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```
setwd("C:/Users/Audrey/Data_Science_Workshop_Capstone/Data")

#library(quantmod)

EURGBP = read.csv("EURGBP.csv", colClasses=c(rep("character",
2),rep("numeric",4)))
datetime <- as.POSIXct(paste(EURGBP$DATE, EURGBP$TIME), format = "%Y%m%d
%H%M%OS")
EURGBP = EURGBP[c(3:6)]

summary(EURGBP)

##          OPEN          HIGH          LOW          CLOSE
## Min.      :0.5954   Min.      :0.5954   Min.      :0.5954   Min.      :0.5954
## 1st Qu.:0.6811   1st Qu.:0.6811   1st Qu.:0.6810   1st Qu.:0.6811
## Median :0.7541   Median :0.7542   Median :0.7541   Median :0.7541
## Mean    :0.7585   Mean     :0.7585   Mean     :0.7584   Mean     :0.7585
## 3rd Qu.:0.8392   3rd Qu.:0.8392   3rd Qu.:0.8391   3rd Qu.:0.8392
## Max.    :0.9801   Max.     :0.9801   Max.     :0.9796   Max.     :0.9800

library(xts)

## Warning: package 'xts' was built under R version 3.2.5
## Loading required package: zoo
## Warning: package 'zoo' was built under R version 3.2.4
##
## Attaching package: 'zoo'
##
## The following objects are masked from 'package:base':
##
##      as.Date, as.Date.numeric

EURGBP.xt <- xts(x = EURGBP, order.by = datetime)
head(EURGBP.xt)

##          OPEN  HIGH  LOW  CLOSE
## 2001-01-03 00:01:00 0.6328 0.6328 0.6328 0.6328
## 2001-01-03 00:02:00 0.6328 0.6328 0.6327 0.6327
## 2001-01-03 00:03:00 0.6327 0.6327 0.6327 0.6327
## 2001-01-03 00:04:00 0.6327 0.6328 0.6327 0.6328
## 2001-01-03 00:05:00 0.6328 0.6328 0.6328 0.6328
## 2001-01-03 00:06:00 0.6327 0.6327 0.6327 0.6327

#chartSeries(EURGBP.xt, up.col='green',dn.col='red')
#chartSeries(to.daily(EURGBP.xt),up.col='green',dn.col='red')
```

```

#chartSeries(to.weekly(EURGBP.xt),up.col='green',dn.col='red')
#chartSeries(to.monthly(EURGBP.xt),up.col='green',dn.col='red')
#chartSeries(to.quarterly(EURGBP.xt),up.col='green',dn.col='red')
#chartSeries(to.yearly(EURGBP.xt),up.col='green',dn.col='red')

#seriesLo(EURGBP.xt)
#seriesHi(EURGBP.xt)

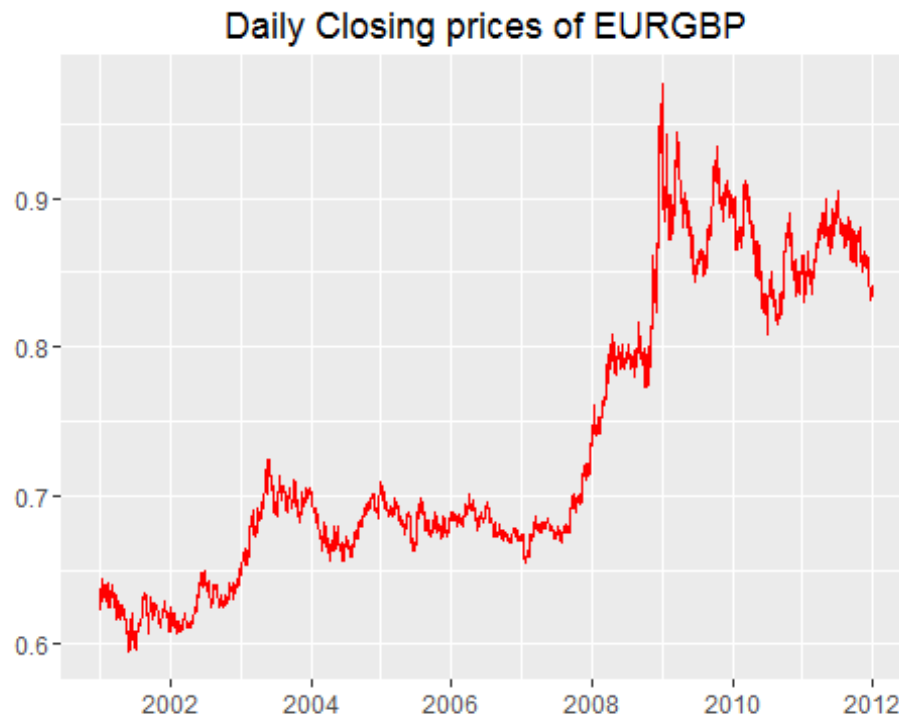
# Testing
EURGBP.testing = EURGBP.xt['2012/2016']
#seriesLo(EURGBP.testing)
#seriesHi(EURGBP.testing)

