

BF_Final_.R

Sun Dec 09 01:58:07 2018

```
library(fpp)
```

```
## Loading required package: forecast
```

```
## Loading required package: fma
```

```
## Loading required package: expsmooth
```

```
## Loading required package: lmtest
```

```
## Loading required package: zoo
```

```
##
```

```
## Attaching package: 'zoo'
```

```
## The following objects are masked from 'package:base':
```

```
##
```

```
##      as.Date, as.Date.numeric
```

```
## Loading required package: tseries
```

```
library(fpp2)
```

```
## Loading required package: ggplot2
```

```
##
```

```
## Attaching package: 'fpp2'
```

```
## The following objects are masked from 'package:fpp':
```

```
##
```

```
##      ausair, ausbeer, austa, austourists, debitcards, departures,
```

```
##      elecequip, euretail, guinearice, oil, sunspotarea, usmelec
```

```
library(dplyr)
```

```
##
```

```
## Attaching package: 'dplyr'
```

```
## The following objects are masked from 'package:stats':
```

```
##
```

```
##      filter, lag
```

```
## The following objects are masked from 'package:base':
```

```
##
```

```
##      intersect, setdiff, setequal, union
```

```

library(forecast)

data<- read.csv("C:\\Users\\ThisPC\\Downloads\\Tractor sales (1).csv")

#1 Create Training and Test data -
set.seed(100) # setting seed to reproduce results of random sampling
trainingRowIndex <- sample(1:nrow(data), 0.8*nrow(data)) # row indices for training data
trainingData <- data[trainingRowIndex, ] # model training data
testData <- data[-trainingRowIndex, ] # test data

#2 Develop the model on the training data and use it to predict the distance on test data
# Build the model on training data -
lmMod <- lm(Number.of.tractors.sold ~ Month.year, data=data) # build the model
salesPred <- predict(lmMod, testData)
summary(lmMod)

##
## Call:
## lm(formula = Number.of.tractors.sold ~ Month.year, data = data)
##
## Residuals:
## ALL 144 residuals are 0: no residual degrees of freedom!
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)         141          NA      NA      NA
## Month.year2003-10          9          NA      NA      NA
## Month.year2003-11         -3          NA      NA      NA
## Month.year2003-12         24          NA      NA      NA
## Month.year2003-2         16          NA      NA      NA
## Month.year2003-3         44          NA      NA      NA
## Month.year2003-4         58          NA      NA      NA
## Month.year2003-5         62          NA      NA      NA
## Month.year2003-6         48          NA      NA      NA
## Month.year2003-7         66          NA      NA      NA
## Month.year2003-8         66          NA      NA      NA
## Month.year2003-9         30          NA      NA      NA
## Month.year2004-1          4          NA      NA      NA
## Month.year2004-10        27          NA      NA      NA
## Month.year2004-11        11          NA      NA      NA
## Month.year2004-12        55          NA      NA      NA
## Month.year2004-2         27          NA      NA      NA
## Month.year2004-3         56          NA      NA      NA
## Month.year2004-4         67          NA      NA      NA
## Month.year2004-5         69          NA      NA      NA
## Month.year2004-6         68          NA      NA      NA
## Month.year2004-7         97          NA      NA      NA

