

Lab 5

Pstat 174/274

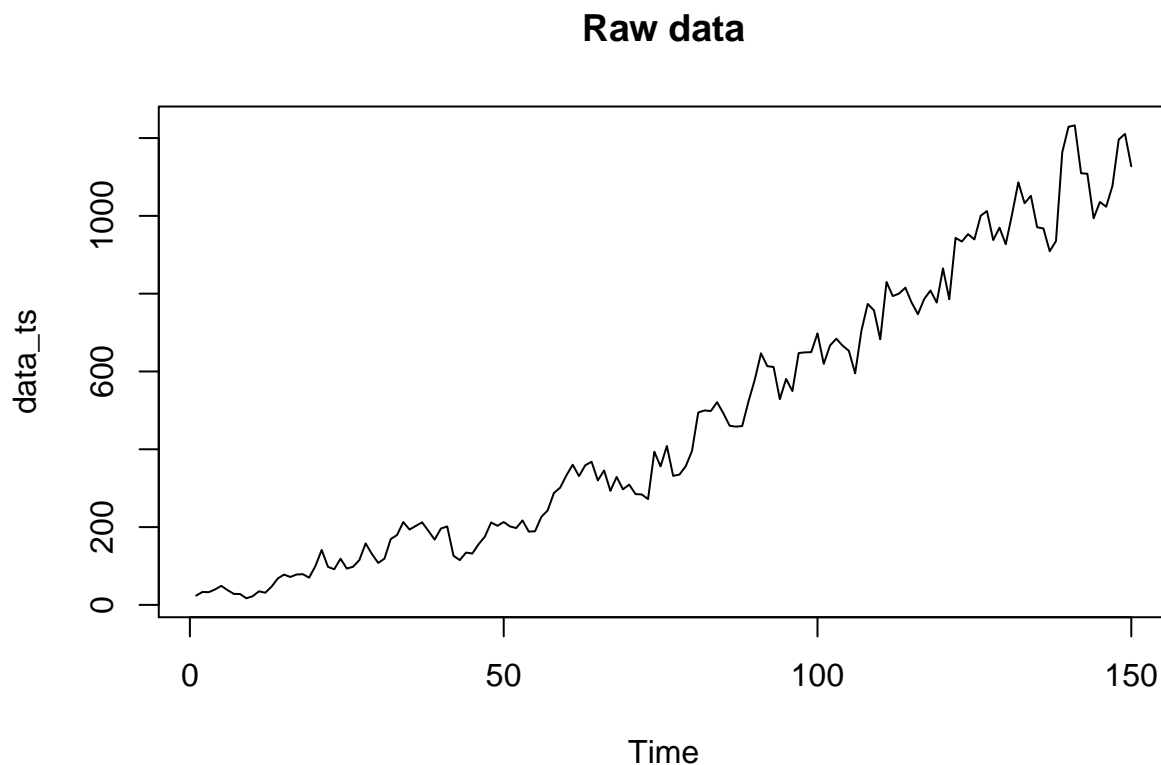
January 31, 2017

- a. Download the file `lab5data.txt` and import the data into R using `scan`.

```
# Set working directory
setwd("/Users/jzapata/Dropbox/Winter 17/TA 174-274/Labs")
# Read data
?scan
data <- scan("lab5data.txt")
```

- b. Plot the time series.

