

Arima

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```
###Forecasting

library(tseries)

## Warning: package 'tseries' was built under R version 4.1.1

## Registered S3 method overwritten by 'quantmod':
##   method      from
##   as.zoo.data.frame zoo

library(forecast)

## Warning: package 'forecast' was built under R version 4.1.1

library(foreign)
library(TSA)

## Warning: package 'TSA' was built under R version 4.1.1

## Registered S3 methods overwritten by 'TSA':
##   method      from
##   fitted.Arima forecast
##   plot.Arima   forecast

##
## Attaching package: 'TSA'

## The following objects are masked from 'package:stats':
## 
##   acf, arima

## The following object is masked from 'package:utils':
## 
##   tar

library(astsa)

## Warning: package 'astsa' was built under R version 4.1.1
```

```

##  

## Attaching package: 'astsa'  

##  

## The following object is masked from 'package:forecast':  

##  

##     gas  

## ##For Nifty50  

price=nifty50$Close  

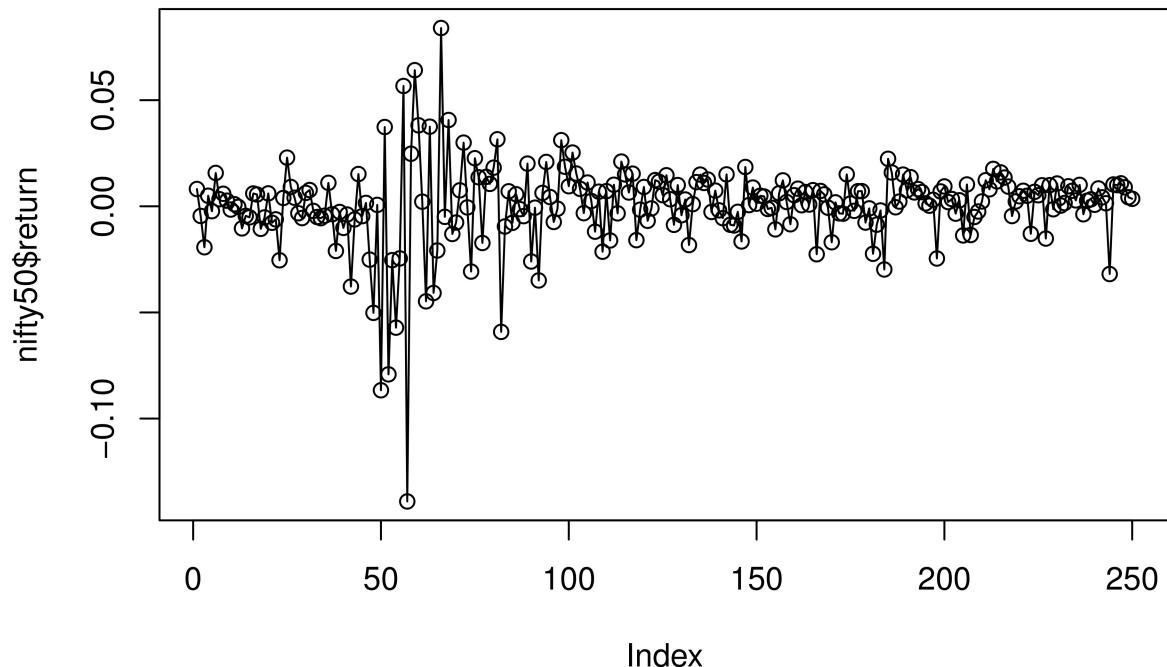
log_return=diff(log(price),lag=1)  

nifty50=nifty50[-1,]  

nifty50['return']=log_return  

plot(nifty50$return,type="o")

```



```

## ##Test for Stationarity  

adf.test(nifty50$return, alternative = "stationary")  

