

Design and Development of ARIMA Model for Bajra (*Pennisetum glaucum*) Production in India

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Abstract

This study deals with design and development of autoregressive integrated moving average (ARIMA) model for Bajra (*Pennisetum glaucum*) production in India based on Bajra (*P. glaucum*) production during the years from 1951 to 2018. The study considers Autoregressive (AR), Moving Average (MA) and Autoregressive Integrated Moving Average (ARIMA) processes to select the appropriate ARIMA model for Bajra (*P. glaucum*) production in India. Based on ARIMA (p, d, q) and its components Autocorrelation Function (ACF), Partial Autocorrelation Function (PACF), Root Mean Square Error (RMSE), Mean Absolute Percentage Error (MAPE), Normalized BIC and Box-Ljung Q statistics estimated, ARIMA (0,1,1) was selected. Based on the chosen model, it could be predicted that Bajra (*P. glaucum*) production would increase from 9.21 million tons in 2018 to 10.52 million tons in 2025 in India.

Key Words: ARIMA, BIC, Forecasting, Bajra Production.

Introduction

Bajra is a widely grown kind of millet. The botanical name of Bajra is

Pennisetum glaucum. This crop has been grown in the Indian subcontinent and Africa since prehistoric times.

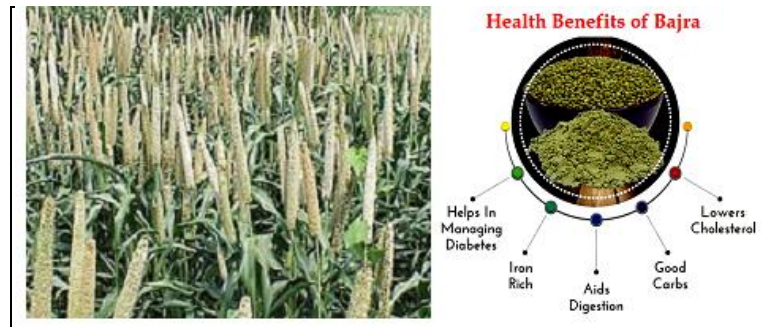


Figure 1 - Health Benefits of Bajra

India is the largest producer of Bajra. India began growing bajra between 1500 and 1100 BCE. Kambu is the Tamil name of Bajra and is an important food across the Indian state of Tamil Nadu. It is the second important food for Tamil people consumed predominantly in the hot humid summer months from February through May every year. It is made into a gruel and consumed along with buttermilk or consumed as dosa or idly. The potential health benefits associated with regularly eating bajra are weight loss, improved diabetes management, and a higher intake of nutrients that support healthy hair, nails, and skin (Figure 1). However, Bajra is a good source of many nutrients known for contributing to healthy hair, skin, and nails, including protein, vitamin B6, niacin folate, iron and zinc.

Material and Methods

As the aim of the study was to implement of ARIMA model for Baajra (*P. glaucum*) production in India, various forecasting techniques were considered for use. ARIMA model, introduced by Box and Jenkins (1976), was frequently applied for discovering the pattern and predicting the future values of the time series data. Box and Pierce (1970) measured the distribution of residual autocorrelations in ARIMA. Akaike (1970) found the stationary time series by an AR (p), where p is finite and bounded by the same integer. Moving Average (MA) models were applied by Slutzky (1973). Md. Moyazzem Hossain and Faruq Abdulla (2016) applied ARIMA (0,2,1) model for yearly potato production in Bangladesh for the period from 1971 to 2013. Borkar et al. (2016) found that ARIMA (2,1,1) is the appropriate model for

forecasting the production of cotton in India. Kour et al., (2017) applied ARIMA model for forecasting of productivity of pearl millet of Gujarat for the period from 1960-61 to 2011-12 and validated ARIMA (0,1,1) model performs quite satisfactorily as the RMAPE value is less than 6 percent. Hemavathi and Prabakaran (2018) found rice production data during 1990-2015 and applied ARIMA (0,1,1) model up to 2020. Bhola Nath et al. (2019) discovered ARIMA (1,1,0) model for wheat production in India for the period from 1949-50 to 2016-17 and forecasted up to 2026-27.

Stochastic time-series ARIMA models were widely used in time series data which are having the characteristics (Alan Pankratz, 1983) of parsimonious, stationary, invertible, significant estimated coefficients and statistically independent and normally distributed residuals. ARIMA model was used in this study, which required a sufficiently large data set and involved four steps: identification, estimation, diagnostic checking and forecasting. Model parameters were estimated to fit the ARIMA models.

