CH5350: Applied Time Series Analysis

Project Report

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Declaration:

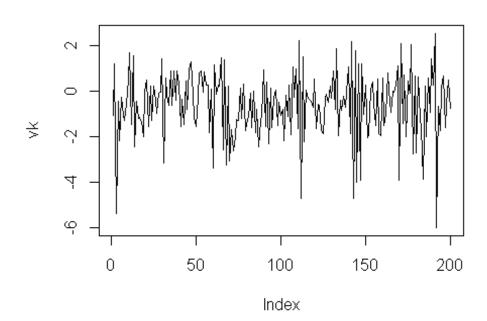
I state that the following work has been done by me and only me without the assistance of any of the other students in this course (CH5350: Applied Time Series Analysis).

Sincerely, Vishal Subbiah (MM12B035)

Vishal Subbiah

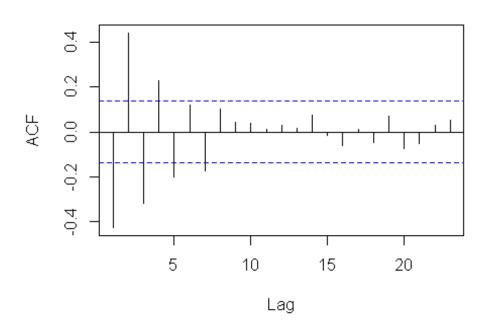
1.R

```
library(TSA)
## Loading required package: leaps
## Loading required package: locfit
## locfit 1.5-9.1 2013-03-22
## Loading required package: mgcv
## Loading required package: nlme
## This is mgcv 1.8-0. For overview type 'help("mgcv-package")'.
## Loading required package: tseries
##
## Attaching package: 'TSA'
##
## The following objects are masked from 'package:stats':
##
##
     acf, arima
##
## The following object is masked from 'package:utils':
##
##
     tar
library(tsDyn)
## Warning: package 'tsDyn' was built under R version 3.1.2
setwd("C:/Users/Toshiba/Desktop/vishal iit/5th sem/Applied time series
analysis/assignments/project")
load("projq1a.Rdata")
plot(vk,type='l')
```



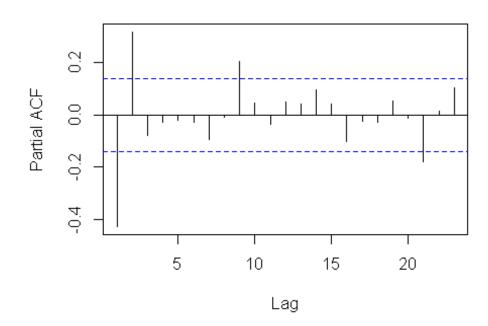
acf(vk)

Series vk



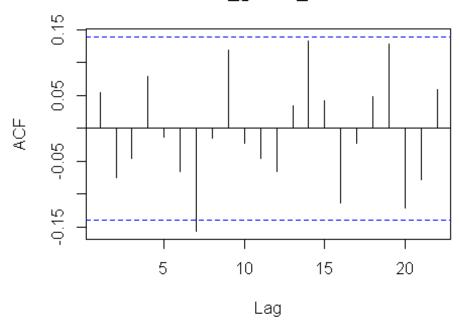
pacf(vk)

Series vk



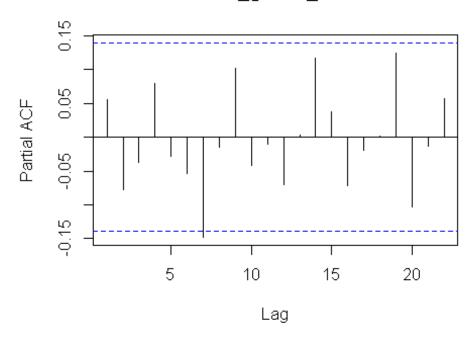
nonlinear_test=**Tsay.test**(vk,order=2)# non linear
vk_guess=**tar**(vk,2,1,1,method="MAIC",estimate.thd=**TRUE**)

Series vk_guess_residuals



pacf(vk_guess_residuals)