```

## Month.year2004-8	97	NA	NA	NA
## Month.year2004-9	58	NA	NA	NA
## Month.year2005-1	42	NA	NA	NA
## Month.year2005-10	63	NA	NA	NA
## Month.year2005-11	53	NA	NA	NA
## Month.year2005-12	91	NA	NA	NA
## Month.year2005-2	59	NA	NA	NA
## Month.year2005-3	108	NA	NA	NA
## Month.year2005-4	110	NA	NA	NA
## Month.year2005-5	148	NA	NA	NA
## Month.year2005-6	108	NA	NA	NA
## Month.year2005-7	138	NA	NA	NA
## Month.year2005-8	138	NA	NA	NA
## Month.year2005-9	91	NA	NA	NA
## Month.year2006-1	74	NA	NA	NA
## Month.year2006-10	100	NA	NA	NA
## Month.year2006-11	88	NA	NA	NA
## Month.year2006-12	131	NA	NA	NA
## Month.year2006-2	98	NA	NA	NA
## Month.year2006-3	129	NA	NA	NA
## Month.year2006-4	138	NA	NA	NA
## Month.year2006-5	166	NA	NA	NA
## Month.year2006-6	164	NA	NA	NA
## Month.year2006-7	181	NA	NA	NA
## Month.year2006-8	198	NA	NA	NA
## Month.year2006-9	122	NA	NA	NA
## Month.year2007-1	106	NA	NA	NA
## Month.year2007-10	125	NA	NA	NA
## Month.year2007-11	98	NA	NA	NA
## Month.year2007-12	140	NA	NA	NA
## Month.year2007-2	120	NA	NA	NA
## Month.year2007-3	189	NA	NA	NA
## Month.year2007-4	221	NA	NA	NA
## Month.year2007-5	244	NA	NA	NA
## Month.year2007-6	199	NA	NA	NA
## Month.year2007-7	229	NA	NA	NA
## Month.year2007-8	240	NA	NA	NA
## Month.year2007-9	158	NA	NA	NA
## Month.year2008-1	116	NA	NA	NA
## Month.year2008-10	148	NA	NA	NA
## Month.year2008-11	129	NA	NA	NA
## Month.year2008-12	180	NA	NA	NA
## Month.year2008-2	109	NA	NA	NA
## Month.year2008-3	188	NA	NA	NA
## Month.year2008-4	209	NA	NA	NA
## Month.year2008-5	252	NA	NA	NA
## Month.year2008-6	229	NA	NA	NA
## Month.year2008-7	282	NA	NA	NA
## Month.year2008-8	269	NA	NA	NA
## Month.year2008-9	185	NA	NA	NA

## Month.year2009-1	164	NA	NA	NA
## Month.year2009-10	204	NA	NA	NA
## Month.year2009-11	174	NA	NA	NA
## Month.year2009-12	248	NA	NA	NA
## Month.year2009-2	169	NA	NA	NA
## Month.year2009-3	233	NA	NA	NA
## Month.year2009-4	273	NA	NA	NA
## Month.year2009-5	313	NA	NA	NA
## Month.year2009-6	300	NA	NA	NA
## Month.year2009-7	369	NA	NA	NA
## Month.year2009-8	345	NA	NA	NA
## Month.year2009-9	252	NA	NA	NA
## Month.year2010-1	217	NA	NA	NA
## Month.year2010-10	245	NA	NA	NA
## Month.year2010-11	219	NA	NA	NA
## Month.year2010-12	287	NA	NA	NA
## Month.year2010-2	227	NA	NA	NA
## Month.year2010-3	303	NA	NA	NA
## Month.year2010-4	341	NA	NA	NA
## Month.year2010-5	393	NA	NA	NA
## Month.year2010-6	383	