Autoregressive process of order (p) is,

$$Y_t = \mu + \phi_1 Y_{t-1} + \phi_2 Y_{t-2} + \dots + \phi_p Y_{t-p} + \varepsilon_t;$$

Moving Average process of order (q) is,

$$Y_t = \mu - \theta_1 \varepsilon_{t-1} - \theta_2 \varepsilon_{t-2} - \dots - \theta_q \varepsilon_{t-q} + \varepsilon_t; \text{ and}$$

The general form of ARIMA model of order (p, d, q) is

$$Y_t = \phi_1 Y_{t-1} + \phi_2 Y_{t-2} + \dots + \phi_p Y_{t-p} + \mu - \theta_1 \varepsilon_{t-1} - \theta_2 \varepsilon_{t-2} - \dots -$$

where Y_t is Bajra (*P. glaucum*) production, ε_t 's are independently and normally distributed with zero mean and constant variance σ^2 for $t = 1, 2, \dots, n$; d is the fraction differenced while interpreting AR and MA and ϕ s and θ s are coefficients to be estimated.

Trend Fitting : The Box-Ljung Q statistics was used to transform the non-stationary data into stationarity data and also to check the adequacy for the residuals. For evaluating the adequacy of AR, MA and ARIMA processes, various reliability statistics like R^2 , Stationary R^2 , Root Mean Square Error (RMSE), Mean Absolute Percentage Error (MAPE), and BIC were used. The reliability statistics viz. RMSE, MAPE, BIC and Q statistics were computed as below:

$$RMSE = \left[\frac{1}{n} \sum_{i=1}^n (Y_i - \hat{Y}_i)^2 \right]^{1/2};$$

$$MAPE = \frac{1}{n} \sum_{i=1}^n \left| \frac{Y_i - \hat{Y}_i}{Y_i} \right| \text{ and } BIC(p,q) = \ln$$

$$v^*(p,q)+(p+q) [\ln(n) / n]$$

where p and q are the order of AR and MA processes respectively and n is the number of observations in the time series and v^* is the estimate of white noise variance σ^2 .

$$Q = \frac{n(n+2) \sum_{i=1}^k r_k^2}{(n-k)}$$

where n is the number of residuals and r_k is the residuals autocorrelation at lag k .

In this study, the data on Bajra (*P. glaucum*) production in India were collected from the Annual Report (2018), Directorate of Economics and Statistics, Department of Agriculture, Cooperation and Farmers Welfare, Ministry of Agriculture and

Farmers Welfare, Government of India for the period from 1951 to 2018 (Table 1) and were used to fit the ARIMA model to predict the future production.

Results and Discussion

In this study, the data for Bajra production in India is collected from the period 1951 to 2018 is given in Table 1. To fit an Autoregressive model, Autoregressive process for any variable involves four steps: identification, estimation, diagnostic and forecasting. ARIMA (p,d,q) model is fitted to check stationarity through examining the graph or time plot of the data. Figure 2 reveals that the data is non-stationary.

Table 1. Actual Bajra (*P. glaucum*) production (million tons) in India

Year	Production	Year	Production	Year	Production	Year	Production
1951	2.60	1969	3.80	1987	4.51	2005	7.93
1952	2.35	1970	5.33	1988	3.30	2006	7.68
1953	3.19	1971	8.03	1989	7.78	2007	8.42
1954	4.55	1972	5.32	1990	6.65	2008	9.97
1955	3.52	1973	3.93	1991	6.89	2009	8.89
1956	3.43	1974	7.52	1992	4.67	2010	6.51
1957	2.87	1975	3.27	1993	8.88	2011	10.37
1958	3.62	1976	5.74	1994	4.97	2012	10.28
1959	3.87	1977	5.85	1995	7.16	2013	8.74

1960	3.49	1978	4.73	1996	5.38	2014	9.25
1961	3.28	1979	5.57	1997	7.87	2015	9.18
1962	3.65	1980	3.95	1998	7.64	2016	8.07
1963	3.96	1981	5.34	1999	6.96	2017	9.73
1964	3.88	1982	5.54	2000	5.78	2018	9.21
1965	4.52	1983	5.13	2001	6.76		
1966	3.75	1984	7.72	2002	8.28		
1967	4.47	1985	6.05	2003	4.72		
1968	5.19	1986	3.66	2004	12.11		