## Warning in adf.test(nifty50$return, alternative = "stationary"): p-value smaller  

## than printed p-value  

##  

## Augmented Dickey-Fuller Test  

##

```

```

## data: nifty50$return
## Dickey-Fuller = -4.8436, Lag order = 6, p-value = 0.01
## alternative hypothesis: stationary

kpss=kpss.test(nifty50$return)
kpss

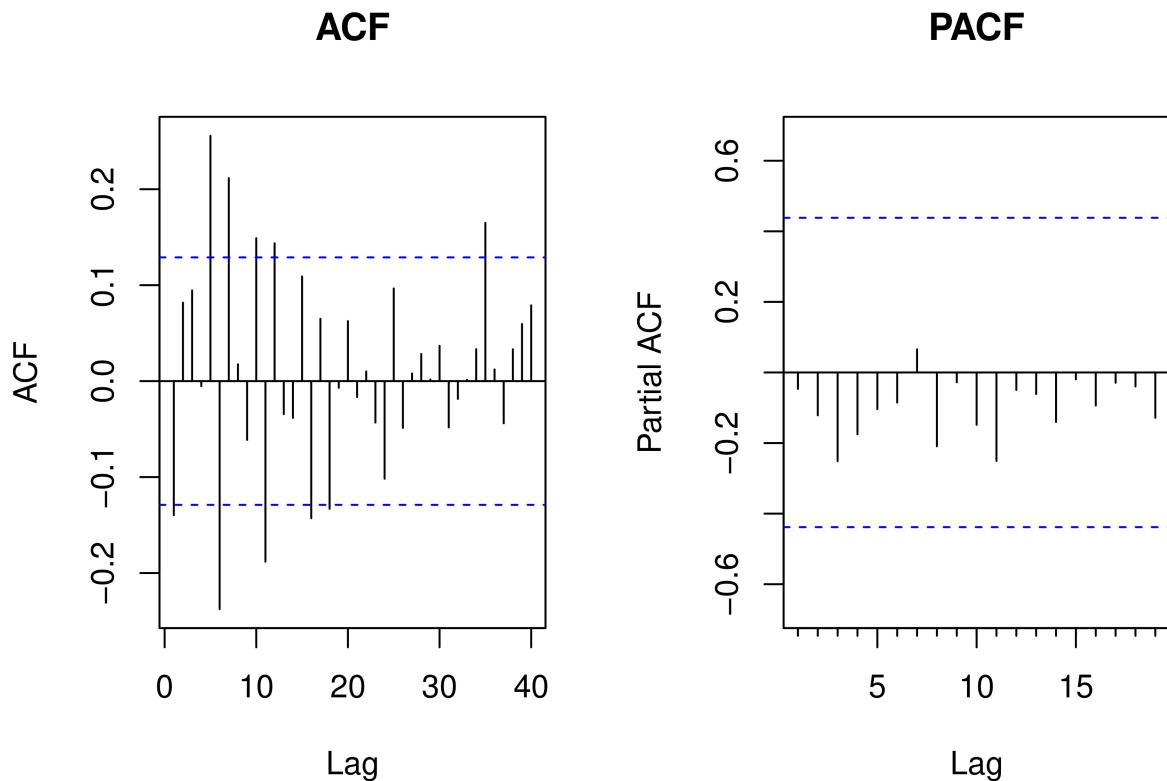
## 
## KPSS Test for Level Stationarity
##
## data: nifty50$return
## KPSS Level = 0.39357, Truncation lag parameter = 5, p-value = 0.07993

pp=PP.test(nifty50$return)
pp

## 
## Phillips-Perron Unit Root Test
##
## data: nifty50$return
## Dickey-Fuller = -18.1, Truncation lag parameter = 5, p-value = 0.01

par(mfrow=c(1,2))
y_train=nifty50[1:231,]
y_test=nifty50[231:250,]
Acf(y_train$return, lag.max = 40, main='ACF')
Pacf(y_test$return, lag.max = 40, main = 'PACF')

```



```

tail(y_test$return)

## [1] 0.010293171 0.009960404 0.010833605 0.008974644 0.004272496 0.003535794

aic_model=auto.arima(y_train$return,trace=TRUE,d=0,max.p=10,max.q=10,ic="aic",approximation=FALSE)

