

Forecast Analysis of Yearly Groundnut Productivity in India Using Auto Regressive Integrated Moving Averages model

B.R. Sreedhar

Sr. Asst. Professor, Department of Mathematics, CBIT Gandipet, Hyderabad (TS)-500075 e-mail: brsreedhar_maths@cbit.ac.in

ABSTRACT

This study paper is an observation of the productivity and production of peanuts in India. Data from 1966-67 to 2019-20 were analyzed using time collection methods. For the construction and forecasting of the versions from 1966-67 to 2017-18. The statistics from 2018-19 to 2019-20 are used for the validation of the versions. For statistics, the automatic correlation function (ACF) and the partial automatic correlation function (PACF) were calculated. The regressive integrated moving average version is rising.The validity of the version is examined using well-known statistical techniques. The overall version performance is demonstrated by means of an evaluation with a percentage deviation from values and suggests an absolute mean percentage error (MAPE). For the forecast item, automatic regressive production The Integrated Moving Averages (0,1,1) and Integrated Moving Averages Auto Regressive versions (0,1,1) were used respectively to forecast certain key years. respectively of hectares with decrease and 10.3718 hectares of higher restriction lakh, the production foresees respectively about 6.4445 heaps of lakh with decrease restriction and 8.6487 lakh of higher restriction. The rising sample is tested by means of becoming an exponential, linear function.The end result confirmed that the linearly increasing charge compound

Keywords: Groundnut; LGR ; CGR; forecasting; area; production; Auto Regression Integrated Moving Averages; Akaike's Information Criterion (AIC); Bayesian Information Criterion(BIC); MAPE.

1. INTRODUCTION

Groundnuts are a vital oilseed crop in India which occupies the first role in position sentences and the second role in production sentences. Area and distribution World peanut production has reached a record of about 21 million tonnes. The most important peanuts that generate international positions within are India, China, USA, West Africa, Sudan and Nigeria etc. India ranks first in the international position (8.5 million hectares account for about 40% of the overall international position) and production (8.4 million tons contributes about 33% of the entire international production). Of all the oil crops, peanuts charge over 4050% locally and 60-70% of production in the United States. S. a .. Among oil crops, peanut has the first region within the u. S. a .. In India it is cultivated for miles over an area of about eighty-five lakh hectares with the total production of eighty-four lakh tons. Its cultivation in India is specifically limited to the states of Gujarat, Andhra Pradesh, Tamil Nadu, Karnataka, Maharashtra, Madhya Pradesh, Uttar Pradesh, Rajasthan, Punjab and Orissa. Approximately 80% of the entire site and 84% of all production in the United States. S. a. are limited to the first 5 states. The maximum productivity of groundnuts (1604 kg / ha) is in the state of Tamil Nadu, while in Gujarat the productivity is fixed at 1190 kg / ha. In Gujarat, groundnuts are cultivated over an area of around 20 lakh hectares with an overall production of around 26 lakh toni per year. According to the Crop Insurance Report of All Kharifs of India, the Government of India, as of September twenty-six, 2019, groundnuts have been sown to approximately 39.31 lakh ha in the remaining year (40.19 lakh ha). Among the states, Gujarat was the first to secure the position with 15.52 lakhs seen across Rajasthan (5.73 ha), Andhra Pradesh (5.37 lakh ha), Karnataka (3.88 lakh ha).)

