for a given year, which is normalised with respect to predicted yield to take care of proportionate inter-annual yield fluctuations. Trend predicted yield values are plotted against the observed yields to assess the visual fit of model to data. The option giving the best fit is selected for computation of normalised yield deviations using the following formula:

$$\Delta Y_i = Y_{ti} - Y_{oi}$$

$$NDY_i = \Delta Y_i / Y_{ti}$$

$$= (Y_{ti} - Y_{oi}) / Y_{ti}$$

Where,

$\Delta Y_i$  = absolute yield deviation for ith year

$Y_{ti}$  = trend predicted yield from trend for ith year

$Y_{oi}$  = observed yield for the ith year

$NDY_i$  = normalized deviations of yield for ith year, with respect to long term trend

#### 3.4.5 Correlation and regression analysis

Correlation matrix of fortnightly weather variables and the normalised yield deviations was generated and examined for magnitude and sign of correlations. Response variables showing significant correlation, at 95 per cent confidence level, with the yield deviations, were noted in the ascending order of magnitude. The procedure also checks the variables for multi-collinearity, and final subset of significant variables was used for regression analysis. Based on two tailed 't'-test (95% confidence) of the regression coefficient, the non-significant variables were eliminated, one at a time. In this backward elimination procedure, exceptions were made if the coefficient of determination ( $R^2$ ) declined disproportionately on the elimination of a particular variable, so as to cover the indirect effect of such a variable in explaining the yield variability. Thus the final equation contained variables returning significant slope as well as overall adjusted  $R^2$ . If individual slope of none of the variables was significant, an equation giving significant adjusted  $R^2$  with minimum number of response variables was selected. The general form of the model realized is as follows:

$$NDY = a_0 + \sum_{i=1}^N (b_i * X_i)$$

where,

$NDY$  = normalized deviation of yield

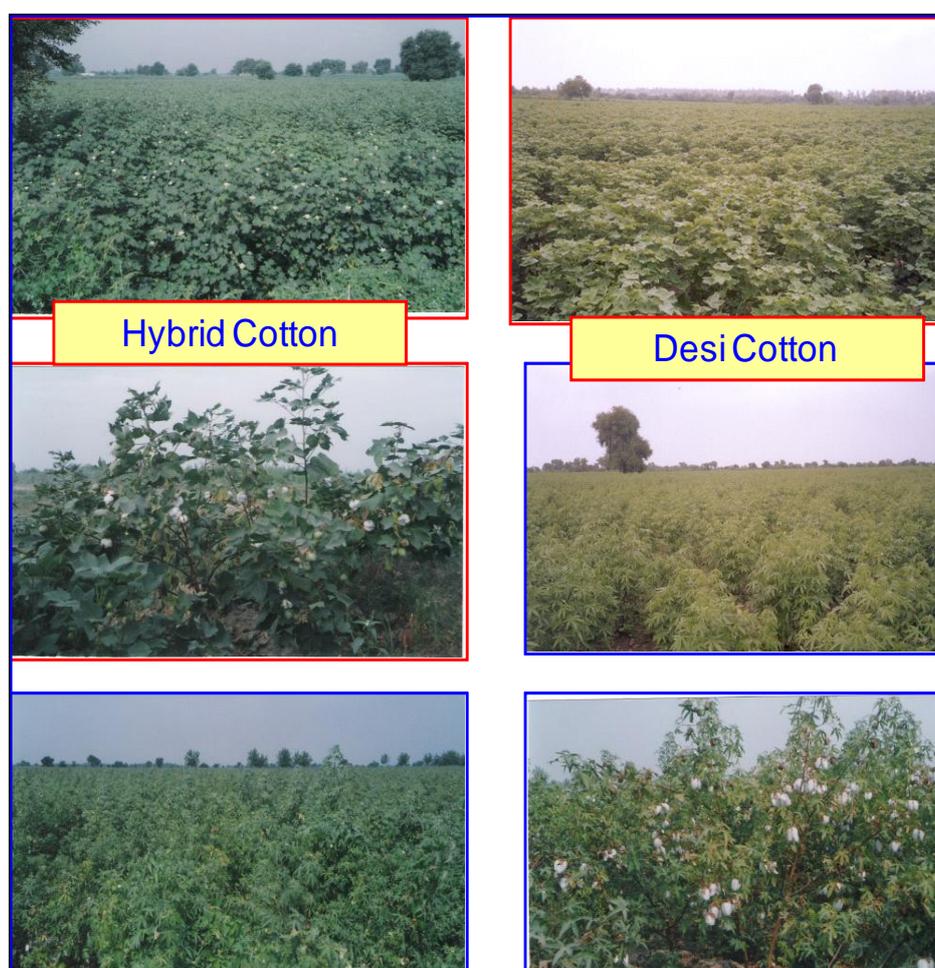
$a_0$  = intercept of multiple regression equation