##
##  ARIMA(2,0,2) with non-zero mean : -1136.634
##  ARIMA(0,0,0) with non-zero mean : -1133.306
##  ARIMA(1,0,0) with non-zero mean : -1135.844
##  ARIMA(0,0,1) with non-zero mean : -1135.181
##  ARIMA(0,0,0) with zero mean     : -1135.251
##  ARIMA(1,0,2) with non-zero mean : -1138.487
##  ARIMA(0,0,2) with non-zero mean : -1135.586
##  ARIMA(1,0,1) with non-zero mean : -1134.183
##  ARIMA(1,0,3) with non-zero mean : -1141.652
##  ARIMA(0,0,3) with non-zero mean : -1136.117
##  ARIMA(2,0,3) with non-zero mean : -1142.896
##  ARIMA(3,0,3) with non-zero mean : -1151.297
##  ARIMA(3,0,2) with non-zero mean : -1144.083
##  ARIMA(4,0,3) with non-zero mean : Inf
##  ARIMA(3,0,4) with non-zero mean : -1149.346
##  ARIMA(2,0,4) with non-zero mean : -1141.211
##  ARIMA(4,0,2) with non-zero mean : -1142.203

```

```

##  ARIMA(4,0,4) with non-zero mean : Inf
##  ARIMA(3,0,3) with zero mean      : -1153.257
##  ARIMA(2,0,3) with zero mean      : -1144.865
##  ARIMA(3,0,2) with zero mean      : -1146.052
##  ARIMA(4,0,3) with zero mean      : Inf
##  ARIMA(3,0,4) with zero mean      : -1151.301
##  ARIMA(2,0,2) with zero mean      : -1138.6
##  ARIMA(2,0,4) with zero mean      : -1143.18
##  ARIMA(4,0,2) with zero mean      : -1144.171
##  ARIMA(4,0,4) with zero mean      : Inf
##
##  Best model: ARIMA(3,0,3) with zero mean

```

aic_model

```

## Series: y_train$return
## ARIMA(3,0,3) with zero mean
##
## Coefficients:
##             ar1      ar2      ar3      ma1      ma2      ma3
##       -0.6529  0.4189  0.8167  0.5909 -0.2907 -0.7253
##   s.e.    0.1662  0.1706  0.1047  0.1452  0.2060  0.2069
##
## sigma^2 estimated as 0.0003827: log likelihood=583.63
## AIC=-1153.26   AICc=-1152.75   BIC=-1129.16

```

```
bic_model=auto.arima(y_train$return,trace=TRUE,d=0,max.p=10,max.q=10,ic="bic",approximation=FALSE)
```

```

##
##  ARIMA(2,0,2) with non-zero mean : -1115.979
##  ARIMA(0,0,0) with non-zero mean : -1126.421
##  ARIMA(1,0,0) with non-zero mean : -1125.517
##  ARIMA(0,0,1) with non-zero mean : -1124.854
##  ARIMA(0,0,0) with zero mean     : -1131.809
##  ARIMA(1,0,1) with non-zero mean : -1120.413
##
##  Best model: ARIMA(0,0,0) with zero mean

```

bic_model

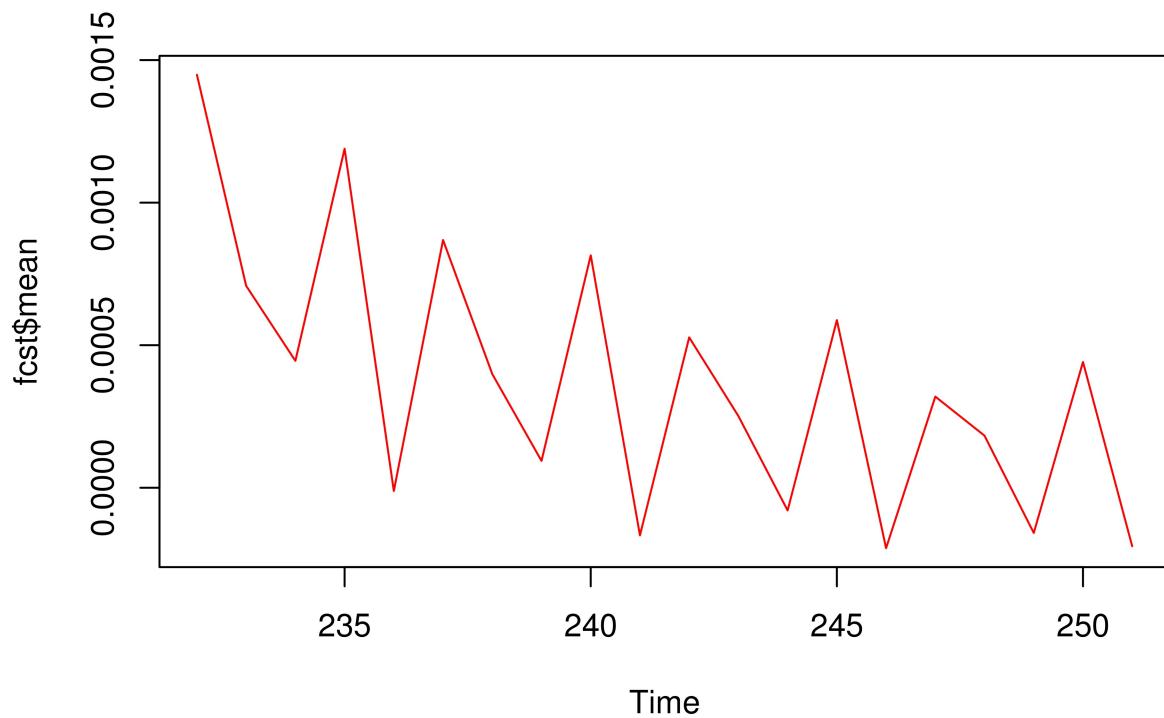
```