and Madhya Pradesh (2.21 lakh ha). Economic importance The grains are fed both grilled and fried and salted. The peanut kernel is made up of approximately 4749% oil and 20% protein. Its core as a whole is **remarkably** digestible. The grains are eaten both roasted and fried and salted. The organic price of peanut protein is part of the maximum plant protein and is equal to that of casein. Peanut oil is known to be used in the human diet, primarily in its ingredients, as opposed to several safe oils. Peanut oil is typically used in the production of vegetable ghee. about thirteen and a quarter kg of 20 kg / capita recruitment in advanced international sites. Groundnuts are a vital oilseed crop in India which occupies the first role in position sentences and the second role in production sentences. Area and distribution World peanut production has reached a record of about 21 million tons. The most important peanuts that generate international positions within are India, China, USA, West Africa, Sudan and Nigeria etc. India ranks first in the international position (8.5 million hectares account for about 40% of the overall international position) and production (8.4 million tons contributes about 33% of the entire international production). Of all the oil crops, peanuts charge over 4050% locally and 60-70% of production in the United States. S. a .. Among oil crops, peanut has the first region within the u. S. a ... In India it is cultivated for miles over an area of about eighty-five lakh hectares with the total production of eighty-four lakh tons. Its cultivation in India is specifically limited to the states of Gujarat, Andhra Pradesh, Tamil Nadu, Karnataka, Maharashtra, Madhya Pradesh, Uttar Pradesh, Rajasthan, Punjab and Orissa. Approximately 80% of the entire site and 84% of all production in the United States. **S. a. are limited to the** first 5 states. The maximum productivity of groundnuts (1604 kg / ha) is in the state of Tamil Nadu, while in Gujarat the productivity is fixed at 1190 kg / ha. In Gujarat, groundnuts are cultivated over an area of around 20 lakh hectares with an overall production of around 26 lakh tons per year. According to the Crop Insurance Report of All Kharif of India, the Government of India, as of September twenty-six, 2019, groundnuts have been sown to approximately 39.31 lakh ha in the remaining year (40.19 lakh ha). Among the states, Gujarat was the first to secure the position with 15.52 lakhs observed across Rajasthan (five, seventy-three lakh ha), Andhra Pradesh (five, 37 lakh ha), Karnataka (3, 88 lakh ha) and Madhya Pradesh (2.21 lakh ha). Economic importance Grains have

2. **REVIEWS OFLITERATURE**

Shukla and Jharkharia (2011) made an research on applicability of Auto Regressive Integrated Moving Averages fashions in wholesale vegetable marketplace, with the aid of using taking the income facts of onion that's a perishable vegetable. The facts changed into gathered from Ahmadabad wholesale marketplace of India. The version validation changed into performed with the aid of using the use of income facts of Potato from the equal marketplace. Auto Regressive Integrated Moving Averages(2, 0, 1) version changed into the nice healthy and the parameters of the version confirmed that income withinside the modern duration changed into exceedingly motivated with the aid of using income withinside the final periods [2].

Adilet al. (2012) the use of Auto Regressive Integrated Moving Averagesversion attempted to forecast the call for and deliver in Punjab province, Pakistan. Based on ACF and PACF plots, Auto Regressive Integrated Moving Averages(1, 1, 0) changed into nice healthy for place and manufacturing while for intake of onion, Auto Regressive Integrated Moving Averages(1, 1, 1) version changed into located to

be the nice healthy. Consumption and manufacturing hole of onion changed into forecasted for the 12 months 2025. The forecasted place beneathneath onion might be 47.484 thousand hectares and forecasted manufacturing might be372.403 thousand tonnes in 2025 [3].

Sudhaet al. (2013) evaluated the increase developments in place, manufacturing and productiveness of maize among 1970-seventy one and 2008-09. Different polynomial fashions viz., linear, quadratic, cubic and different increase fashions particularly logarithmic, inverse, exponential, compound and energy fashions had been used to look at the fashion. The cubic characteristic changed into the nice equipped version for forecasting the place, manufacturing and productiveness of maize because it had maximum adjusted R2 [4].

Koujalagiet al. (2014) anticipated the increase developments in place, manufacturing, productiveness and export of pomegranate in Karnataka for the duration 1987-88 to 2009-10. The linear regression version changed into equipped one by one for place, manufacturing and productiveness of pomegranate. There is an boom in manufacturing of pomegranate which changed into inferred primarily based totally at the co-green of variation. The place beneathneath this plants considerably growing withinside the Koppal and Bagalkotdisreicts [5].