$b_i$  = regression coefficient of ith variable

$X_i$  =  $i$ th fortnightly weather (response) variable

### 3.4.6 Ground Truth Data Collection

Ground Truth (GT) was collected during second fortnight of September to first fortnight of October each year, which coincided with flowering, boll formation and boll opening stage of cotton. During this period first picking of cotton also commences in most of the cotton growing districts. Large homogeneous sites of cotton grown under different cultivation practices, soils, and management practices in different districts were identified using the previous seasons IRS LISS-III FCC and transferred on 1:50, 000 scale base maps. These sites were visited and agronomic observations like crop growth stage, crop density/vigour, irrigation; moisture stress, variety, number of pickings, disease/pest attack etc. were recorded for each site. The GPS readings of important sites in the cotton growing districts were recorded using global positioning system (GPS) receiver. Field photographs of cotton varieties at different growth stages are shown in **Figure-4**.

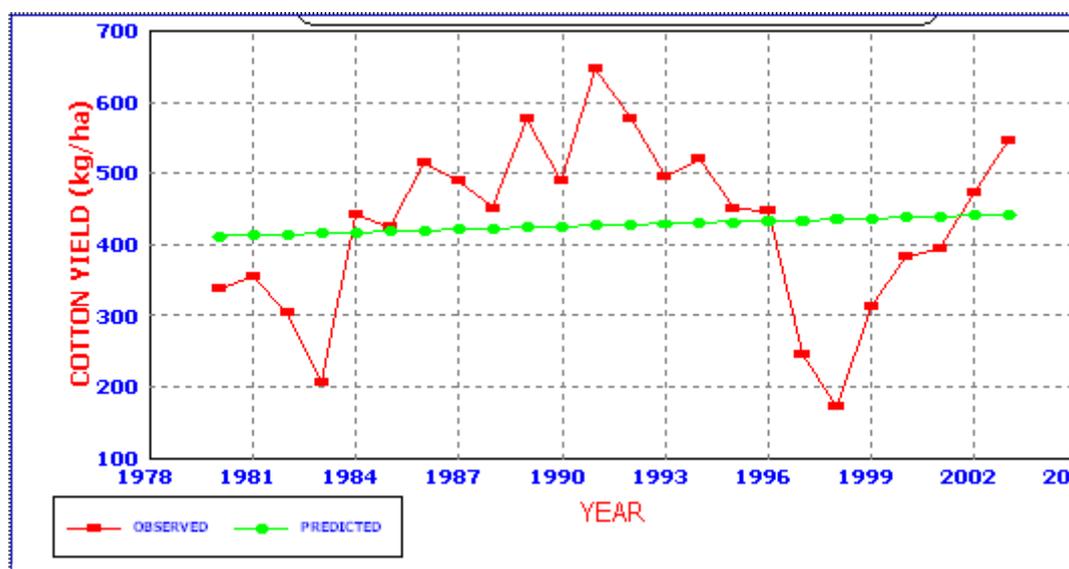


**Figure-4:** Field Photographs of Cotton Varieties grown in Punjab State

## 4. RESULTS AND DISCUSSIONS

### 4.1 Trend Predicted Cotton Yields

The scatter plots of year vs. cotton yields for five districts were generated to assess the fluctuations and trends in cotton yields over the years. It was observed that the cotton yields were highly fluctuating over the period of 24 years in cotton growing districts of Punjab State depending upon the climatic conditions and incidence of pest/disease. During the kharif seasons from 1984 to 1996, cotton yields in all the districts were between 400 to 650 kg/ha and showed an increasing trend. However, cotton yields during the periods of 1980 to 1983 and 1997 to 2002 were below 400 kg/ha and cotton seasons of 1983, 1997 and 1998 showed drastic reductions in cotton yields and were below 250 kg/ha in all the cotton growing districts. A single trend was fitted to the yield time series data for all the districts and trend predicted yields were computed. The scatter plot of observed yields along with trend predicted yields for one of the districts (Bathinda) is given in **Figure-5**.