NA	NA	NA
## Month.year2010-7	437	NA	NA	NA
## Month.year2010-8	426	NA	NA	NA
## Month.year2010-9	306	NA	NA	NA
## Month.year2011-1	256	NA	NA	NA
## Month.year2011-10	296	NA	NA	NA
## Month.year2011-11	265	NA	NA	NA
## Month.year2011-12	329	NA	NA	NA
## Month.year2011-2	259	NA	NA	NA
## Month.year2011-3	357	NA	NA	NA
## Month.year2011-4	395	NA	NA	NA
## Month.year2011-5	455	NA	NA	NA
## Month.year2011-6	450	NA	NA	NA
## Month.year2011-7	510	NA	NA	NA
## Month.year2011-8	513	NA	NA	NA
## Month.year2011-9	368	NA	NA	NA
## Month.year2012-1	287	NA	NA	NA
## Month.year2012-10	311	NA	NA	NA
## Month.year2012-11	271	NA	NA	NA
## Month.year2012-12	331	NA	NA	NA
## Month.year2012-2	282	NA	NA	NA
## Month.year2012-3	366	NA	NA	NA
## Month.year2012-4	395	NA	NA	NA
## Month.year2012-5	469	NA	NA	NA
## Month.year2012-6	468	NA	NA	NA
## Month.year2012-7	546	NA	NA	NA
## Month.year2012-8	566	NA	NA	NA
## Month.year2012-9	368	NA	NA	NA
## Month.year2013-1	313	NA	NA	NA
## Month.year2013-10	372	NA	NA	NA

```
## Month.year2013-11      340      NA      NA      NA
## Month.year2013-12      426      NA      NA      NA
## Month.year2013-2       314      NA      NA      NA
## Month.year2013-3       427      NA      NA      NA
## Month.year2013-4       469      NA      NA      NA
## Month.year2013-5       565      NA      NA      NA
## Month.year2013-6       520      NA      NA      NA
## Month.year2013-7       626      NA      NA      NA
## Month.year2013-8       642      NA      NA      NA
## Month.year2013-9       442      NA      NA      NA
## Month.year2014-1       384      NA      NA      NA
## Month.year2014-10      440      NA      NA      NA
## Month.year2014-11      378      NA      NA      NA
## Month.year2014-12      464      NA      NA      NA
## Month.year2014-2       379      NA      NA      NA
## Month.year2014-3       446      NA      NA      NA
## Month.year2014-4       569      NA      NA      NA
## Month.year2014-5       652      NA      NA      NA
## Month.year2014-6       608      NA      NA      NA
## Month.year2014-7       730      NA      NA      NA
## Month.year2014-8       707      NA      NA      NA
## Month.year2014-9       499      NA      NA      NA
##
## Residual standard error: NaN on 0 degrees of freedom
## Multiple R-squared:      1, Adjusted R-squared:      NaN
## F-statistic:      NaN on 143 and 0 DF, p-value: NA
```