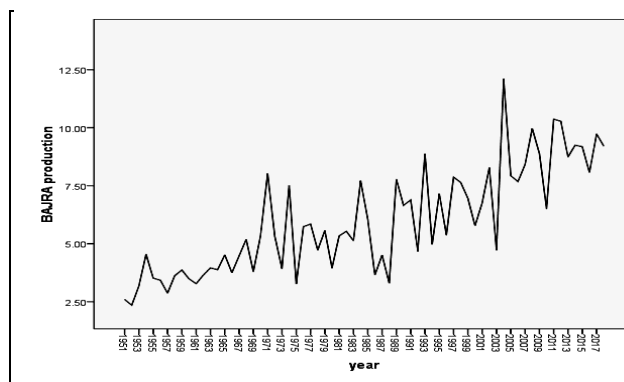


Figure 2 - Time plot of Bajra (*P. glaucum*) Production

The autocorrelation and partial autocorrelation coefficients of various orders of Y_t are computed Table 2. The graphs of ACF and PACF are given in Figure 3.

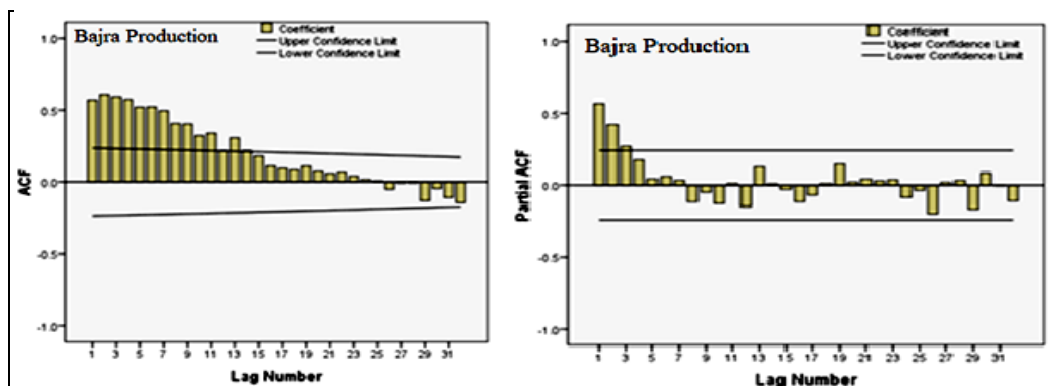


Figure 3 ACF and PACF of differenced data

Table 2 - ACF and PACF of Bajra (*P. glaucum*) Production

Lag	AC	Std. Error ^a	Box-Ljung Statistic	PAC	Std. Error	Lag	AC	Std. Error ^a	Box-Ljung Statistic	PAC	Std. Error
	Value	Df	Sig. ^b	Value	Df		Value	Df	Sig. ^b	Value	Df
1	0.569	0.119	23.008	0.569	0.121	17	0.098	0.104	226.404	-0.070	0.121
2	0.607	0.118	49.554	0.418	0.121	18	0.088	0.102	227.135	0.012	0.121
3	0.591	0.117	75.155	0.273	0.121	19	0.114	0.101	228.394	0.146	0.121
4	0.573	0.116	99.580	0.175	0.121	20	0.076	0.100	228.970	0.022	0.121
5	0.521	0.115	120.065	0.043	0.121	21	0.056	0.099	229.293	0.042	0.121
6	0.522	0.114	140.998	0.061	0.121	22	0.068	0.098	229.775	0.031	0.121
7	0.495	0.113	160.152	0.035	0.121	23	0.037	0.097	229.922	0.038	0.121
8	0.407	0.112	173.266	-0.118	0.121	24	0.015	0.096	229.947	-0.086	0.121
9	0.404	0.111	186.412	-0.049	0.121	25	0.005	0.095	229.950	-0.036	0.121
10	0.323	0.110	194.987	-0.129	0.121	26	-0.051	0.094	230.242	-0.199	0.121
11	0.340	0.109	204.654	0.012	0.121	27	-0.008	0.093	230.250	0.019	0.121
12	0.221	0.108	208.790	-0.149	0.121	28	-0.008	0.092	230.256	0.034	0.121
13	0.308	0.107	216.979	0.127	0.121	29	-0.128	0.091	232.250	-0.167	0.121
14	0.221	0.107	221.270	0.010	0.121	30	-	0.089	232.509	0.085	0.121

							0.045				
15	0.184	0.106	224.298	- 0.030	0.121	31	- 0.106	0.088	233.968	- 0.003	0.121
16	0.115	0.105	225.501	- 0.116	0.121	32	- 0.141	0.087	236.583	- 0.111	0.121

^aThe underlying process assumed is independence (white noise).