**Figure-5:** Trend Predicted Cotton Yields in Bathinda District of Punjab State

### 4.2 Correlation and regression analysis with backward elimination

The trend predicted yields in all the districts were used to compute normalized yield deviations. The computed yield deviations were regressed against the fortnightly meteorological variables like mean maximum temperature (MXT), mean minimum temperature (MNT) and total rainfall (TRF) from first fortnight of June to First fortnight of November of every year for 24 years (1980-2003). The data-set of yield deviations along with Crop Condition (CC) term and meteorological variables generated for one of the districts (Bathinda) is shown in **Table-1**.

From the correlation matrix of yield deviations and meteorological variables a subset of top eight variables showing significant correlations, at 95 % confidence level, with yield deviations was selected for further regression analysis (**Table-2**). In the multiple

regression analysis of eight variable with yield deviations, the non-significant variables were eliminated, one at a time, by backward elimination procedure based on two tailed 't'-test (95% confidence).

**Table-1:** Correlation data set generated for Agromet-Yield model development in Bathinda

YEAR	MYR	COM_YLD	CC	TRFJUN1	MXTJUN1	MNTJUN1	TRFJUN2	MXTJUN2	MNTJUN2	TRFJUL	MXTJUL	MNTJUL
1980	10	-0.198449	3	0.00	43.17	27.78	5.70	42.55	28.34	199.20	37.25	27.57
1981	11	-0.233668	3	0.00	43.35	24.44	22.00	44.23	28.07	132.40	36.61	25.55
1982	12	-0.250185	3	0.00	43.25	24.97	13.80	39.85	25.78	0.00	41.53	27.99
1983	13	-0.59578	1	4.00	40.67	24.69	14.80	40.48	25.85	26.20	38.93	25.36
1984	14	0.132088	4	12.40	43.79	28.79	0.00	41.04	30.16	67.00	36.92	25.79
1985	15	0.194774	4	46.80	39.49	26.37	41.00	42.03	28.03	69.50	37.07	26.37
1986	16	0.241582	5	1.20	39.97	25.36	425.60	40.23	27.73	65.60	33.56	25.35
1987	17	0.017958	4	95.20	40.29	25.98	21.40	41.53	26.81	50.80	40.60	26.49
1988	18	0.138771	4	0.00	43.45	27.44	36.20	40.43	27.97	34.40	37.55	27.99
1989	19	0.303592	5	32.80	40.03	25.53	4.50	39.04	26.01	35.90	37.75	26.43
1990	20	0.118829	4	9.20	41.11	25.21	6.40	42.28	29.85	437.20	33.23	26.30
1991	21	0.392176	5	58.50	41.96	25.61	27.40	39.39	26.20	12.80	41.27	28.65
1992	22	0.372966	5	8.60	40.77	25.31	31.70	42.49	28.19	114.50	36.82	26.10
1993	23	0.144106	5	69.60	38.75	26.57	68.90	34.27	26.08	69.90	32.87	24.39
1994	24	0.218528	5	44.30	42.57	26.33	32.40	41.91	27.75	128.50	37.38	27.87
1995	25	0.155718	5	0.00	45.85	27.66	84.00	39.80	26.79	0.00	40.43	29.33
1996	26	0.06168	4	17.00	40.94	26.75	113.20	35.60	25.84	40.40	36.55	27.14
1997	27	-0.502366	2	0.00	36.97	24.39	66.60	39.37	27.36	18.00	38.52	27.87
1998	28	-0.505182	2	20.30	39.25	25.79	11.00	41.66	28.41	89.80	38.02	27.27
1999	29	-0.232475	3	15.60	38.42	24.73	11.00	39.72	28.65	69.00	38.91	28.51
2000	30	0.059913	5	7.20	39.57	26.76	10.80	40.93	29.80	126.80	37.71	27.97
2001	31	-0.179216	3	46.50	37.02	25.18	84.10	35.45	25.65	118.00	36.19	27.03
2002	32	-0.160498	4	0.00	40.07	27.97	25.00	38.80	27.83	18.00	38.90	29.45
2003	33	0.290845	5	0.00	43.30	28.47	22.50	38.19	27.65	68.00	35.59	28.67
2004	34		5	5.50	38.77	25.47	63.52	35.46	25.71	0.00	37.87	28.70