Series vk_guess_residuals

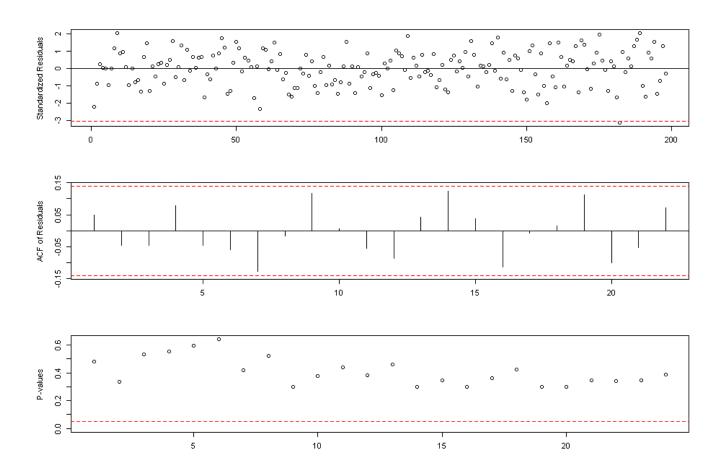


Threshold=vk_guess\$n1 # Threshold vk_guess_1_coeff=vk_guess\$qr1\$coefficients # cofficients for first regime vk_guess_2_coeff=vk_guess\$qr2\$coefficients # cofficients for second regime

```
phi_intercept_1=vk_guess_1_coeff[1]
phi_1_1=vk_guess_1_coeff[2]
phi_2_1=vk_guess_1_coeff[3]

phi_intercept_2=vk_guess_2_coeff[1]
phi_1_2=vk_guess_2_coeff[2]

tsdiag.TAR(vk_guess)
```



```
# part B

epsk=vk_guess_residuals

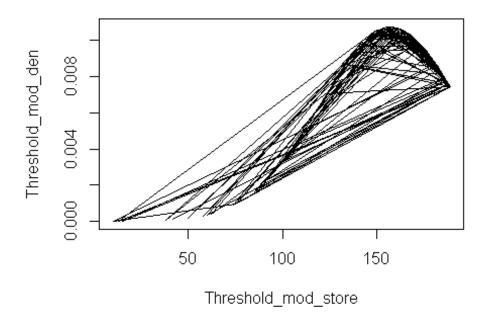
vk_guess_y=vk_guess$y

Threshold_mod_store=0
phi_mod_intercept_1_store=0
phi_mod_1_1_store=0
phi_mod_2_1_store=0

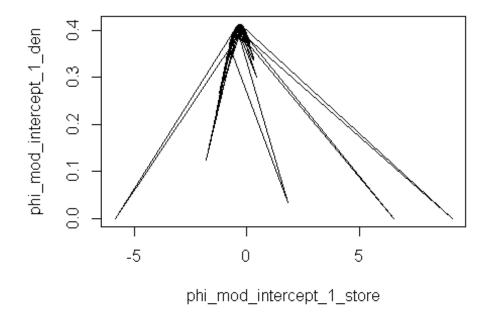
phi_mod_intercept_2_store=0
phi_mod_1_2_store=0
j=1
k=1
```

```
1=1
m=1
n=1
o=1
for (i in 1:200)
{
epskr1<-sample(epsk,size=200,replace=T)
vk_mod=vk_guess_y+epskr1
vk_guess_mod=tar(vk_mod,2,1,1,method="MAIC",estimate.thd=TRUE)
if(!is.na(vk_guess_mod$n1))
Threshold_mod_store[j]=vk_guess_mod$n1
j=j+1
}
vk_guess_mod_1_coeff=vk_guess_mod$qr1$coefficients # cofficients for first regime
vk_guess_mod_2_coeff=vk_guess_mod$qr2$coefficients # cofficients for second regime
if(!is.na(vk_guess_mod_1_coeff[1]))
phi_mod_intercept_1_store[k]=vk_guess_mod_1_coeff[1]
k=k+1
}
if(!is.na(vk_guess_mod_1_coeff[2]))
phi_mod_1_1_store[l]=vk_guess_mod_1_coeff[2]
1=1+1
}
if(!is.na(vk_guess_mod_1_coeff[3]))
phi_mod_2_1_store[m]=vk_guess_mod_1_coeff[3]
m=m+1
if(!is.na(vk_guess_mod_2_coeff[1]))
phi_mod_intercept_2_store[n]=vk_guess_mod_2_coeff[1]
n=n+1
}
if(!is.na(vk_guess_mod_2_coeff[2]))
phi_mod_1_2_store[o]=vk_guess_mod_2_coeff[2]
0=0+1
}
}
Threshold_mod_mean=mean(Threshold_mod_store)
```