# Training
EURGBP.training = EURGBP.xt['/2011']
#seriesLo(EURGBP.training)
#seriesHi(EURGBP.training)

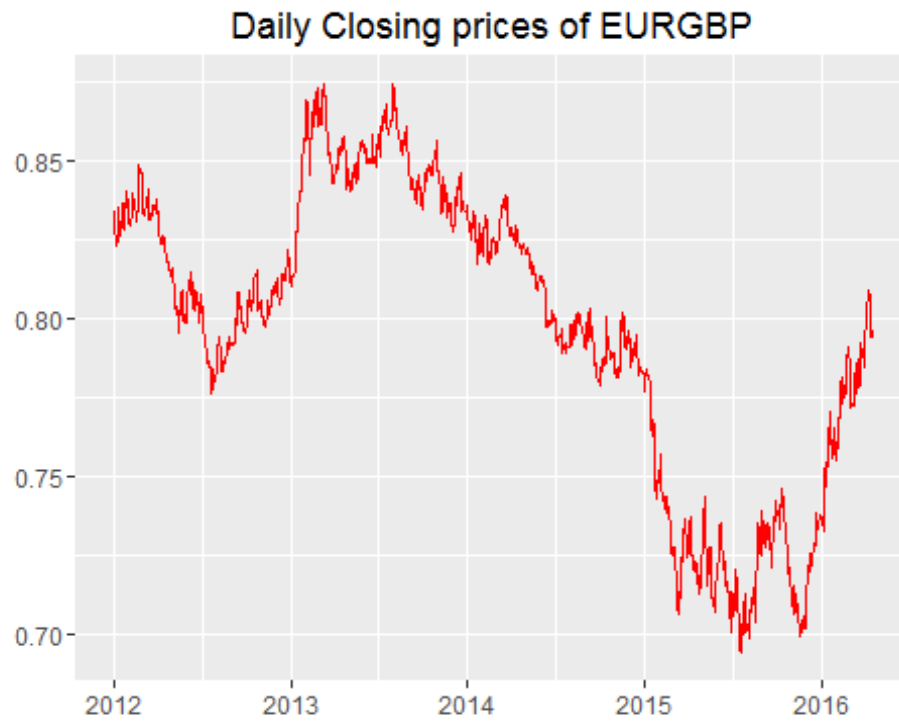
library(forecast)
## Warning: package 'forecast' was built under R version 3.2.5
## Loading required package: timeDate
## This is forecast 7.1
library(ggfortify)
## Loading required package: proto
## Loading required package: ggplot2
## Warning: package 'ggplot2' was built under R version 3.2.5
x = to.daily(EURGBP.training)[,4]
xtest = to.daily(EURGBP.testing)[,4]