**Table-2:** A subset of eight variables selected for regression analysis of Bathinda District

MYR	COM_YLD	CC	TRFOCT2	TRFJUL2	MNTJUN1	MNTOCT1	MXTJUN1	MXTNOV1	MNTOCT2	
108	10	-0.198449	3	0.60	113.20	27.78	19.61	43.17	28.56	16.44
109	11	-0.233668	3	4.70	25.40	24.44	16.03	43.35	27.05	14.24
110	12	-0.250185	3	0.00	0.20	24.97	18.96	43.25	29.78	15.50
111	13	-0.59578	1	0.00	40.10	24.69	18.39	40.67	29.57	12.04
112	14	0.132088	4	1.00	75.20	28.79	15.01	43.79	29.15	13.83
113	15	0.194774	4	0.00	158.30	25.37	19.06	39.49	29.85	12.15
114	16	0.241582	5	33.40	31.40	25.36	20.53	39.97	31.49	13.63
115	17	0.017958	4	31.00	32.80	25.98	19.55	40.29	31.08	16.29
116	18	0.138771	4	0.00	64.40	27.44	18.06	43.45	30.76	15.86
117	19	0.303592	5	0.00	78.40	25.53	18.31	40.03	30.15	13.08
118	20	0.118829	4	0.00	67.70	25.21	19.33	41.11	30.74	12.58
119	21	0.392176	5	2.10	57.30	25.61	15.40	41.96	29.47	14.68
120	22	0.372966	5	1.00	126.30	25.31	17.87	40.77	29.60	14.93
121	23	0.144106	5	1.00	115.30	26.57	14.92	38.75	27.37	14.13
122	24	0.218528	5	1.00	139.20	26.33	16.57	42.57	29.40	13.53
123	25	0.155718	5	0.80	127.10	27.66	20.05	45.85	30.39	14.75
124	26	0.06168	4	0.00	14.70	26.75	18.13	40.94	31.13	13.56
125	27	-0.502366	2	59.90	93.60	24.39	18.20	36.97	25.68	15.29
126	28	-0.505182	2	35.00	46.20	25.79	21.51	39.25	29.35	17.07
127	29	-0.232475	3	21.65	26.20	24.73	19.71	38.42	31.38	17.31
128	30	0.059913	5	0.00	35.30	26.76	20.43	39.57	31.36	16.94
129	31	-0.179216	3	0.00	103.00	25.18	19.35	37.02	31.29	16.68
130	32	-0.160498	4	10.00	60.00	27.97	19.49	40.07	30.75	16.78
131	33	0.290845	5	0.00	136.00	28.47	18.53	43.30	29.91	15.09
132	34		5	0.00	38.50	25.47	17.77	38.77	29.5	15.02

### 4.3 District-level Agrometeorological Cotton Yield Models

The final subset of 3-4 explanatory meteorological variables, which resulted in highest  $R^2$  with minimum SEOE were selected for each district independently. The regression output of Bathinda district showing  $R^2$ , SEOE, t-value etc. is given in **Table-3**. The combined  $R^2$  ( $RC^2$ ) of meteorological indices and trend was computed which indicates the variability explained by these variables in each district. Based on multiple regression analysis a set of variables selected for yield model development and yield prediction for each district along with their coefficients,  $R^2$ , Adjusted  $R^2$ , SEOE for five districts were also generated.