#3 Density plot to check if the response variable is close to normality

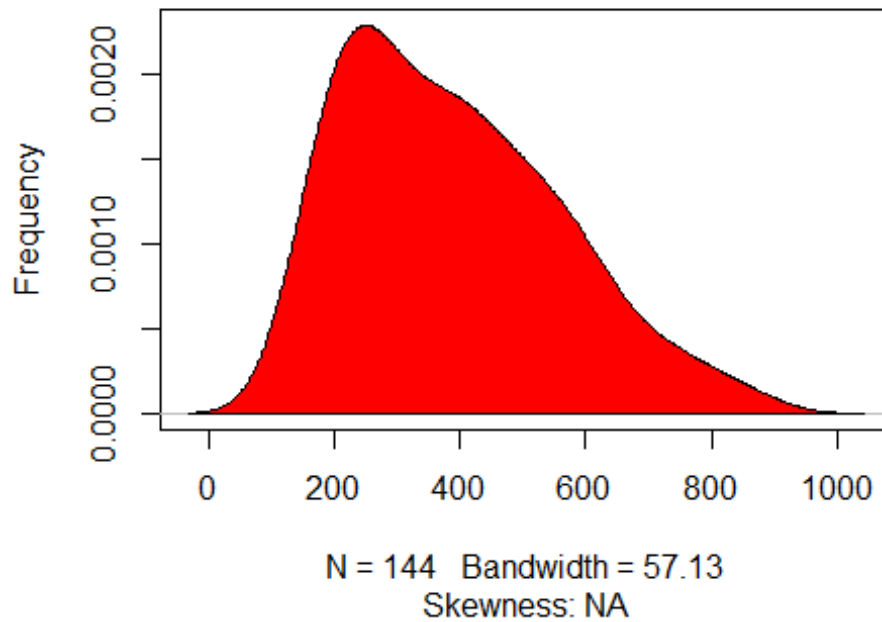
```
library(e1071)
par(mfrow=c(1, 1)) # divide graph area in 2 columns
plot(density(data$Number.of.tractors.sold), main="Density Plot: Month.year",
ylab="Frequency", sub=paste("Skewness:", round(e1071::skewness(data$Month.yea
r), 2))) # density plot for 'Month.year'

## Warning in mean.default(x): argument is not numeric or logical: returning
## NA

## Warning in Ops.factor(x, mean(x)): '-' not meaningful for factors

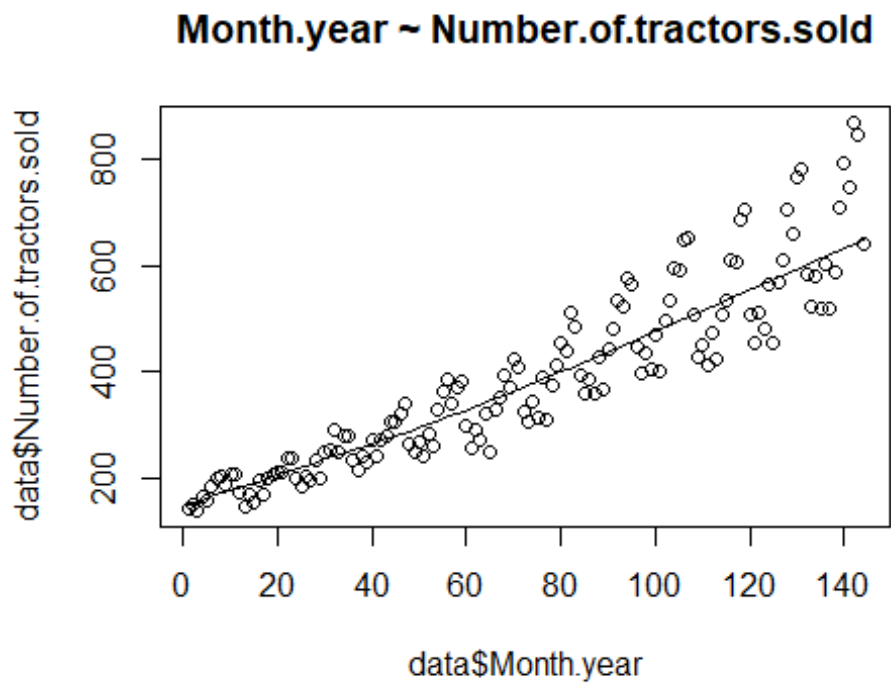
polygon(density(data$Number.of.tractors.sold), col="red")
```

Density Plot: Month.year



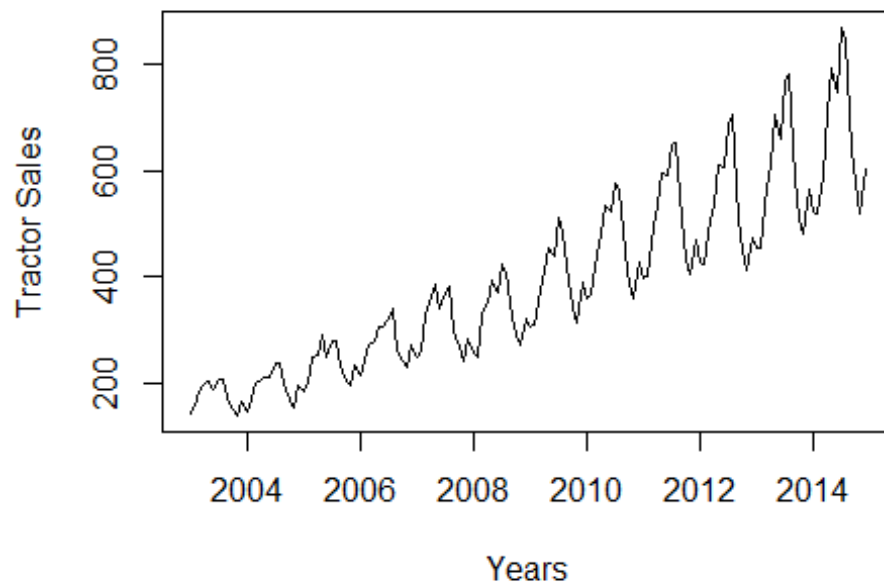
#4 Scatter plot to visualize the linear relationship between the predictor and response

```
scatter.smooth(x=data$Month.year, y=data$Number.of.tractors.sold, main="Month  
.year ~ Number.of.tractors.sold")
```