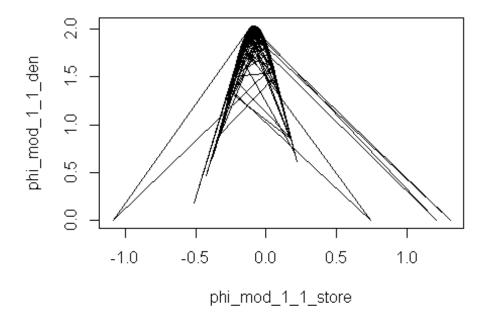
```
Threshold_mod_var=var(Threshold_mod_store)
Threshold_mod_den=dnorm(Threshold_mod_store,mean=Threshold_mod_mean,sd=sqrt(Threshold_mod_var))
plot(Threshold_mod_store,Threshold_mod_den,type="l")
```



```
phi_mod_intercept_1_mean=mean(phi_mod_intercept_1_store)
phi_mod_intercept_1_var=var(phi_mod_intercept_1_store)
phi_mod_intercept_1_den=dnorm(phi_mod_intercept_1_store,mean=phi_mod_intercept_1_mean,sd
=sqrt(phi_mod_intercept_1_var))
plot(phi_mod_intercept_1_store,phi_mod_intercept_1_den,type='l')
```

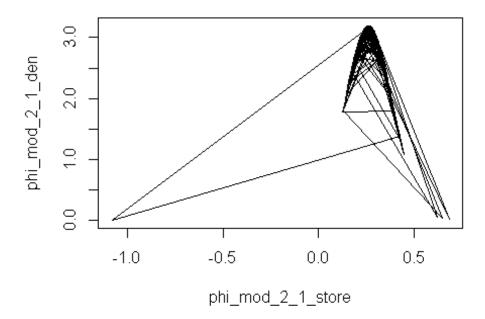


```
phi_mod_1_1_mean=mean(phi_mod_1_1_store)
phi_mod_1_1_var=var(phi_mod_1_1_store)
phi_mod_1_1_den=dnorm(phi_mod_1_1_store,mean=phi_mod_1_1_mean,sd=sqrt(phi_mod_1_1_v ar))
plot(phi_mod_1_1_store,phi_mod_1_1_den,type='l')
```

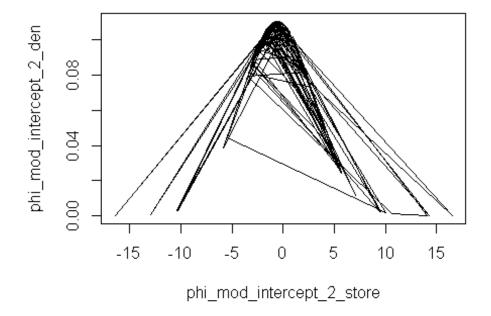


```
phi_mod_2_1_mean=mean(phi_mod_2_1_store)
phi_mod_2_1_var=var(phi_mod_2_1_store)
```

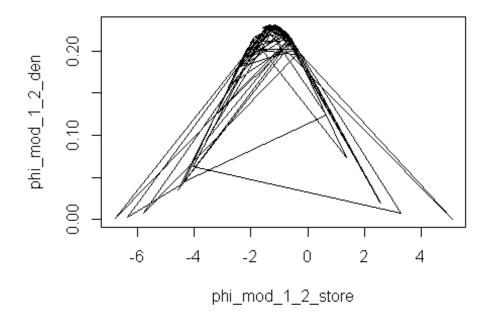
phi_mod_2_1_den=**dnorm**(phi_mod_2_1_store,mean=phi_mod_2_1_mean,sd=**sqrt**(phi_mod_2_1_v ar)) **plot**(phi_mod_2_1_store,phi_mod_2_1_den,type='l')



phi_mod_intercept_2_mean=mean(phi_mod_intercept_2_store)
phi_mod_intercept_2_var=var(phi_mod_intercept_2_store)
phi_mod_intercept_2_den=dnorm(phi_mod_intercept_2_store,mean=phi_mod_intercept_2_mean,sd
=sqrt(phi_mod_intercept_2_var))
plot(phi_mod_intercept_2_store,phi_mod_intercept_2_den,type="I")



```
phi_mod_1_2_mean=mean(phi_mod_1_2_store)
phi_mod_1_2_var=var(phi_mod_1_2_store)
phi_mod_1_2_den=dnorm(phi_mod_1_2_store,mean=phi_mod_1_2_mean,sd=sqrt(phi_mod_1_2_v
ar))
plot(phi_mod_1_2_store,phi_mod_1_2_den,type='l')
```



Variable	From tar routine	Mean	Variance
Threshold	175	156.66	1290.74
Intercept ¹	-0.165	-0.419	0.3845
-d ₁ ¹	-0.0248	-0.107	0.0262
-d ₂ ¹	0.422	0.272	0.0042
Intercept ²	0.366	-0.13	6.9465
-d ₁ ²	-1.88	-1.74	1.5018

The delay is 1.