**Table-3:** The regression output of Bathinda district showing  $R^2$ , SEOE, t-value and selected variables

Regression Output:				
Constant			-1.53564	
Std Err of Y Est			0.09396	
R Squared			0.93895	
No. of Observations			24	
Degrees of Freedom			20	
		<b>CC</b>	<b>MXTNOV1</b>	<b>TRFJUN1</b>
X Coefficient(s)	0.22163		0.02273	0.00174
Std Err of Coef.	0.01499		0.01412	0.00077
T-Ratio	14.7825		1.60966	2.2697

The results indicate that the adjusted  $R^2$  for all the five districts range from 0.91 to 0.95. This indicates that these variables explain around 91 to 95 per cent variability in the cotton yields in five districts. The most significant variables in the regression equations of five districts are: Crop Condition (CC), Total Rainfall-June-I fortnight (TRFJUN1), Minimum Temperature-October-I fortnight (MNT OCT1) and Maximum Temperature-November-I fortnight (MXTNOV1). The agromet model predicted and observed cotton yields for two districts, Bathinda and Muktasar in Punjab State are given in **Figures-6 and 7**.

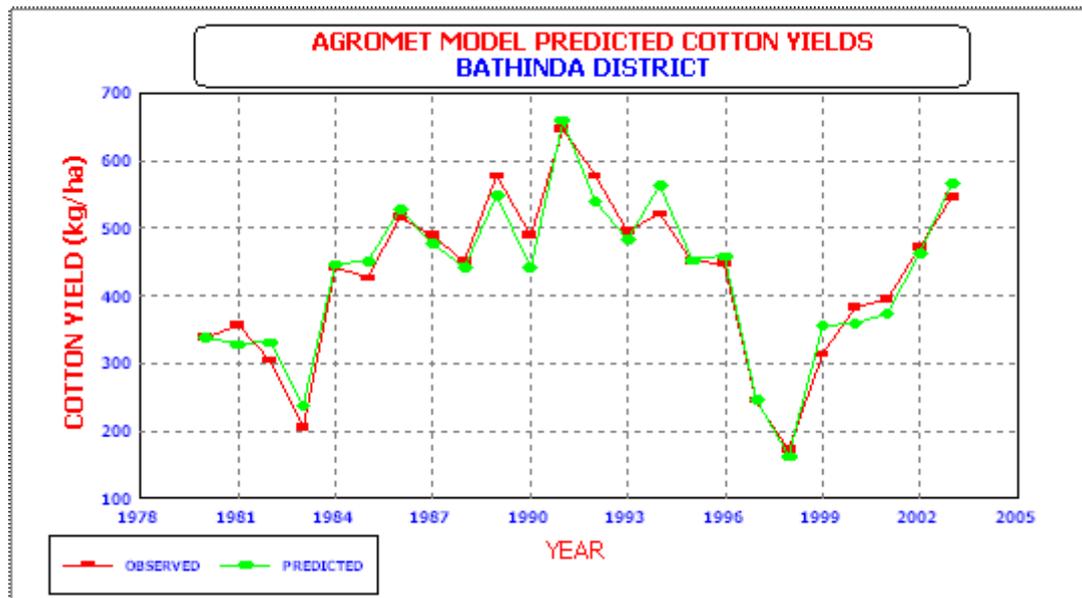


Figure-6: Agromet Model Predicted Cotton Yields in Bathinda District, Punjab State