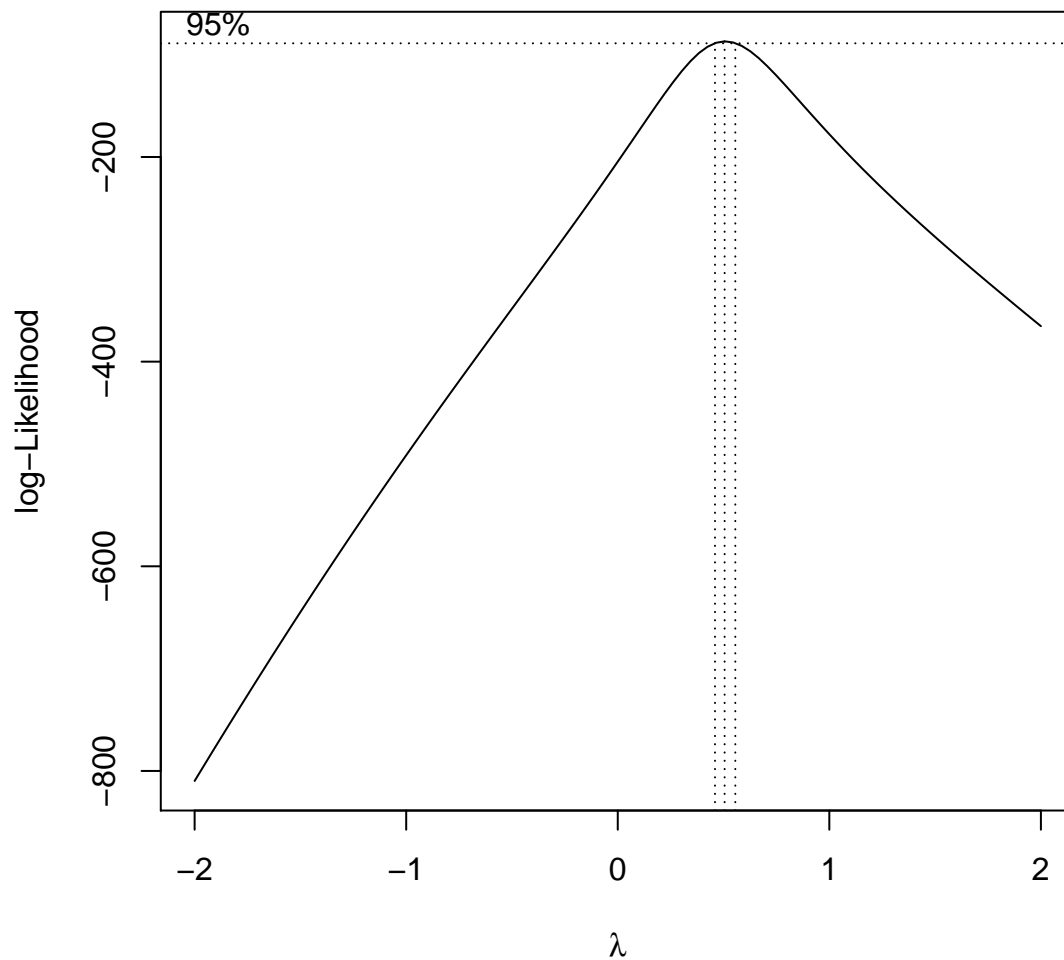
```
# Convert to time series object
data_ts <- ts(data)
ts.plot(data_ts, main = "Raw data")
```



- c. Is the variance constant over time? Is a transformation necessary? Try square root, log, and boxcox transformations to stabilize the variance.

```
# Transform data using boxcox()
require(MASS)
bcTransform <- boxcox(data_ts ~ as.numeric(1:length(data_ts)))
lambda = bcTransform$x[which(bcTransform$y == max(bcTransform$y))]
lambda
```

```
## [1] 0.5050505
```



Notice that the 95% confidence interval for λ includes the value $\lambda = 1/2$ (in fact the estimated MLE of the Box-Cox transformation is $\hat{\lambda} \approx 1/2$). Therefore, we implement the Box-Cox transformation as:

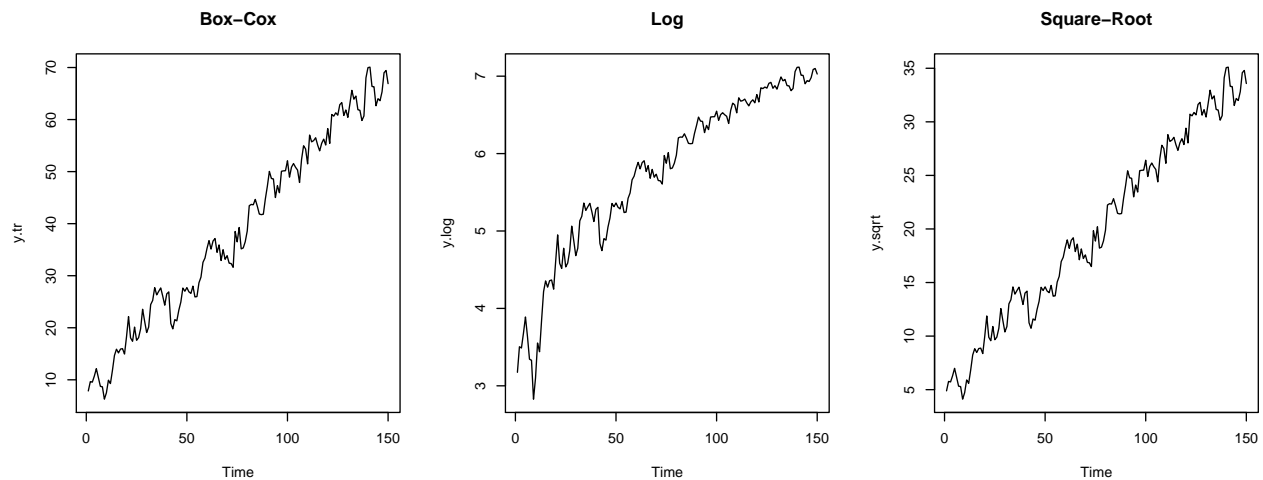
```
y.tr <- (1/lambda)*(data_ts^lambda - 1) # Box-cox
```

as well as the other transformations:

```
y.log <- log(data_ts) # log
y.sqrt <- sqrt(data_ts) # sqrt
```

We now plot the transformed time-series:

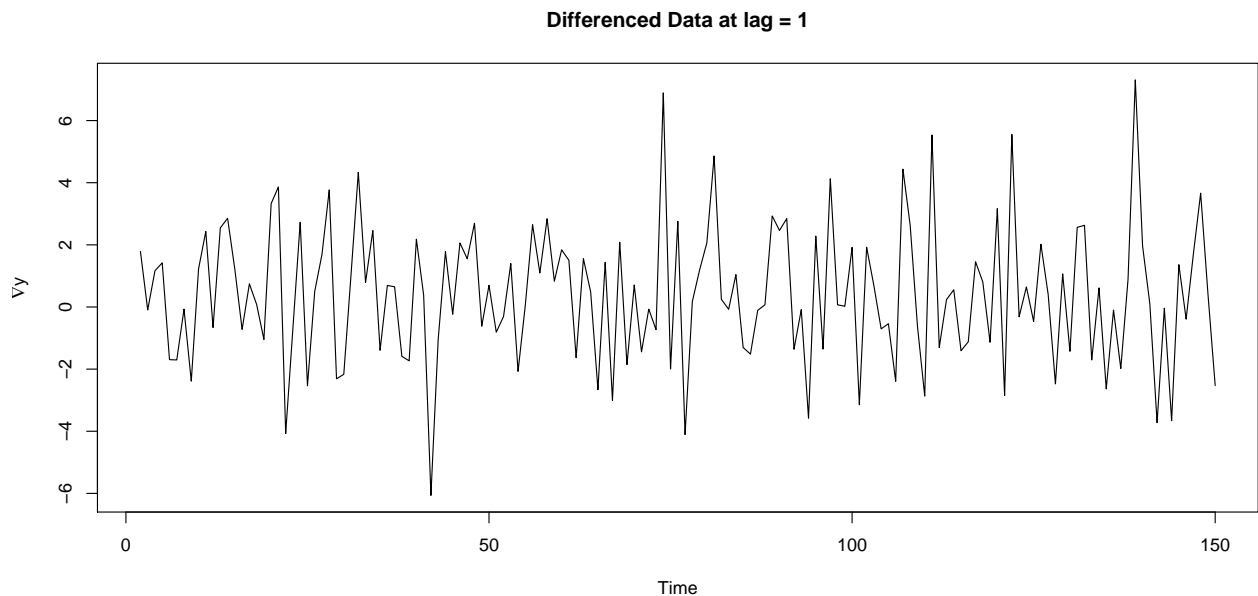
```
op <- par(mfrow = c(1,3))
ts.plot(y.tr,main = "Box-Cox")
ts.plot(y.log,main = "Log")
ts.plot(y.sqrt,main = "Square-Root")
par(op)
```



d. Is there a trend in the data? linear? quadratic? seasonal? Try differencing the data to remove the trend.