^bBased on the asymptotic chi-square approximation.

The models and corresponding BIC values are given in Table 3. The value of normalized BIC is 0.807 and R Squared value is 0.654. So the most suitable model for Bajra Production is ARIMA(0,1,1) as this model has the lowest BIC value.

Model Estimation: Model parameters were estimated and reported in Table 3 and Table 4. The model verification is concerned with checking the residuals of

the model to improve on the chosen ARIMA (p,d,q). This is done through examining the autocorrelations and partial autocorrelations of the residuals of various orders, up to 32 lags were computed and the same along with their significance which is tested by Box-Ljung test are provided in Table 5. This proves that the selected ARIMA model is an appropriate model.

Table - 2 BIC values of ARIMA(p,d,q)

ARIMA (p,d,q)	BIC Values
0,1,0	1.599
0,1,1	0.807
0,1,2	0.828
1,1,0	1.233
1,1,1	0.829
1,1,2	0.908
2,1,0	1.127

2,1,1	0.905
2,1,2	0.985
3,1,0	1.109
3,1,1	0.985
3,1,2	1.091

Table 3 - Estimated AR Model of Bajra (*P. glaucum*) Production

	Estimate	SE	t	Sig.
Constant	-2.118	1.873	-1.131	0.262
MA 1	0.994	0.480	2.072	0.042

Table 4 - Estimated AR Model Fit Statistics

ARIMA (p,d,q)	Stationary R ²	R ²	RMSE	MAPE	MaxAPE	MAE	MaxAE	Normalized BIC
0,1,0	0.000	0.174	2.089	26.125	133.056	1.519	7.296	1.599
0,1,1	0.581	0.654	1.362	17.051	85.595	0.953	4.398	0.807
0,1,2	0.604	0.673	1.334	16.936	99.218	0.937	3.794	0.828
1,1,0	0.358	0.47	1.686	21.434	96.773	1.234	5.127	1.233
1,1,1	0.604	0.673	1.335	16.974	96.51	0.94	3.742	0.829
1,1,2	0.604	0.673	1.346	17.132	101.383	0.944	3.738	0.908
2,1,0	0.467	0.56	1.549	19.619	81.019	1.106	4.819	1.127
2,1,1	0.605	0.674	1.344	16.973	100.966	0.936	3.716	0.905
2,1,2	0.605	0.674	1.356	16.974	99.402	0.941	3.736	0.985
3,1,0	0.516	0.6	1.488	19.256	85.948	1.051	4.986	1.109
3,1,1	0.605	0.674	1.356	16.99	98.818	0.939	3.779	0.985
3,1,2	0.594	0.665	1.385	18.163	93.688	0.997	3.915	1.091

The ACF and PACF of the residuals are given in Figure 4. It also indicates ‘good fit’ of the model. So the fitted ARIMA model for the Bajra Production data is

$$Y_t = \mu - \theta_1 \varepsilon_{t-1} + \varepsilon_t$$

$$Y_t = -2.118 - 0.994\varepsilon_{t-1} + \varepsilon_t$$

Forecasting: Forecasted value of Bajra Production (Quantity in numbers) for the year 2019 through 2025 respectively given by 9.70, 9.83, 9.97, 10.11, 10.24, 10.38 and 10.52 are given in Table 6. To assess the forecasting ability of the fitted ARIMA (p,d,q) model, important

measures of the sample period forecasts’ accuracy were computed. This measure indicates that the forecasting inaccuracy is low. Figure 5 shows that the actual and forecasted value of Bajra Production data with 95% confidence limits.