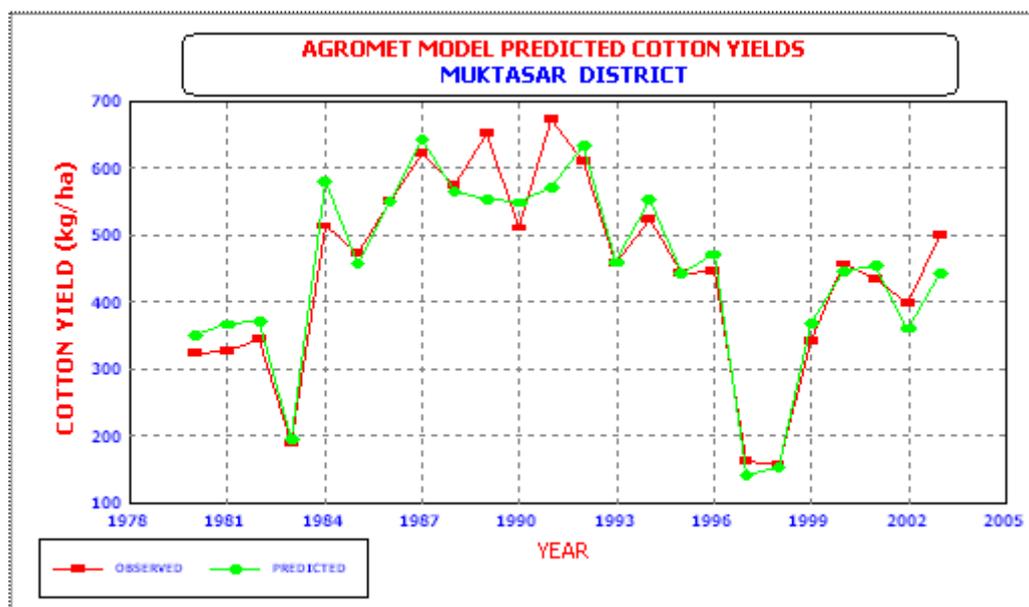


Figure-7: Agromet Model Predicted Cotton Yields in Muktasar District, Punjab State

#### 4.4 Performance Evaluation of District-level Agrometeorological Cotton-Yield Models

The performance of district-level Agrometeorological cotton-yield models developed using both meteorological variables in terms of their yield prediction capability was evaluated by computing Mean Absolute Per cent Error (MAPE) in comparison with

yields estimated by the Department of Agriculture, Government of Punjab. The MPAE of one of the districts (Bathinda) during 1980 to 2003 period are presented in **Table-4**.

**Table-4:** Observed and Predicted Cotton Yields along with MAPE for Bathinda District

Year	Cotton Yield (kg/ha)		Relative Devi (%)
	Observed	Predicted	
1980	339	326	-3.8
1981	356	342	-3.9
1982	304	320	5.3
1983	206	222	7.8
1984	443	447	0.9
1985	426	442	3.8
1986	516	520	0.8
1987	490	482	-1.6
1988	452	446	-1.3
1989	577	559	-3.1
1990	489	457	-6.5
1991	647	652	-0.8
1992	577	559	-3.1
1993	495	489	-1.2
1994	521	552	6.0
1995	451	455	0.9
1996	449	450	0.2
1997	245	249	1.2
1998	173	159	-8.1
1999	314	350	11.5
2000	383	372	-2.9
2001	395	377	-4.6
2002	472	467	-1.1
2003	546	555	1.8
		MAPE	-0.2

$$\text{Mean Absolute Percent Error (MAPE)} = \frac{1}{N} \sum \frac{Yld\ Pr\ ed - YldObs}{YldObs} * 100$$

It can be seen from Table-4 that in general, the MAPE of the predicted yields are within 10 per cent, except for one data point of 1999. The agro-meteorological model predicted and observed cotton yields in all the five districts indicated that the relative deviations were in the range of -0.5 to 10 per cent in all the districts from 1980 to 2003 period.

## CONCLUSIONS

This study was conducted in the five major cotton producing districts namely Bathinda, Faridkot, Firozpur, Mansa and Muktasar which account for 93.9 per cent of acreage and 95.5 per cent of cotton production in Punjab State. In this study Agro-meteorological yield modelling approach was adopted to develop district-level cotton yield models using Agro-meteorological data of past 25-years.

- ❖ District-level cotton yield models were developed using fortnightly meteorological variables like mean maximum temperature (MXT), mean minimum temperature (MNT) and total rainfall (TRF) from first fortnight of June to First fortnight of November of every year for 24 years (1980-2003).
- ❖ The crop condition term was also incorporated into the yield model to account for yield losses due to pest/disease or drought conditions. Technology trend has been separately modelled using time series data of district-level cotton yield data of 24 years. Yield deviations from their respective trends have been regressed against the selected subset of explanatory variables.
- ❖ The weather variable subset for each district was selected by backward elimination procedure based on the strength and significance of association observed from the correlation matrix. The regression coefficients were tested for significance at 95 percent confidence level using two tailed 't' test. Repetitive analysis for different districts has been facilitated by a semi-automated procedure based on macros of different steps in data analysis.
- ❖ The best sub-set of 3-4 explanatory agro-meteorological variables, which resulted in highest  $R^2$  with minimum SEOE were selected for each district independently. Using these set of variables cotton yields were predicted for each district. The results indicate that the adjusted  $R^2$  for all the five districts range from 0.91 to 0.95. This indicates that these variables explain around 91 to 95 per cent variability in the cotton yields in five districts.
- ❖ The most significant variables in the regression equations of five districts are: Crop Condition (CC), TRFJUN1, MNT OCT1 and MXT NOV1. The agro-meteorological model predicted and observed cotton yields in five districts indicated that the relative deviations were in the range of-0.5 to 10 per cent in all the districts from 1980 to 2003 period.

