

An Application of Integrative Approach of Time Based Data – Application of ARIMA Model for Dairy Sector of Gujarat State

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ABSTRACT

Time is the factor which affects the entire planet. Adequacy to the relevant changes according to the time dilemma can be maintained and forecasting for the stated data used for future glimpses. This research is presented with the integrated impact of autoregressive and moving average based to the exogenous variables to the X specified vectors. The relevancy of the time matrix to the dependent and independent study is derived. The ARIMAX is applied to the dairy sector data of Gujarat state. The predicted model can result out for future prediction or estimations of each of the parameter.

1. INTRODUCTION:

There are two principle objectives of time arrangement investigation: (a) recognizing the idea of the marvel spoken to by the grouping of perceptions, and (b) estimating or foreseeing future estimations of the time arrangement variable. Both of these objectives necessitate that the example of watched time arrangement information is distinguished and pretty much formally depicted. When the example is set up, we can decipher and coordinate it with other information. Notwithstanding the profundity of our comprehension and the legitimacy of our translation of the wonder, it can extrapolate the recognized example to anticipate future occasions.

Most time arrangement examples can be portrayed as far as two essential classes of parts: pattern and regularity. The previous speaks to a general deliberate direct or nonlinear segment that changes after some time and does not rehash or if nothing else does not rehash inside the time go caught by our information. The last may have a formally comparable nature. In any case, it rehashes itself in efficient interims after some time. Those two general classes of time arrangement parts may coincide, in actuality, information.

2. AUTOREGRESSIVE-INTEGRATED MOVING-AVERAGE (ARIMA)MODEL:

In the statistical analysis of time series autoregressive moving average (ARIMA) model provide a miserly explanation of

a inert random procedure in relations of two models, one for the AR and the second is MA. The common ARIMA model has defined in the 1952 in research thesis of Whittle, and was interpreted in the 1970 by Box and Jenkins. Given a time series of data Y_{it} , the ARIMA model is a tool for considerate for estimating future values of time series. It contains of two measures, AR and MA. The model is usually then mentioned as ARIMA (p, q) model here p is the order of the AR term and q is the order of the MA term.

3. AUTOREGRESSIVE-MOVING-AVERAGE MODEL

The presentation in time series of ARIMA (p, q) shows that the defined model is an integrated model with p order AR terms and q ordered, MA terms. Thus it has order of (p, q),

$$Y_{it} = A + \epsilon_i + \sum_{i=0}^p \theta_i X_{i-1} + \sum_{i=1}^q \rho_i \epsilon_{i-1}$$

The first application of ARIMA was given by Whittle and was popularized in 1970 by Box and Jenkins. This model is useful to test the models for lower order polynomials. The error ϵ_i expected to be i.i.d and consider for $\epsilon_i \sim N(0, \sigma^2)$.

Table 1.1: ARIMA, using observations 1-650
Dependent variable: MP

	Coefficient	SE	z	p-value	
Const.	-3.17	14.7	-0.2157	0.8293	
Row_1	0.868	0.0234	37.19	<0.0001	***
Theta_1	0.0073	0.043	0.1696	0.8653	
TB_POP	-0.0263	0.059	-0.4402	0.6598	
FA_POP	0.536	0.055	9.743	<0.0001	***
BR_POP	-0.0589	0.124	-0.4749	0.6349	
IMA_POP	0.907	0.139	6.519	<0.0001	***
MI_A	-0.271	0.102	-2.659	0.0078	***
MA	-0.361	0.074	-4.859	<0.0001	***
SR	0.647	0.172	3.760	0.0002	***

WDR	-0.739	0.088	-8.413	<0.0001	***
LS	-0.015	0.055	-0.2705	0.7868	
PC_MA	0.336	0.0157	21.43	<0.0001	***
VI	0.868	0.174	5.000	<0.0001	***
VA	-0.00872	0.014	-0.6177	0.5367	