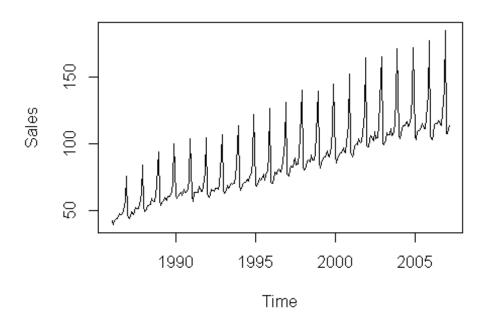
This model satisfies since the residuals are white based on looking at the ACF and PACF and by running Box.test on the residuals.

2.R

```
library(TSA)
## Loading required package: leaps
## Loading required package: locfit
## locfit 1.5-9.1 2013-03-22
## Loading required package: mgcv
## Loading required package: nlme
## This is mgcv 1.8-0. For overview type 'help("mgcv-package")'.
## Loading required package: tseries
## Attaching package: 'TSA'
##
## The following objects are masked from 'package:stats':
##
##
     acf, arima
##
## The following object is masked from 'package:utils':
##
##
     tar
library(astsa)
library(forecast)
## Loading required package: zoo
##
## Attaching package: 'zoo'
## The following objects are masked from 'package:base':
##
##
     as.Date, as.Date.numeric
##
## Loading required package: timeDate
## Attaching package: 'timeDate'
##
## The following objects are masked from 'package:TSA':
##
##
     kurtosis, skewness
##
## This is forecast 5.6
##
##
## Attaching package: 'forecast'
## The following object is masked from 'package:astsa':
##
##
     gas
## The following object is masked from 'package:TSA':
##
```

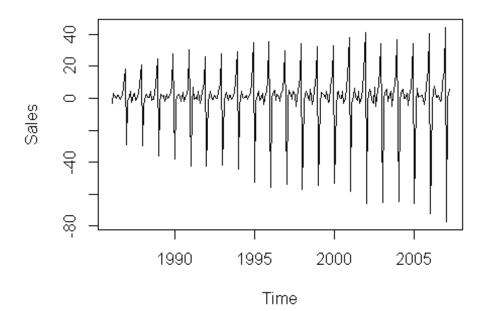
```
## fitted.Arima
##
## The following object is masked from 'package:nlme':
##
## getResponse

data(retail)
#retail_new=retail[1:255]
plot(retail,type='l')
```



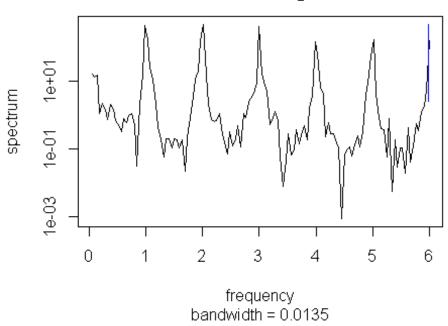
```
retail_diff=diff(retail)

plot(retail_diff)# d=1, D=1
```



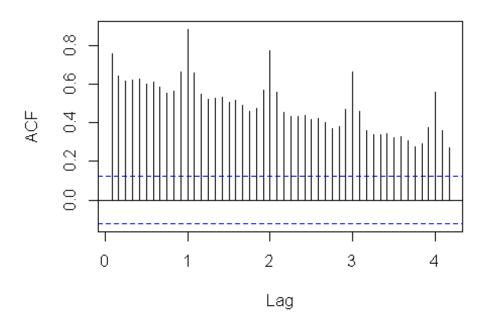
spectrum(retail)# S=12





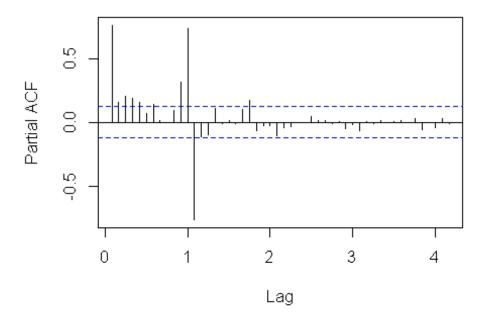
retail_acf=acf(retail,lag.max=50)#q=2,Q=1