## REFERENCE

1. Anonymous, 1996. Cotton Acreage and Condition Assessment (CACA): Project Proposal for Phase-II. RSAM/SAC/CACA/PP/18/96, Space Applications Centre, Ahmedabad.
2. Bhagia, N. and Dubey, R.P., 1994. Yield modelling of cotton and sorghum using time series analysis. Scientific Note: RSAM/SAC/CAPE-II/SN/35/94, Space Applications Centre, Ahmedabad, 53.
3. Chakraborty, M., 1992. In Status Report of Cotton Acreage and Condition Assessment Project. RSAM/SAC/CACA/SR/01/92. Space Applications Centre (ISRO), Ahmedabad-53.

4. Dubey, R.C., Chaudhary, A., and Kale, J.D., 1995. The estimation of cotton yield based on weather parameters in Maharashtra. *Mausam*, 46(3): 275-278.
5. Dubey, R.P. and Sehgal, V.K., 1997. Development of weather based zonal models for cotton crop in India. Paper presented at INTROMET (International Symposium on Advances in Tropical Meteorology) under theme, "Monsoon and Management of Agricultural activities, IIT, New Delhi, Dec. 2-5, 1997.
6. Jahagirdar, S.W. and Ratnalikar, D.V., 1995. Cotton yield model based on monthly rainfall and rainy days. Paper submitted to the Research Review Subcommittee, P.K.V., Akola.
7. Kalubarme, M.H., Hooda, R.S., Yadav, Manoj, Saroha, G.P., Arya, V.S., Bhatt, H.P., Ruhel, D.S., Khera, A.P., Hooda, I.S. and Singh, C.P., 1998. Cotton production forecast using remote sensing data and agro-meteorological yield models in Haryana. Paper presented in National seminar on, 'Applications of Remote Sensing and GIS for sustainable development', Nov. 26-28, 1997, NRSA, Hyderabad.
8. Kalubarme, M.H., and Dubey, R.P., 1999. State-level agro-meteorological models for national cotton production forecast. Scientific Note: RSAM/SAC/FASAL-TD/SN/05/99, Space Applications Centre, Ahmedabad.
9. Kalubarme, M.H., Yadav, Manoj, Ruhel, D. S., Khera, A. P. and Hooda, I.S., 1997. Agro-meteorological cotton yield modelling in Haryana, Paper presented in TROPMET 97: Monsoon, Climate and Agriculture, Bangalore, February 10-14.
10. Kharche, S.G., 1984. Simulation models on growth and yield of cotton. A Dissertation submitted to the faculty of Mississippi State University for award of Ph.D. Degree in Agronomy.
11. Lhomme, J.P. and Katerji, N., 1991. A simple modeling of crop water balance for agrometeorological applications. *Ecological Modeling*, 57, 11-25.
12. Patil, S.M., Sehgal, V.K., Mehta, N.S., Shrivastava, S.K., Dubey, R.P. and Nalamwar, R.V., 1997. Taluqwise technology trend and rainfall based cotton yield models for seven districts in Khandesh, Marathwada and Vidarbha regions of Maharashtra. Scientific Note: RSAM/SAC/CAPE-II/SN/72/97.
13. Ray, S.S., Pokharna, S.S. and Ajai, 1994. Cotton production estimation using IRS-1B and meteorological data. *Int. J. Remote Sens.*, 15(5): 1085-1090.
14. Ray, S.S., Pokharna, S.S. and Ajai, 1999. Cotton yield estimation using agro-meteorological model and satellite-derived profile. *Int. J. Remote Sensing*, 1999, Vol. 20, No. 14, 2693-2702.
15. Verma, U., Piepho, H.P., Ogutu, J.O., Kalubarme, M.H. and Goyal, M., 2014. Development of agromet models for district-level cotton yield forecasts in Haryana State. *International J. of Agricultural and Statistical Sciences*, 10(1), 59-65.