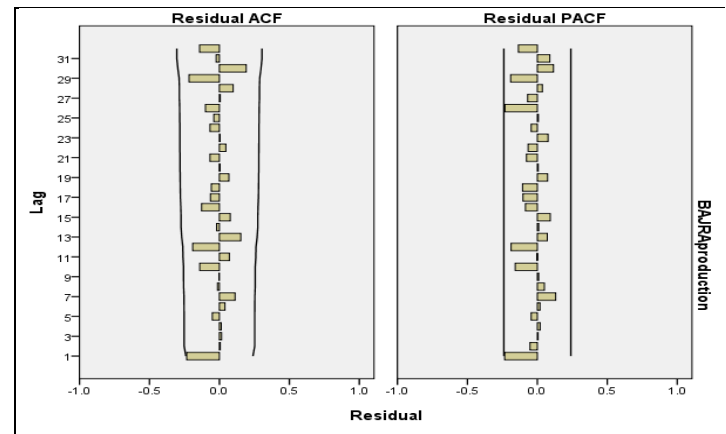
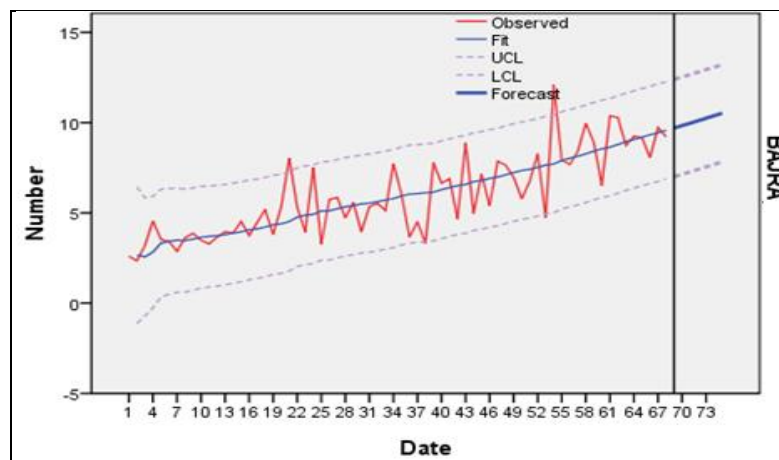


Figure 4 - Residuals of ACF and PACF

Table - 5 Residual of ACF and PACF of Bajra (*P. glaucum*) Production

Lag	ACF		PACF		Lag	ACF		PACF	
	Mean	SE	Mean	SE		Mean	SE	Mean	SE
1	-0.234	0.122	-0.234	0.122	17	-0.064	0.142	-0.103	0.122
2	0.006	0.129	-0.052	0.122	18	-0.057	0.143	-0.105	0.122
3	0.016	0.129	0.005	0.122	19	0.068	0.143	0.073	0.122
4	0.013	0.129	0.020	0.122	20	0.004	0.144	0.006	0.122
5	-0.050	0.129	-0.044	0.122	21	-0.067	0.144	-0.077	0.122
6	0.040	0.129	0.019	0.122	22	0.046	0.144	-0.066	0.122

7	0.112	0.129	0.131	0.122	23	0.005	0.144	0.076	0.122
8	-0.014	0.131	0.050	0.122	24	-0.067	0.144	-0.044	0.122
9	-0.004	0.131	0.009	0.122	25	-0.040	0.145	0.007	0.122
10	-0.140	0.131	- 0.159	0.122	26	-0.100	0.145	-0.231	0.122
11	0.071	0.133	- 0.001	0.122	27	0.006	0.146	-0.069	0.122
12	-0.191	0.133	- 0.187	0.122	28	0.098	0.146	0.035	0.122
13	0.153	0.137	0.072	0.122	29	-0.217	0.147	-0.190	0.122
14	-0.020	0.140	0.009	0.122	30	0.192	0.152	0.115	0.122
15	0.078	0.140	0.093	0.122	31	-0.023	0.155	0.089	0.122
16	-0.127	0.141	- 0.085	0.122	32	-0.141	0.155	-0.136	0.122

Figure 5 - Actual and Estimate of Bajra (*P. glaucum*) ProductionTable 6 - Forecast of Bajra (*P. glaucum*) Production

Year	Predicted	LCL	UCL
2019	9.70	7.01	12.39

2020	9.83	7.14	12.53
2021	9.97	7.28	12.66
2022	10.11	7.41	12.80
2023	10.24	7.55	12.94
2024	10.38	7.69	13.07
2025	10.52	7.83	13.21

Conclusion

The most appropriate ARIMA model for Bajra (*P. glaucum*) production forecasting of data was found to be ARIMA (0,1,1). From the time series data, it can be found that forecasted production would increase to 10.52 million tons in 2025 from 9.1 million tons in 2018 in India for using time series data from 1951 to 2018 on Bajra (*P. glaucum*) production, this study provides an evidence on future Bajra (*P. glaucum*) production in the country, which can be considered for future policy making and formulating strategies for augmenting and sustaining Bajra (*P. glaucum*) production in India.

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