# STEP 1
autoplot(x, ts.colour = 'red', main='Daily Closing prices of EURGBP')

```

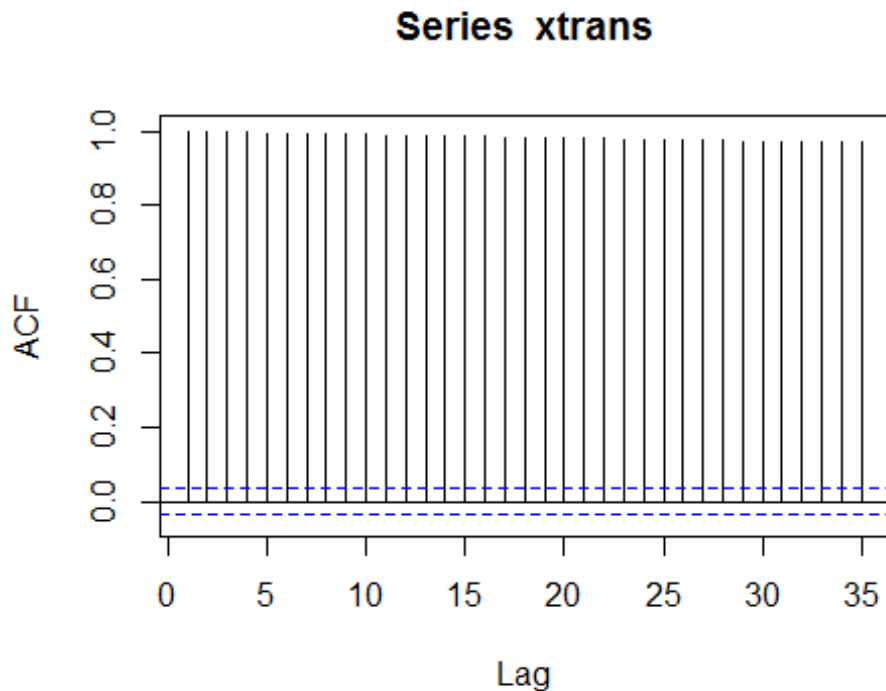


```
autoplot(xtest, ts.colour = 'red', main='Daily Closing prices of EURGBP')
```

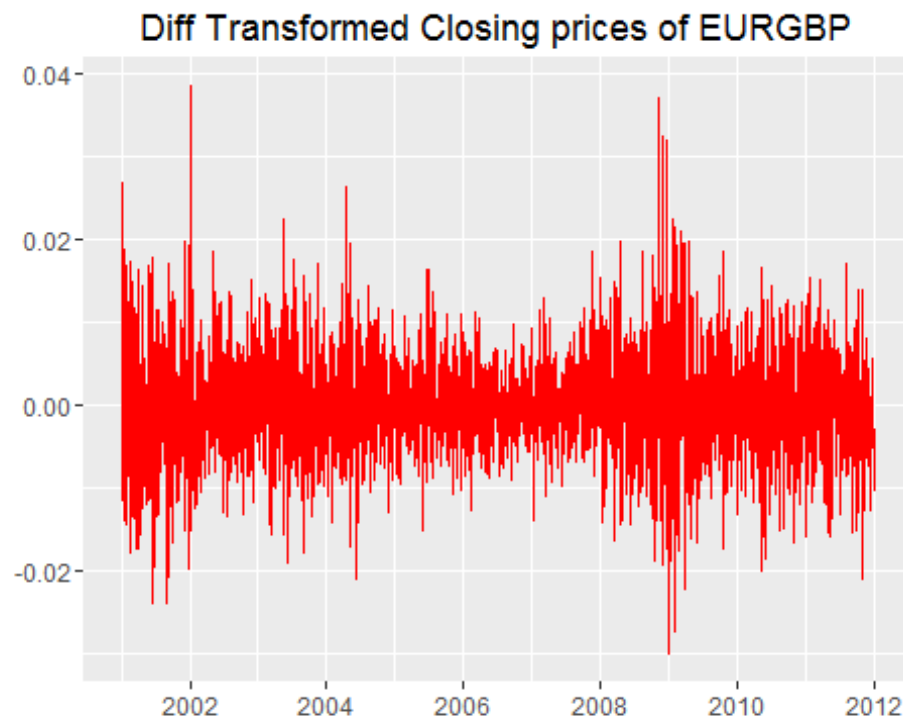