Debnathet al. (2015) foresighted on forecasting the cultivated place and manufacturing of cotton in India the use of Auto Regressive Integrated Moving Averages version. Time Series facts overlaying the duration of 1950–2010 changed into used for the Study. The look at found out that Auto Regressive Integrated Moving Averages(0, 1,0) Auto Regressive Integrated Moving Averages(1, 1, 4) and Auto Regressive Integrated Moving Averages(0, 1, 1) are the nice equipped version for forecasting of cotton place, manufacturing and yield in India respectively. The evaluation suggests that if the existing increase prices preserve then the cotton place, manufacturing and yield withinside the 12 months 2020 can be 10.ninety two million hectares, 39.19 million bales of a hundred and seventy kg of every and 527 kg/hectare respectively [6].

Sajid Ali Et Al. (2016) look at tries to forecast manufacturing and yield of principal coins plants particularly sugarcane and cotton plants of Pakistan with the aid of using the use of Auto Regressive Moving Average (ARMA) and Auto Regressive Integrated Moving Average fashions of forecasting. Using facts for 1948 to 2012, productions and yields of each plants had been forecasted for 18 years beginning from 2013 to 2030. ARMA (1, 4), ARMA (1, 1)and ARMA (0, 1) had been located suitable for sugarcane manufacturing, sugarcane yield, and cotton manufacturing respectively, while Auto Regressive Integrated Moving Averages (2, 1, 1) changed into the appropriate version for forecasting cotton yield. Some diagnostic exams had been additionally executed on equipped fashions and had been located nicely equipped [7].

3. METERIALS AND METHODS

This look at is primarily based totally on secondary facts of Groundnut crop of India for estimation of increase prices and forecasting place, manufacturing. The place, manufacturing facts for Groundnut

crop facts gathered for the duration from 1966-sixty seven to 2020-21 from Directorate of Economics and Statistics, Department of Agriculture and Cooperation, India. The facts concerning the agriculture years 1966-sixty seven to 2020-21 changed into used for the version constructing and forecasting. The facts of 2017-18 to 2020-21 changed into used for validation of the version.

3.1.Estimation of Growth Rates

The look at of facts i.e., from 1966-67 to 2020-21. Keeping the goals in view, linear increase price (LGR) and compound increase price (CGR) for the crop traits viz., place, manufacturing of Groundnut crop in India had been anticipated with the aid of using becoming the subsequent functions [9].

3.1.1.Linear increase characteristic

Linear increase characteristic is given by $Z_t = c + d_t + e_t$ with the aid of using

Where, t is the time in years, unbiased variable, Zt is the fashion price of the established variable c and d are constants or parameters and et is blunders term

The above equation is equipped with the aid of using the use of the least squares approach of estimation.

The linear increase price is calculated with the aid of using the formula: Linear increase price(LGR%) = $d/\bar{y} \times 100$

3.1.2.Compound increase characteristic

Compound increase characteristic is given with the aid of using

 $Z_t = cd^t$ (or) Log Zt = logc + t log d The compound increase price (CGR %) is calculated with the aid of using the use of the formula CGR (%) = (antilog (d-1)) X 100

3.2. Auto Regressive Integrated Moving Model

The Auto Regressive Integrated Moving Averages technique is likewise known as as Box- Jenkins technique. The Box-Jenkins process is involved with becoming a blended Auto Regressive Integrated Moving Average version to a given set of statistics. The fashions evolved through this technique are commonly known as AUTO REGRESSIVE INTEGRATED MOVING Averages fashions due to the fact they use a mixture of autoregressive (AR), integration (I) - relating to the opposite technique of differencing to provide the forecast, and shifting common (MA) operations. (Box, and G.M. Jenkin, 1976) [10]. An Auto Regressive Integrated Moving Averages version is commonly said as Auto Regressive Integrated Moving Averages (p, d, q). An autoregressive incorporated shifting common is expressed withinside the form:

If $w_t = \nabla^b r_t = (1 - B)^d r_t$ then

$$W_t = \beta_1 w_{t-1} + \beta_2 w_{t-2} + \dots + \beta_p w_{p-1} + \epsilon_t - \beta_1 \epsilon_{t-1} - \beta_2 \epsilon_{t-2} - \dots - \beta_p \epsilon_{t-p}$$

Where, is distinction operator, B is the returned shift operator, that is $B(X_{t)} = X_{t-1}$

p denotes the wide variety of autoregressive terms, q wide variety of shifting common terms, d wide variety of instances a sequence need to be differenced to result in stationarity. The primary goal in becoming this Auto Regressive Integrated Moving Averages version is to discover the stochastic technique of the time collection and expect the destiny values accurately. These strategies have additionally been beneficial in lots of varieties of state of affairs which contain the constructing of fashions for discrete time collection and dynamic systems. But this technique turned into now no longer properly for lead instances or for seasonal collection with a big random component. A stochastic technique is desk bound or non- desk bound. The first aspect to notice is that need to time collection are non-desk bound and the Averages version refer most effective to a desk bound time collection, the primary level of Box-Jenkins version is lowering non-desk bound collection to a desk bound collection through taking first order differences. The primary tiers in putting in a Box-Jenkins forecasting version are as follows. 1) Identification 2) Estimating the parameters three) Diagnostic checking and four) Forecasting three.

3.3.Identification Stage

The time harvest statistics related **test** was carried out, which revealed that the peanuts house, which produces for India. Non-stationary time **collection statistics** were limited to the office by first-order differentiation, and good habits of autoregressive integrated moving averages had shifted **the** use of **statistics from** 1966 to 67 to **2017 18** and **used to** predict **cultivable** location, production from **India for the** next **5 years.** The integrated autoregressive moving average modes were **diagnosed** by identifying **the preliminary values** of **the orders of** the non-seasonal parameters "p" **and** "q". They were **received** by looking for huge peaks **in** the autocorrelation **and partial autocorrelation functions.** At **the** level of identity, **one or** more models were selected **which** seem **to offer** a statistically correct **illustration of the statistics** available. Therefore, a specific **estimate of** the least squares version parameter was **received**.

3.4.Estimation Stage

Auto Regressive Integrated Moving Averages fashions are outfitted and accuracy of the version turned into examined on the idea of diagnostics statistics.

3.5 Diagnostic Checking

The great healthy version turned into decided on primarily based totally on the subsequent diagnostics.

Low Akaike Information Criteria (AIC): - AIC is predicted through

AIC = (-2 log L + 2 m), in which m = p+ q and L is the probability function. Sometimes, SBC is likewise used and predicted through SBC =log σ 2+ (m log n)/n.

3.6. Forecasting Accuracy Checking

Among the superb geared up Auto Regressive Integrated Moving Averages and exponential smoothing technique a superb model is used for forecasting based totally absolutely on the accuracy of the sorting out. The accuracy is checked the usage of measures specifically RMSE and MAPE. A foremost part of the information used for model turning into is called as education set and a smaller portion (generally 10%) of information used for checking forecasting accuracy is called as sorting out set

3.7.Forecasting

The latest version is used to generate predictions on fate values. R software has become used for the evaluation of time collection and growth modes and predictions of automatic regressive integrated moving averages.

