Sae Hee Kim April 2, 1981 VEE Time Series Student Project Session: Fall 2012

Introduction

I am expecting my first child and one of the changes that I had to make was to cut my consumption of caffeine. This meant I had to cut out coffee and I had to find a good substitute. I found it in a cup of herbal tea. I was never a big fan of tea, but now I came to enjoy the morning cup of tea before starting my day. Therefore, I decided to examine the price of tea over time for my Time Series project.

Data

I obtained my data from the Index Mundi website at

http://www.indexmundi.com/commodities/?commodity=tea&months=120 and the data that I collected was from November 2003 to November 2013. This data contains the average monthly price in US cents per kilogram.



The prices range from a low of 177.95 cents and high of 374.41 cents. From the graph, you see the price has been rising overall until the end of 2012 and decreasing in 2013.

Seasonality

To check further for seasonality, I graphed each of the years by month to observe any monthly trends in the graph below. There do not appear to be any seasonal trends that will require adjustments to the data.



<u>Analysis</u>

Firstly, it must be determined if the data is stationary. In order to do so, we look at the sample autocorrelations by lag period.



We look for a correlogram that declines to zero after several lags to demonstrate stationarity. The correlogram above does not fall to zero quickly and when it does drop to zero at lag 40, it does not stay zero at subsequent lags with minimum fluctuations. It decreases until negative, then rises back up. Thus the series could be represented by an AR(1) or AR(2) process.

First Difference



Below is the graph of the first difference with lags. It does not show any particular trend.

The regression results are as follows:

SUMMARY OUTPUT

Regression Statistics				
Multiple R	0.944742			
R Square	0.892537			
Adjusted R Square	0.891627			
Standard Error	19.53097			
Observations	120			

ANOVA

					Significance
	df	SS	MS	F	F
Regression	1	373850.6	373850.6	980.0548	5.41E-59
Residual	118	45012.15	381.4589		
Total	119	418862.8			

		Standard			Lower	Upper	Lower	Upper
	Coefficients	Error	t Stat	P-value	95%	95%	95.0%	95.0%
					-			
Intercept	16.51733	8.37102	1.97315	0.05081	0.05959	33.0942	-0.05959	33.0942
				5.41E-				
X Var 1	0.940414	0.03004	31.3058	59	0.88092	0.999901	0.880928	0.999901

The result of the equation is $Y_t=16.51733+0.940414Y_{t-1}$. The R² and adjusted R² values are fairly high, so this is a good model. Significance value is close to zero and it indicates that regression is significant.



Below is the predicted AR(1) prices and the original prices.

The AR(1) is almost a perfect fit. Second Difference will be examined to see if it provides better fit.

Second Difference



Again, the graph above, the second difference with lags, does not show any particular trend.

The AR(2) regression is as follows:

SUMMARY OUTPUT

Regression Statistics				
Multiple R	0.945314			
R Square	0.893618			
Adjusted R Square	0.891784			
Standard Error	19.49198			
Observations	119			

ANOVA

					Significance
	df	SS	MS	F	F
Regression	2	370216.4114	185108.2	487.2072	3.62E-57
Residual	116	44072.72747	379.9373		
Total	118	414289.1388			

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} }

The equation is $Y_t=18.51248+1.07713Y_{t-1}-0.14414Y_{t-2}$. The R2 and adjusted R2 calues are a little bit higher than AR(1) model. The coefficients follow the rules of AR(2) stationarity: X1+X2<1 \rightarrow 1.07713+(-0.14414)=0.93299<1 X2<1 \rightarrow -0.14414<1 X2-X1<1 \rightarrow -0.14414-1.07713=-1.22127<1.



The graph above shows the predicted prices vs. the actual prices for the AR(2) model. It is also a very good fit, but it is not as good as AR(1).

Conclusion:

Based on the analysis, AR(1) model, $Y_t=16.51733+0.940414Y_{t-1}$, provides a great prediction of the tea price. The model has proven to provide a very close result when compared to the actual prices.