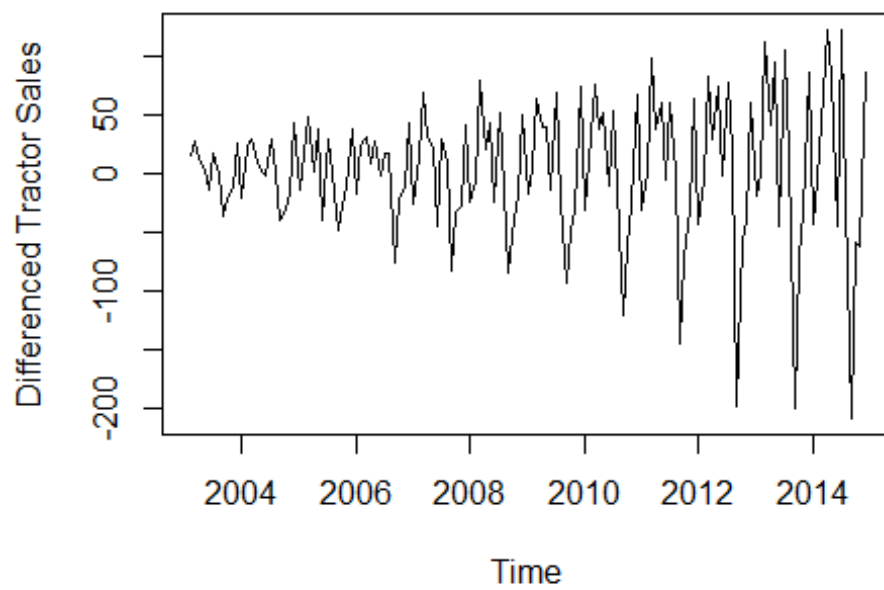
#5 Reading and Plotting the tractor sales data as time series

```
data<-read.csv("C:\\Users\\APEKSHA\\Downloads\\Tractor sales (1).csv")
data = ts(data[,2],start = c(2003,1),frequency = 12)
plot(data, xlab='Years', ylab = 'Tractor Sales')
```



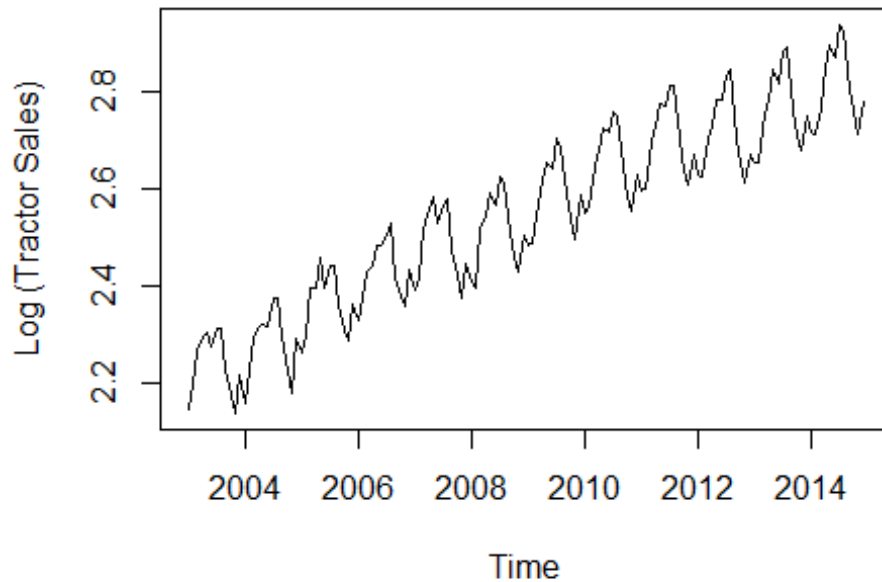
#6 Differencing the data to make data stationary on mean (remove trend)

```
plot(diff(data),ylab='Differenced Tractor Sales')
```



