### PSTAT 174 Lab 6

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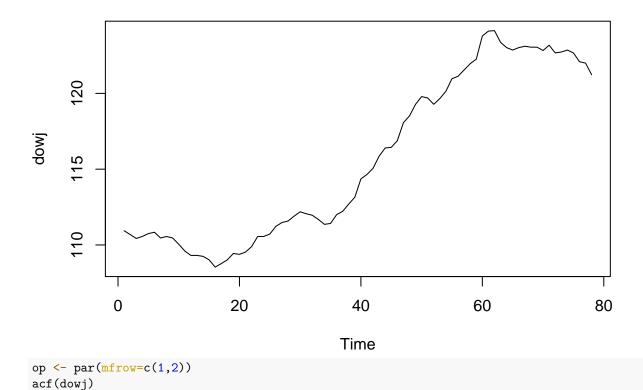
\*\* 1 \*\*

- (a) Summarize how you would carry out the following steps in time series analysis. You can describe briefly by words, or write down R commands you would use to implement these steps
- Step 1 Data processing (make the data stationary) To remove trend take lag at 1, to remove seasonality take lag at 12. Do this until the data doesn't have trend or seasonality. Step 2 Model identification: Analyze the acf and pacf for patterns or shocks, use these to estimate the p,d,q and/or P,D,Q values. So these are your potential models. Step 3 Model estimation Use Yule-Walker and MLE and AICc estimates Step 4 Model selection Select the best fit model by fitting using the potential models, calculate the coffecients and variance. Step 5 Model diagnostics Test for roots to see if it is causal/invertible Test for residuals for independence Step 6 Data forecast Test confidence intervals for forecasting
  - (b) Review this week's lab material, Dow Jones Index question part 3) ('Make the data stationary'). Is differencing once at lag 1 sufficient to make the data stationary? If yes, justify it. If no, try to make it stationary. Please write related R Codes. Yes, differencing once at lag 1 is sufficient since it got rid of the upward trend. If the data had a quadratic trend then we would difference twice. After the data has been differenced the plot is stationary, meaning there are no trends or seasonality.

```
# Load data
dowj_data <- scan("dowj.txt")

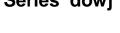
dowj <- ts(dowj_data)
# Plot data
ts.plot(dowj,main = "Dow Jones Index")</pre>
```

### **Dow Jones Index**

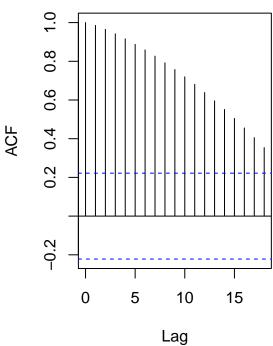


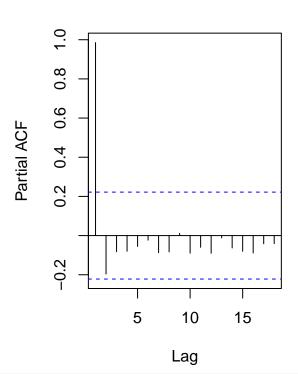
## pacf(dowj)





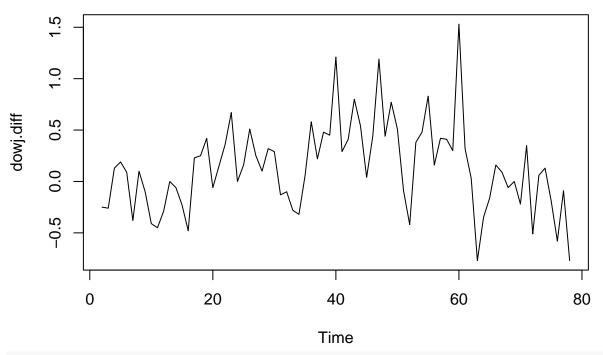
# Series dowj





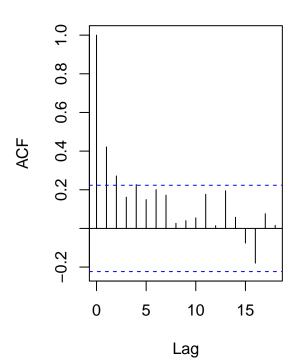
```
par(op)
dowj.diff <- diff(dowj,1)</pre>
ts.plot(dowj.diff, main = "De-trended data")
```

## De-trended data

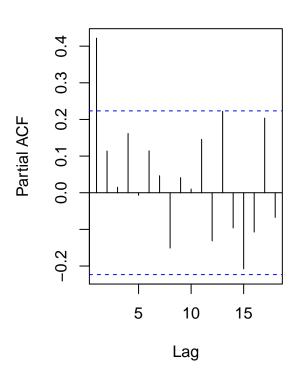


op <- par(mfrow=c(1,2))
acf(dowj.diff)
pacf(dowj.diff)</pre>

# Series dowj.diff



# Series dowj.diff



par(op)