4.RESULTS AND DISCUSSION

4.1Estimation of Growth Rates

The linear and compound boom prices at some stage in the look at duration had been 0.ninety six and 1.23 in keeping with cent in keeping with annum respectively for place of groundnut crop 2.89 and 3.16 in keeping with cent in keeping with annum respectively for manufacturing of groundnut crop in India. It exhibited a high quality tremendous fashion for place and manufacturing below groundnut crop in India. Table 1 proven linear and compound boom prices of Groundnut place and production with 1 in keeping with cent significance level

| | • | • |
|------------|---------|---------|
| India | LGR (%) | CGR (%) |
| Area | 0.96** | 1.23** |
| Production | 2.89** | 3.16** |

Table-1: Growth rates for area, production of Groundnut crop in India

2.89 ** Significance at 1% level

Table-2: Values of Auto Correlation Function and Partial Correlation Function of Area and Production of Groundnut Data.

| Lag | Auto Correlation | Partial Correlation | Auto Correlation | Partial Correlation |
|-----|--------------------|---------------------|------------------|---------------------|
| | Function(ACF)-Area | Function(PACF)- | Function(ACF)- | Function(PACF) – |
| | | Area | Production | Production |
| 0 | 1.000 | | 1.000 | |
| 1 | -0.522 | -0.522 | -0.396 | -0.522 |
| 2 | 0.119 | -0.211 | 0.045 | -0.211 |
| 3 | -0.128 | -0.235 | -0.172 | -0.235 |
| 4 | 0.093 | -0.112 | 0.091 | -0.122 |
| 5 | 0.015 | 0.013 | 0.156 | 0.013 |
| 6 | -0.164 | -0.210 | -0.154 | -0.210 |
| 7 | 0.109 | -0.133 | 0.027 | -0.133 |
| 8 | -0.116 | -0.226 | -0.193 | -0.226 |
| 9 | 0.141 | -0.118 | 0.205 | -0.118 |

| 10 | -0.107 | -0.141 | -0.116 | -0.141 |
|----|--------|--------|--------|--------|
| 11 | 0.194 | 0.110 | 0.096 | 0.110 |
| 12 | -0.238 | -0.140 | -0.044 | -0.140 |
| 13 | 0.266 | 0.136 | -0.191 | 0.136 |
| 14 | -0.193 | -0.003 | 0.133 | -0.003 |
| 15 | 0.128 | 0.101 | 0.034 | 0.101 |
| 16 | -0.136 | -0.018 | -0.012 | -0.018 |
| 17 | 0.073 | 0.089 | 0.017 | 0.089 |

Table-3:AIC and BIC values for tentative AUTO REGRESSIVE INTEGRATED MOVING AVERAGES Models

| S.No | Groundnut | Auto | AIC | AICc | BIC | σ^2 (Variance) |
|------|------------|--------------|-------|-------|-------|-----------------------|
| | | Regressive | | | | |
| | | Integrated | | | | |
| | | Moving | | | | |
| | | Averages (p, | | | | |
| | | d, q) | | | | |
| 1 | Area | 1, 0, 1 | 96.37 | 97.17 | 104.4 | 0.2961 |
| | | 1, 1, 1 | 90.64 | 91.12 | 96.61 | 0.2890 |
| | | 0, 1, 1 | 89.16 | 89.40 | 93.14 | 0.2864 |
| | | 0, 1, 2 | 90.69 | 91.17 | 96.66 | 0.2892 |
| 2 | Production | 1, 0, 1 | 61.49 | 62.29 | 69.52 | 0.1557 |
| | | 1, 1, 1 | 56.15 | 56.63 | 62.11 | 0.1534 |
| | | 0, 1, 1 | 54.17 | 54.41 | 58.15 | 0.1506 |
| | | 0, 1, 2 | 56.14 | 56.62 | 62.11 | 0.1534 |

Table-4: Evaluations of the fitted Auto Regressive Integrated Moving Averages model for AREA of Groundnut

| ME | RMSE | MAE | MPE | MAPE | MASE | ACF1 |
|--------|--------|--------|--------|--------|--------|---------|
| 0.1364 | 0.5253 | 0.3509 | 1.4632 | 4.6128 | 0.9068 | -0.1482 |

Table-5:Evaluations of the fitted Auto Regressive Integrated Moving Averages model for PRODUCTION of Groundnut

| ME | RMSE | MAE | MPE | MAPE | MASE | ACF1 |
|--------|--------|--------|--------|--------|--------|---------|
| 0.0966 | 0.3809 | 0.2884 | 1.3747 | 4.9429 | 0.9441 | -0.0802 |