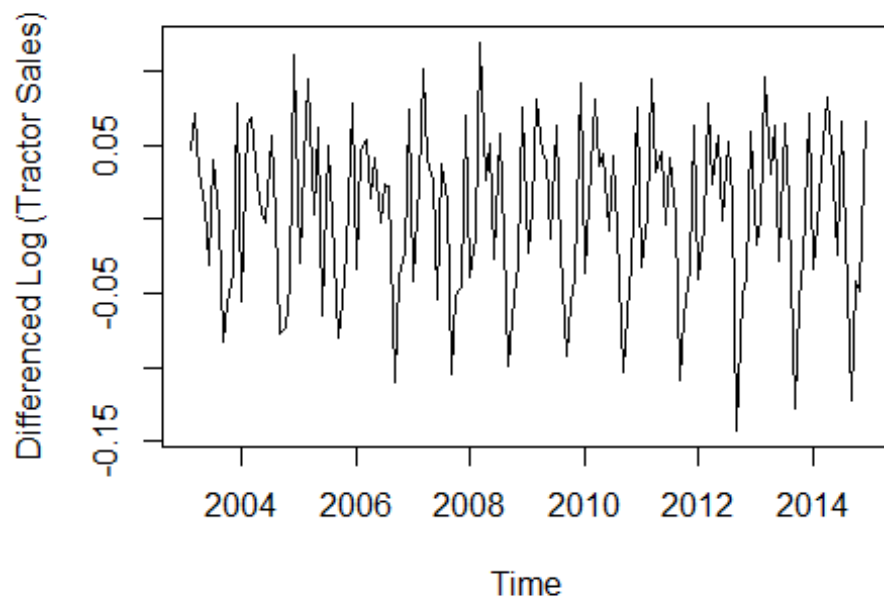
#7 Log transforming the data to make data stationary on variance

```
plot(log10(data),ylab='Log (Tractor Sales)')
```



#8 Differencing the Log transform data to make data stationary on both mean and variance.

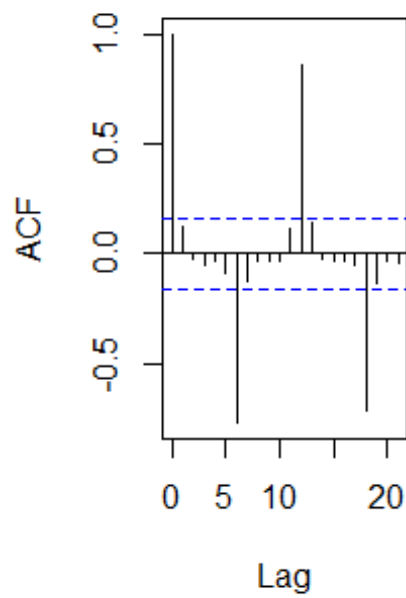
```
plot(diff(log10(data)),ylab='Differenced Log (Tractor Sales)')
```



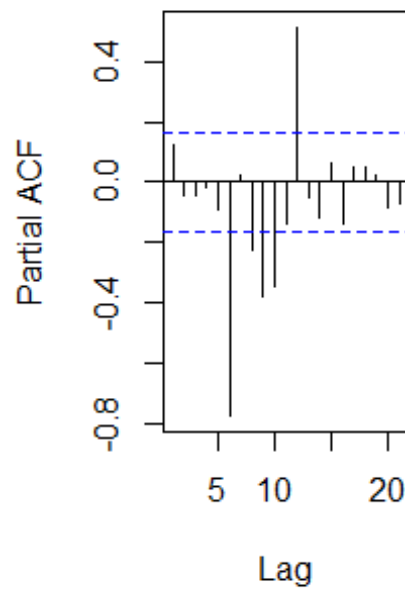
#9 Plotting ACF and PACF to identify potential AR and MA model

```
par(mfrow = c(1,2))  
acf(ts(diff(log10(data))),main='ACF Tractor Sales')  
pacf(ts(diff(log10(data))),main='PACF Tractor Sales')
```