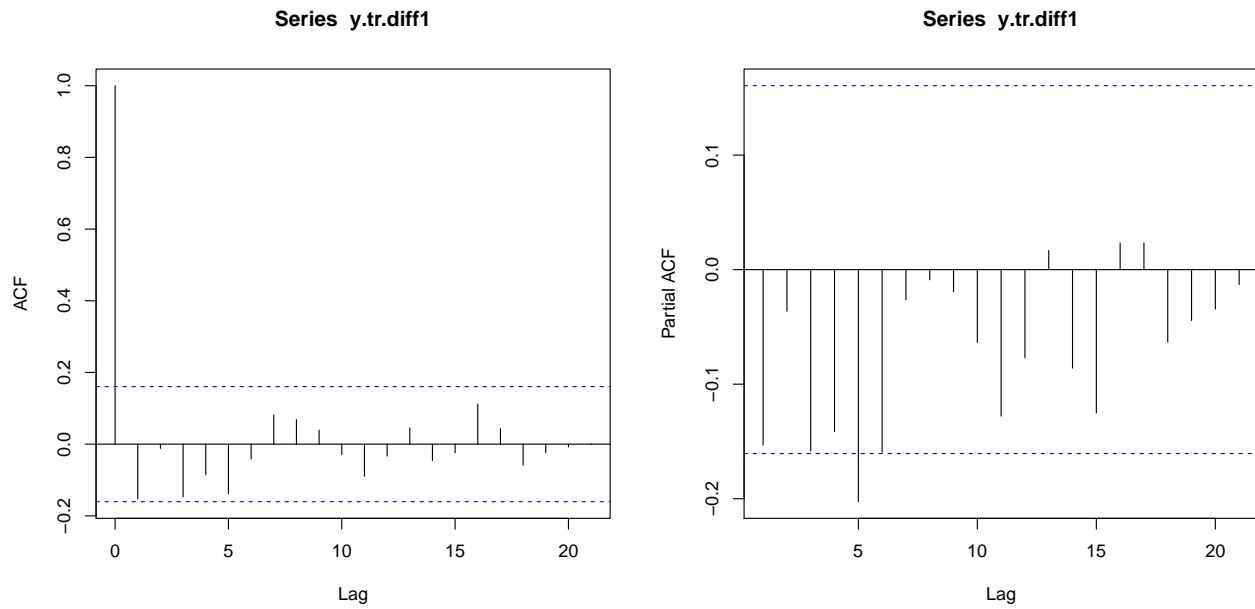
There is a strong linear trend in either of the transformed series (e.g. Box-Cox). Therefore, in order to remove the trend component it is enough to difference the data at lag 1, i.e.:

```
y.tr.diff1 <- diff(y.tr,1)
ts.plot(y.tr.diff1,main = "Differenced Data at lag = 1",ylab=expression(paste(nabla,y)))
```



e. Plot the ACF and PACF. What model do you think generated the data?

```
op <- par(mfrow = c(1,2))
acf(y.tr.diff1)
pacf(y.tr.diff1)
par(op)
```



- ACF: Since all the sample ACF values lie within the 95% confidence bounds we conclude that no significant MA component is present (since otherwise we would have significant ACF values at lags below the order q of the MA process). Moreover, notice the oscillating and decaying behavior of the ACF values indicate the presence of an AR process – which we now confirm via the PACF.
 - PACF: There is a significant sample PACF at lag 5. Moreover, after lag 5 all sample PACF are not significant (lie within the confidence bounds) which match the theoretical behaviour of the PACF of an AR(5) model.
- f. If you suspect the data follows an AR model, estimate the coefficients using Yule-Walker estimation (Type `help(ar)` for reference on how to estimate AR model coefficients using Yule-Walker estimation).

```
# Fit ar model using yule-walker equations
?ar
(fit <- ar(y.tr.diff1, method="yule-walker"))
```

```
##
## Call:
## ar(x = y.tr.diff1, method = "yule-walker")
##
## Coefficients:
##      1      2      3      4      5      6
## -0.2478 -0.1354 -0.2270 -0.1964 -0.2369 -0.1593
##
## Order selected 6  sigma^2 estimated as  4.523
```

Notice that `ar` automatically determines the order of an the AR model via AIC.