```
lambda = forecast::BoxCox.lambda(x, method="loglik")
xtrans = forecast::BoxCox(x, lambda)
```

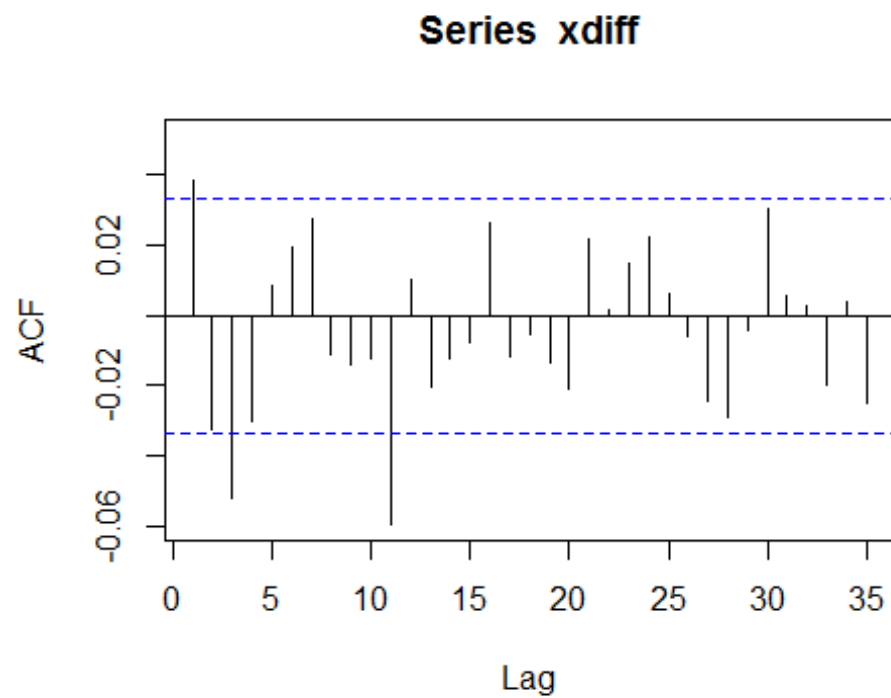
```
xtranstest = forecast::BoxCox(xtest, lambda)
Acf(xtrans)
```



```
tseries::adf.test(xtrans, k = 0)
##
## Augmented Dickey-Fuller Test
##
## data: xtrans
## Dickey-Fuller = -2.4005, Lag order = 0, p-value = 0.4088
## alternative hypothesis: stationary
ndiffs(xtrans)
## [1] 1
# nsdiffs(xtrans)
# STEP 2
xdiff = diff(xtrans)
autoplot(xdiff, ts.colour = 'red', main='Diff Transformed Closing prices of
EURGBP')
## Warning: Removed 1 rows containing missing values (geom_path).
```

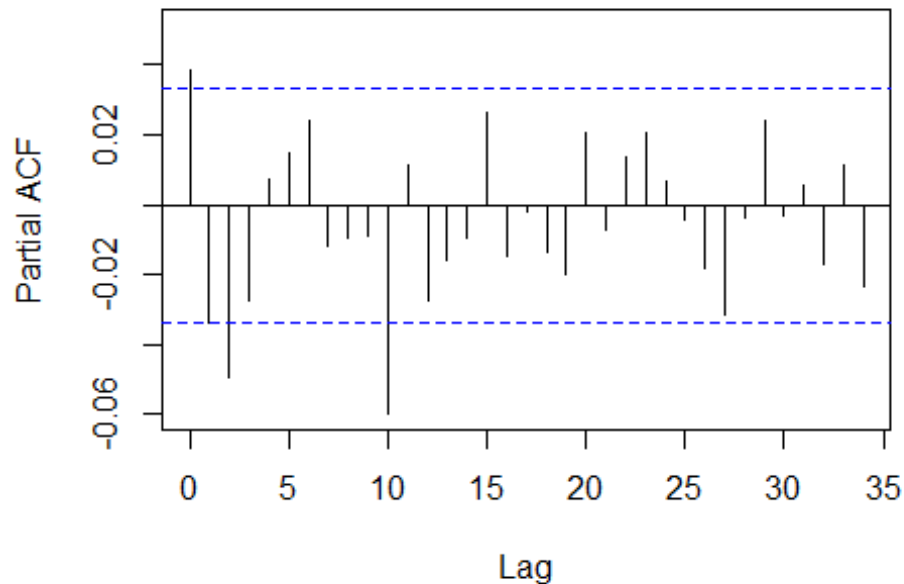