ACF Tractor Sales

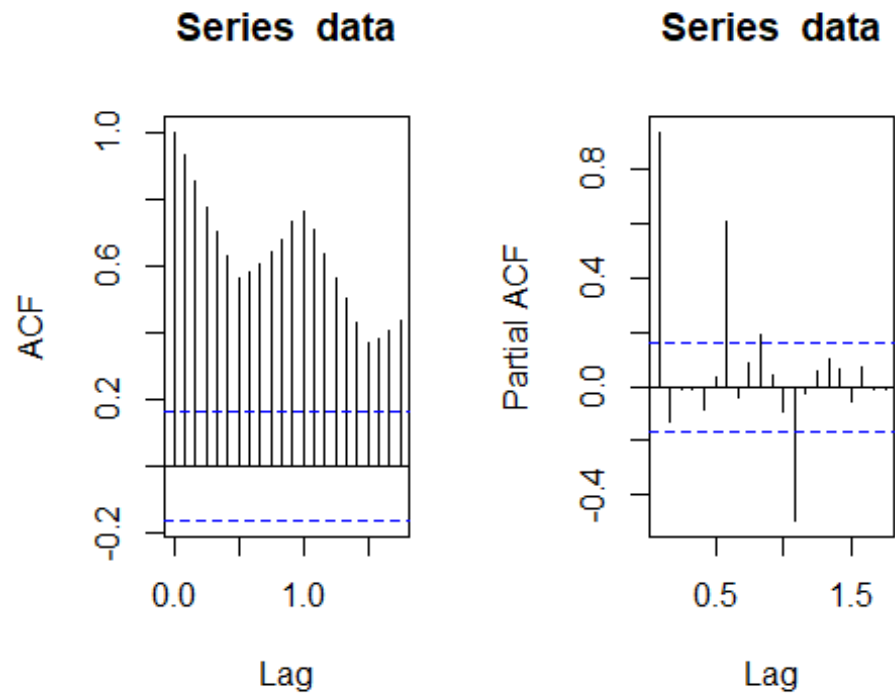


PACF Tractor Sales



#10 Both `acf()` and `pacf()` generates plots by default

```
acfRes <- acf(data) # autocorrelation  
pacfRes <- pacf(data) # partial autocorrelation
```



```
##ARIMA
```

```
#11 Identification of the best fit ARIMA model
```

```
library(tseries)
ARIMAfit = auto.arima(log10(data), approximation=FALSE,trace=FALSE)
summary(ARIMAfit)

## Series: log10(data)
## ARIMA(0,1,1)(0,1,1)[12]
##
## Coefficients:
##          ma1      sma1
##      -0.4047 -0.5529
## s.e.   0.0885  0.0734
##
## sigma^2 estimated as 0.0002571: log likelihood=354.4
## AIC=-702.79  AICc=-702.6  BIC=-694.17
##
## Training set error measures:
##              ME      RMSE      MAE      MPE      MAPE
## Training set 0.0002410698 0.01517695 0.01135312 0.008335713 0.4462212
##              MASE      ACF1
## Training set 0.2158968 0.01062604
```

```
#12 Forecasting the sales using the best fit ARIMA model
```