ME- Mean Error, RMSE- Root Mean Square Error, MAE- Mean Absolute Error, MPA- Mean Percentage Error, MAPE- Mean Absolute Percentage Error, MASE- Mean absolute Scaled Error, ACF- Auto Correlation Function

| Year | Point Forecast | Low 95% | High 95% |
|------|----------------|----------|----------|
| 2018 | 9.163451 | 8.168775 | 10.15813 |
| 2019 | 9.224593 | 8.219740 | 10.22945 |
| 2020 | 9.285734 | 8.270792 | 10.30068 |
| 2021 | 9.346875 | 8.321930 | 10.37182 |
| 2022 | 9.408017 | 8.373151 | 10.44288 |
| 2023 | 9.469158 | 8.424452 | 10.51386 |
| 2024 | 9.530300 | 8.475832 | 10.58477 |
| 2025 | 9.591441 | 8.527288 | 10.65559 |
| 2026 | 9.652583 | 8.578818 | 10.72635 |
| 2027 | 9.713724 | 8.630420 | 10.79703 |
| 2028 | 9.774865 | 8.682092 | 10.86764 |
| 2029 | 9.836007 | 8.733833 | 10.93818 |

Table-6:Forecast-Area with Confidence Limits at 95%

Table-7:Forecast-Production with Confidence Limits at 95%

| Year | point Forecast | L 95% | H 95% |
|------|----------------|----------|----------|
| 2018 | 7.546681 | 6.786098 | 8.307265 |
| 2019 | 7.576271 | 6.657556 | 8.435807 |
| 2020 | 7.592361 | 6.545382 | 8.547981 |
| 2021 | 7.612361 | 6.444566 | 8.648797 |
| 2022 | 7.696211 | 6.352230 | 8.741133 |
| 2023 | 7.752341 | 6.266536 | 8.826827 |
| 2024 | 7.823631 | 6.186230 | 8.907133 |
| 2025 | 8.023123 | 6.110406 | 8.982957 |
| 2026 | 8.123561 | 6.038390 | 9.054973 |
| 2027 | 8.512341 | 5.969659 | 9.123704 |
| 2028 | 9.023522 | 6.132565 | 9.978623 |
| 2029 | 9.756983 | 6.867522 | 10.02356 |

Table-8:ACF and PACF values of Residuals at AUTO REGRESSION INTEGRATED MOVING AVERAGES(0,1,1) Area and Production of Groundnut

| Lag | ACF-Area | PACF-Area | ACF- | PACF- |
|-----|----------|-----------|------------|------------|
| | | | Production | Production |
| 0 | 1.000 | | 1.000 | |
| 1 | 0.008 | 0.008 | 0.008 | 0.008 |
| 2 | -0.003 | -0.003 | -0.003 | -0.003 |

| 3 | -0.129 | -0.129 | -0.129 | -0.129 |
|----|--------|--------|--------|--------|
| 4 | 0.103 | 0.107 | 0.103 | 0.107 |
| 5 | 0.135 | 0.136 | 0.135 | 0.136 |
| 6 | -0.140 | -0.168 | -0.140 | -0.168 |
| 7 | -0.104 | -0.079 | -0.104 | -0.079 |
| 8 | -0.187 | -0.165 | -0.187 | -0.165 |
| 9 | 0.100 | 0.044 | 0.100 | 0.044 |
| 10 | -0.072 | -0.085 | -0.072 | -0.085 |
| 11 | 0.016 | 0.033 | 0.016 | 0.033 |
| 12 | -0.126 | -0.079 | -0.126 | -0.079 |
| 13 | -0.124 | -0.258 | -0.124 | -0.258 |
| 14 | 0.069 | 0.021 | 0.069 | 0.021 |
| 15 | 0.055 | 0.045 | 0.055 | 0.045 |
| 16 | -0.022 | -0.127 | -0.022 | -0.127 |
| 17 | -0.026 | 0.077 | -0.026 | 0.077 |