```
Acf(xdiff)
```



```
Pacf(xdiff)
```

### Series x



```
tseries::adf.test(xdiff[-1,], k=0)

## Warning in tseries::adf.test(xdiff[-1, ], k = 0): p-value smaller than
## printed p-value

##
## Augmented Dickey-Fuller Test
##
## data: xdiff[-1, ]
## Dickey-Fuller = -56.437, Lag order = 0, p-value = 0.01
## alternative hypothesis: stationary

summary(auto.arima(xdiff[-1,], stationary=TRUE))

## Series: xdiff[-1, ]
## ARIMA(2,0,2) with zero mean
##
## Coefficients:
##      ar1      ar2      ma1      ma2
##      1.0158 -0.6589 -0.9841  0.5967
## s.e.  0.1950  0.1370  0.2085  0.1476
##
## sigma^2 estimated as 3.81e-05: log likelihood=12578.14
## AIC=-25146.27 AICc=-25146.25 BIC=-25115.57
##
## Training set error measures:
##              ME              RMSE              MAE MPE MAPE              MASE
## Training set 0.0001245453 0.006169261 0.004458355 NaN  Inf 0.6913796
```

```
##                                ACF1
## Training set 0.004395102

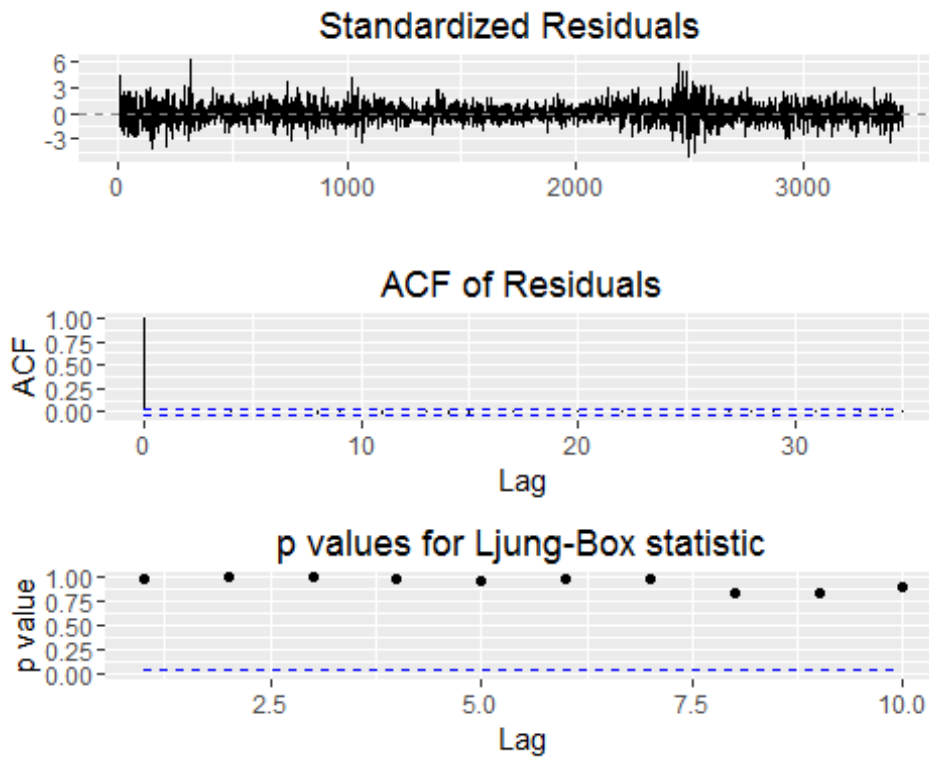
summary(Arima(xtrans, c(2,1,2)))