Series retail



retail_pacf=pacf(retail,lag.max=50)#p=2,P=0

Series retail



```
retail_test=sarima(retail,2,1,2,D=1,Q=1,S=12)

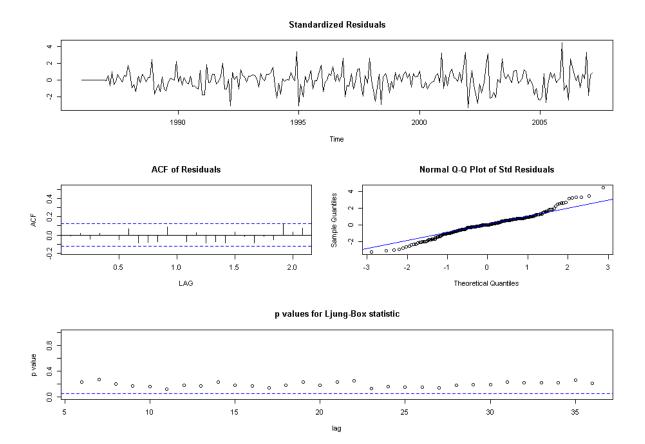
## initial value 0.840627

## iter 2 value 0.685920

## iter 3 value 0.645495

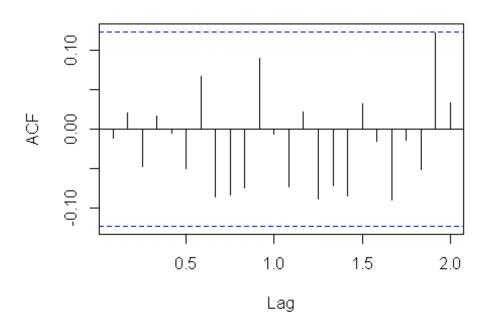
## iter 4 value 0.645147
```

```
## iter 5 value 0.644846
## iter 6 value 0.644707
## iter 7 value 0.644628
## iter 8 value 0.644592
## iter 9 value 0.644499
## iter 10 value 0.644275
## iter 11 value 0.643740
## iter 12 value 0.642994
## iter 13 value 0.642226
## iter 14 value 0.641474
## iter 15 value 0.641163
## iter 16 value 0.641076
## iter 17 value 0.641016
## iter 18 value 0.640740
## iter 19 value 0.640226
## iter 20 value 0.639682
## iter 21 value 0.639121
## iter 22 value 0.638806
## iter 23 value 0.638802
## iter 24 value 0.638802
## iter 25 value 0.638802
## iter 26 value 0.638802
## iter 27 value 0.638802
## iter 28 value 0.638802
## iter 29 value 0.638802
## iter 30 value 0.638802
## iter 31 value 0.638801
## iter 32 value 0.638801
## iter 32 value 0.638801
## iter 32 value 0.638801
## final value 0.638801
## converged
## initial value 0.632797
## iter 2 value 0.632772
## iter 3 value 0.632748
## iter 4 value 0.632689
## iter 5 value 0.632626
## iter 6 value 0.632520
## iter 7 value 0.632494
## iter 8 value 0.632480
## iter 9 value 0.632477
## iter 10 value 0.632475
## iter 11 value 0.632475
## iter 11 value 0.632475
## iter 11 value 0.632475
## final value 0.632475
## converged
```



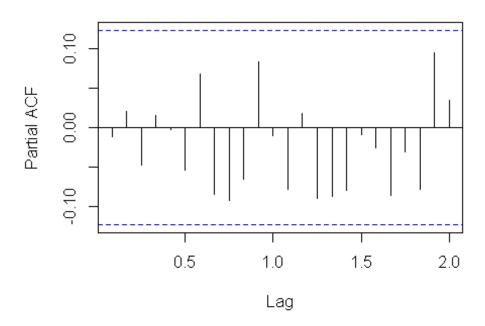
retail_residuals=retail_test\$fit\$residuals
acf(retail_residuals)

Series retail_residuals



pacf(retail_residuals)

Series retail_residuals



The Model is $ARIMA(2,1,2) \times (0,1,1)_{12}$ as seen based on the values and plots above.

This model satisfies since the residuals are white based on looking at the ACF and PACF and by running Box.test on the residuals.