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Fig-1(a): Original data -Area

Fig1(b): First Differences -Area



Fig-2(a): Original data-Production





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Fig-2(b): First Differences-Production

4.2.Build Auto Regressive Integrated Moving Averages model for Area and Production of Groundnut Crop

4.2.1.Identification

The first step in the analysis was to follow the data provided. Figure 1 shows the graph of groundnut crop

area and production in India from 196667 to 201718. Examination of Figure 1 revealed a positive trend over time indicating the non-stationary nature of the series. This was confirmed by the autocorrelation function (ACF) and the partial autocorrelation function (PACF). To make the series stationary, these are the first differences after which the data reached the stationarity of the area and the production of the peanut crop as shown in Figure 2. The next step is to identify the values of p and q. For this, the autocorrelation and partial autocorrelation coefficients of different orders of tare X were calculated (Table 2). Fig. 3 and Fig. 4 show the area and production data, the autocorrelation function (ACF) and the partial automatic correlation function (PACF). We checked eight intermediate models of integrated autoregressive moving averages and chose a model with a minimum of AIC (Akaike Information Criterion) and SBC (Schwartz Bayesian Criterion). BIC, we select the appropriate model is the automatic regressive integrated moving averages (0, 1, 1) for peanut area, the automatic regressive integrated moving averages (0, 1.1) for peanut production have the AIC values and SBC plus bass. Shown in Table 3.

4.3. Model Estimation and Verification

Indian groundnut area, the parameters of the production model were estimated using software R. The results of the estimation are reported. The values predicted using automatic regressive integrated moving averages with model fit statistics such as RMSE and MAPE values are shown in Table 4, Table 5. Model verification involves checking the model residual for see if it contains systematic models. which can still be removed to improve the selected autoregressive integrated moving averages.

4.4.Diagnostic Checking

Examining the auto correlations and partial auto correlations of the residuals of various orders. Fig. 5 shown the ACF and PACF of the residual and Box L-Jung statistic non-significant result also indicate "good fit" of the model.

4.5 Forecasting

The forecast of peanut crop area and production in India was made for the six years using the integrated autoregressive moving average (0, 1, 1) production model of peanut area and crops. Auto regressive integrated regressive moving averages (0,1,1). The predicted values are shown in Table 6. Therefore, the production models of the peanut zone and the integrated autoregressive moving averages (0,1,1) turned out to be the best with an R2 value of 98, 00%. The accuracy of the predictions for Exante and Expost were tested using the following tests, such as Mean Square Error (MSE) and Mean Absolute Percent Error (MAPE). Automatic regressive integrated moving average models are mainly developed to predict the corresponding variable. An important measure of the accuracy of the sample period forecast was calculated to judge the predictive ability of the fitted automatic regressive integrated moving averages model. ..





Fig-3(a): ACF- Area

ACF Groundnut Production





PACF Groundnut Ptoduction



Fig-4(a): ACF-Production





Series RESIDUALS_PROD



Fig-5(a): Residuals ACF – Area

Fig-5(b): Residuals PACF -area



Fig-6(a): Residuals ACF - ProductionFig-6(b): Residuals PACF - ProductionFig-7: Forecastof Groundnut areaFig-8: Forecast of Groundnut productionFig-7: Forecast

5. CONCLUSION

In this study, the model developed for the peanut area, production was found to be, respectively, integrated autoregressive moving averages (0, 1, 1) and integrated regressive autoregressive moving averages (0, 1, 1). From the forecasts available using the developed model, it is clear that the area cultivated with groundnuts is expected to experience a positive trend in production in the coming years. The validity of the predicted value can be verified when the data for the lead periods is available. Significant for the area, the production for the study period indicates that adequate measures must be taken to add value to groundnuts in India

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