## Series: xtrans
## ARIMA(2,1,2)
##
## Coefficients:
##          ar1          ar2          ma1          ma2
##          1.0158 -0.6589 -0.9841  0.5967
## s.e.  0.1950  0.1370  0.2085  0.1476
##
## sigma^2 estimated as 3.81e-05:  log likelihood=12578.14
## AIC=-25146.27  AICc=-25146.25  BIC=-25115.57
##
## Training set error measures:
##              ME              RMSE              MAE              MPE              MAPE
## Training set 0.0001243319 0.00616837 0.004457232 -0.1005911 1.792451
##              MASE              ACF1
## Training set 0.999585 0.004245334
```

```
summary(Arima(xtrans, c(2,1,5)))

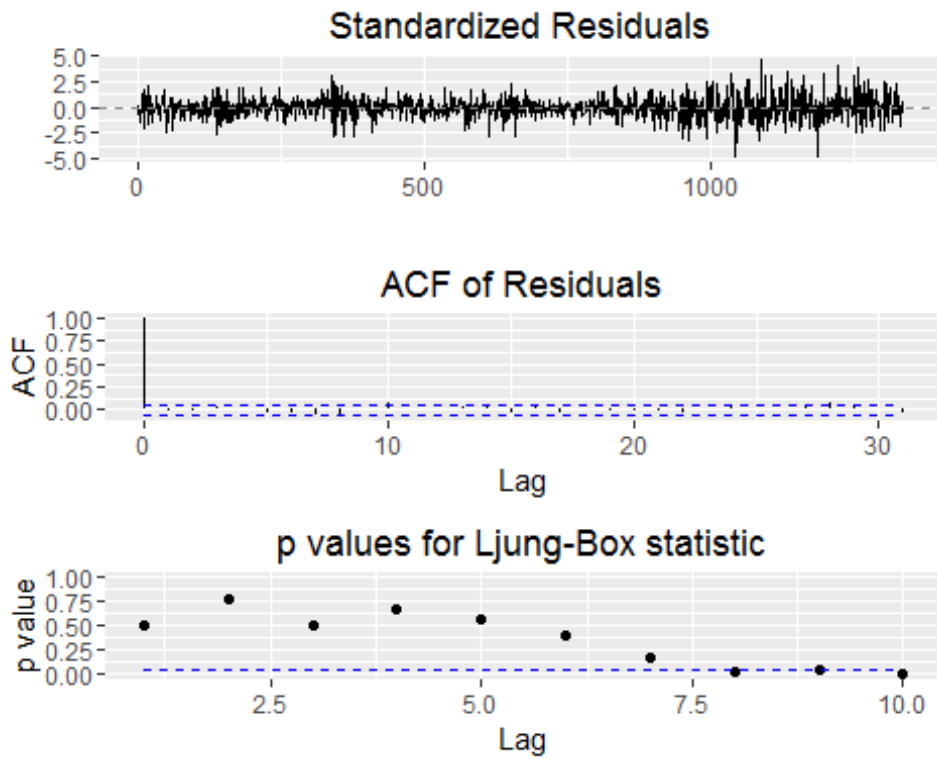
## Series: xtrans
## ARIMA(2,1,5)
##
## Coefficients:
##          ar1          ar2          ma1          ma2          ma3          ma4          ma5
##          1.3502 -0.9782 -1.3158  0.9013  0.0287  0.0192 -0.0291
## s.e.  0.0151  0.0136  0.0229  0.0314  0.0328  0.0289  0.0174
##
## sigma^2 estimated as 3.803e-05:  log likelihood=12582.99
## AIC=-25149.97  AICc=-25149.93  BIC=-25100.86
##
## Training set error measures:
##              ME              RMSE              MAE              MPE              MAPE
## Training set 0.0001230457 0.006159517 0.004453085 -0.09785779 1.786207
##              MASE              ACF1
## Training set 0.9986549 -0.0006372632
```