Mean dependent var	290.5	S.D. dependent var	215.3
Mean of innovations	-0.068	S.D. of innovations	32.29
Log-likelihood	-3181.7	Akaike criterion	6395.4
Schwarz criterion	6467.1	Hannan-Quinn	6423.2

		Real	Imaginary	Modulus	Frequency
AR					
	Root 1	1.15	0.00	1.15	0.00
MA					
	Root 1	137.19	0.00	137.19	0.5

The integrated model of autoregressive and moving average, ARIMA presented in table 1.1. The Row and Theta are the first order regressive lag to the autoregressive and moving average (p, q) respectively. The first difference change in the value of model affects by 86.8% and 0.07%. The first difference order of autocorrelation shows best fit estimated model on time based lag. Other than that the model constant is computed negative. Thus, the root values of each of the lag parameters have defined positive and negative impact for future estimations. The model is negatively affected by total bovine population, breadable animal population, in milk animal population, male animal population, wet dry ratio in animals, livestock and vaccinations to animals. Defining the first order difference for ARIMA these parameters have change the estimation of total milk production.

The time arrangement information of dairy area has made negative autocorrelations towards the slacks, the primary distinction requesting towards AR is processed positive as 1.15 it demonstrates that there is no over contrast in dairy parameters. The estimation of SD is giving real estimation of slack distinction. Lower the estimation of SD displaying the best attack of estimation and the information are to think about stationary.

The incentive for AR (1) is 1.15 for MA it is 137.19 and coordinated is 0.5 which are near AR, along these lines AR (1) is the best fit. The unit establishes in both the slack demonstrating is discovered multiple. It shows to lessen for requesting to differencing by one.

References:

1. Angrist, J. and J. Pischke (2009). Mostly harmless econometrics: An Empiricist's Companion. Princeton University Press.
2. Bishop, Y., Fienberg, S. and Holland, P. 1975. *Discrete Multivariate Analysis*. Cambridge, Mass.: MIT Press
3. Box, G.E.P. and D.R. Cox. 1964. An analysis of transformations. *Journal of the Royal Statistical Society B* 26: 211-252.
4. Bradu and K. R. Gabriel. The boxplot as a diagnostic tool for models of two-way tables. *Technometrics*, 20:47-68, 1978.
5. Draper, D. (1995), "Assessment and Propagation of Model Uncertainty," *Journal of the Royal Statistical Society Series B* 57: 45-97. Fisher, R. A. (1956), *Statistical Methods and Scientific Inference*. Edinburgh: Oliver and Boyd.
6. Emerson, J.D. 1983. Mathematical aspects of transformation. In Hoaglin, D.C., F. Mosteller and J.W. Tukey (eds) *Understanding Robust and Exploratory Data Analysis*. New York: John Wiley, 247-282.
7. Farrell, M. 1954. The demand for motor cars in the United States. *Journal of the Royal Statistical Society, Series A* 117, 171-93.
8. Gaddum, J. 1933. Reports on biological standards III. Methods of biological assay depending on a quantal response. *Special Report Series Medical Research Council* No. 183, London.
9. Judge et al, *The Theory and Practice of Econometrics*, John Wiley & Sons, New York, 1981, pp. 531-533.
10. Kalvelagen, Erwin. 2007. *Least Squares Calculations with GAMS*.
11. Liker, J., Augustyniak, S. and Duncan, G. (1985). Panel data and models of change: A comparison of first difference and conventional two-wave models. *Social Science Research*, 14(1), pp.80-101.
12. Maddala, G.S. 1983. *Limited Dependent and Qualitative Variables in Econometrics*. Cambridge: Cambridge University Press.
13. McFadden, D. 1974. Conditional logit analysis of qualitative choice behavior. In *Frontiers in Econometrics*, ed. P. Zarembka, New York: Academic Press.
14. Morgan, S. and C. Winship (2015). *Counterfactuals and Causal Inference. Methods and Principles for Social Research*. 2nd ed. Cambridge University Press.
15. Norman R. Draper, Harry Smith, 3rd Edition. *Applied Regression Analysis*. Wiley Series in Probability and Statistics. John Wiley & Sons, 2014.