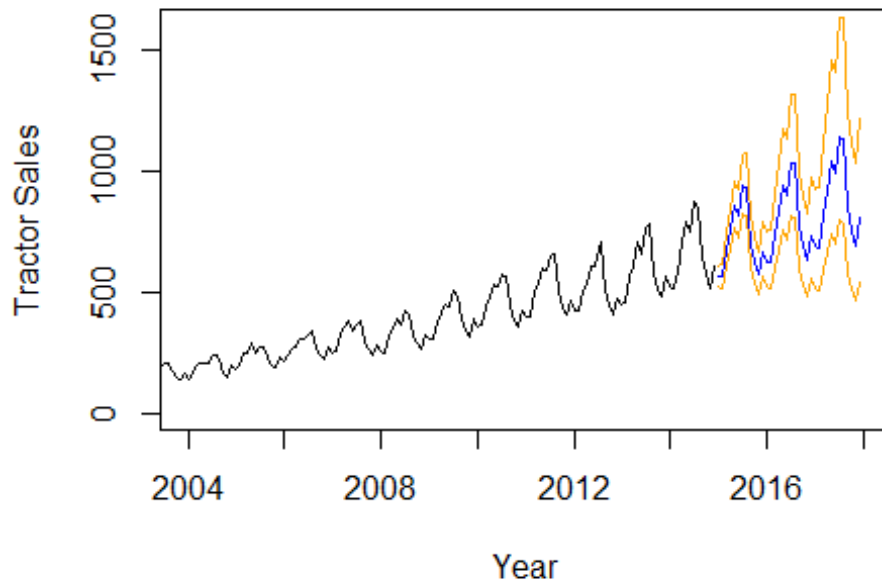
```

par(mfrow = c(1,1))
pred = predict(ARIMAfit, n.ahead = 36)
pred

## $pred
##           Jan           Feb           Mar           Apr           May           Jun           Jul
## 2015 2.754168 2.753182 2.826608 2.880192 2.932447 2.912372 2.972538
## 2016 2.796051 2.795065 2.868491 2.922075 2.974330 2.954255 3.014421
## 2017 2.837934 2.836948 2.910374 2.963958 3.016213 2.996138 3.056304
##           Aug           Sep           Oct           Nov           Dec
## 2015 2.970585 2.847264 2.797259 2.757395 2.825125
## 2016 3.012468 2.889147 2.839142 2.799278 2.867008
## 2017 3.054351 2.931030 2.881025 2.841161 2.908891
##
## $se
##           Jan           Feb           Mar           Apr           May           Jun
## 2015 0.01603508 0.01866159 0.02096153 0.02303295 0.02493287 0.02669792
## 2016 0.03923008 0.04159145 0.04382576 0.04595157 0.04798329 0.04993241
## 2017 0.06386474 0.06637555 0.06879478 0.07113179 0.07339441 0.07558934
##           Jul           Aug           Sep           Oct           Nov           Dec
## 2015 0.02835330 0.02991723 0.03140337 0.03282229 0.03418236 0.03549035
## 2016 0.05180825 0.05361850 0.05536960 0.05706700 0.05871534 0.06031866
## 2017 0.07772231 0.07979828 0.08182160 0.08379608 0.08572510 0.08761165

plot(data,type='l',xlim=c(2004,2018),ylim=c(1,1600),xlab = 'Year',ylab = 'Tractor Sales')
lines(10^(pred$pred),col='blue')
lines(10^(pred$pred+2*pred$se),col='orange')
lines(10^(pred$pred-2*pred$se),col='orange')

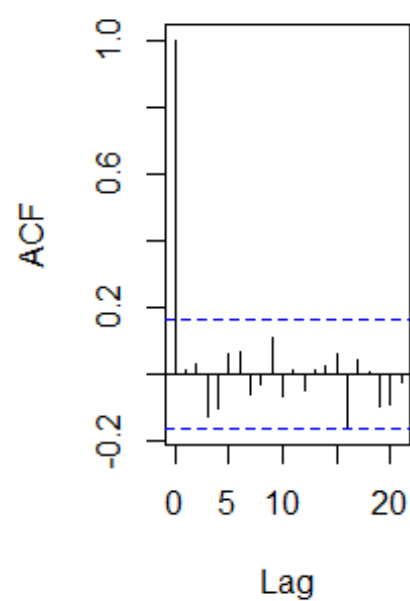
```



#13 Plotting ACF and PACF for residuals of ARIMA model to check if there's more information left for extraction.

```
par(mfrow=c(1,2))  
acf(ts(ARIMAfit$residuals),main='ACF Residual')  
pacf(ts(ARIMAfit$residuals),main='PACF Residual')
```

ACF Residual



PACF Residual