```
# Model
xtrans.fit <- Arima(xtrans, c(2,1,5))
ggetsdiag(xtrans.fit)
```

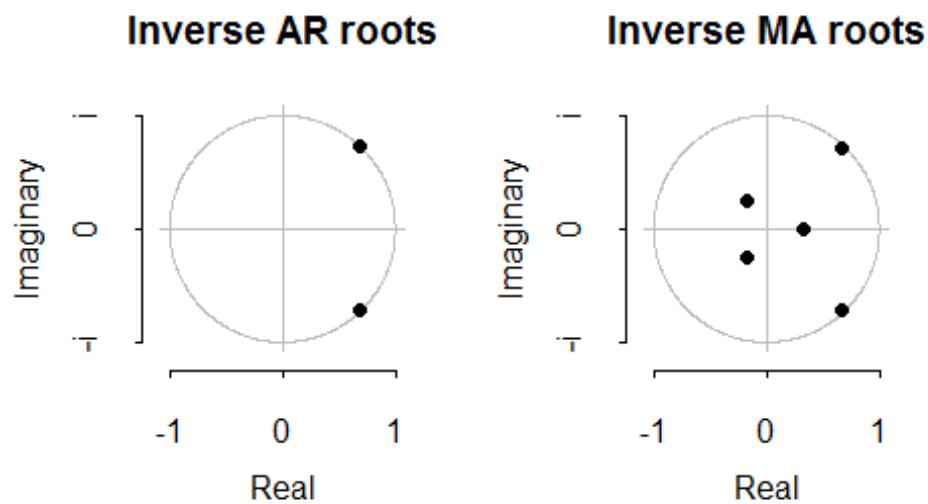


```
# Apply model to test
xtranstest = xtranstest[-1336]
xtranstest.fit = Arima(xtranstest, model=xtrans.fit)
ggtsdiag(xtranstest.fit)
```

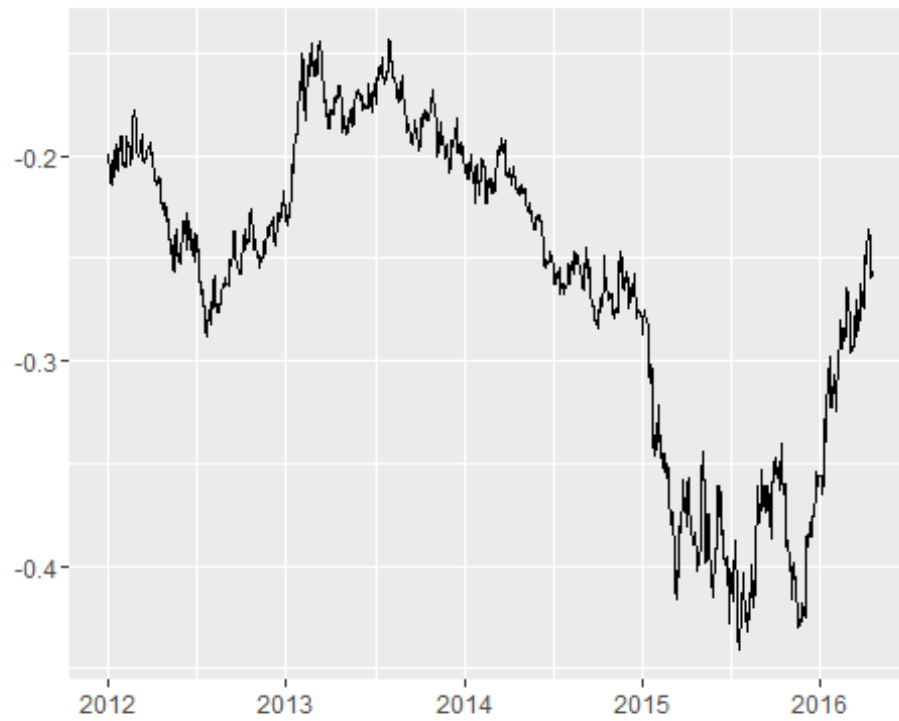




```
plot.Arima(xtranstest.fit)
```



```
autoplot(xtranstest)
```



```
autoplot(xtranstest.fit)
```