## Series: y_train$return
## ARIMA(0,0,0) with zero mean
##
## sigma^2 estimated as 0.000426: log likelihood=568.63
## AIC=-1135.25   AICc=-1135.23   BIC=-1131.81

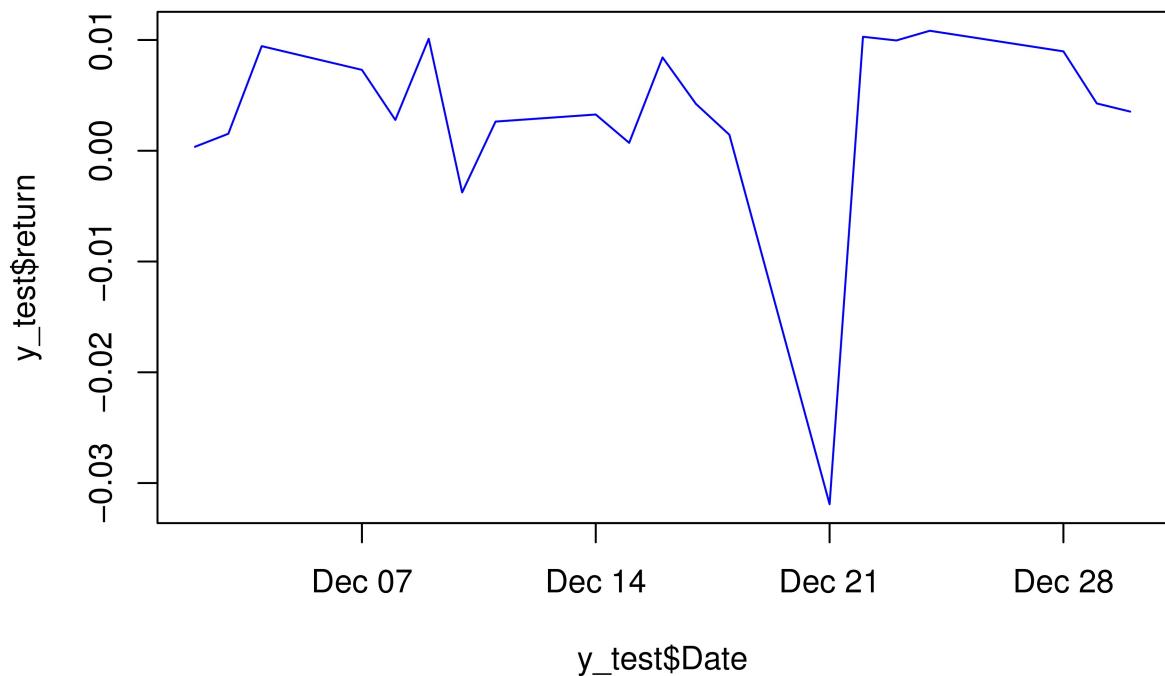
```

##Forecasting using AIC_model

```
fcst=forecast(aic_model,h=20)
plot(fcst$mean,col='red')
```

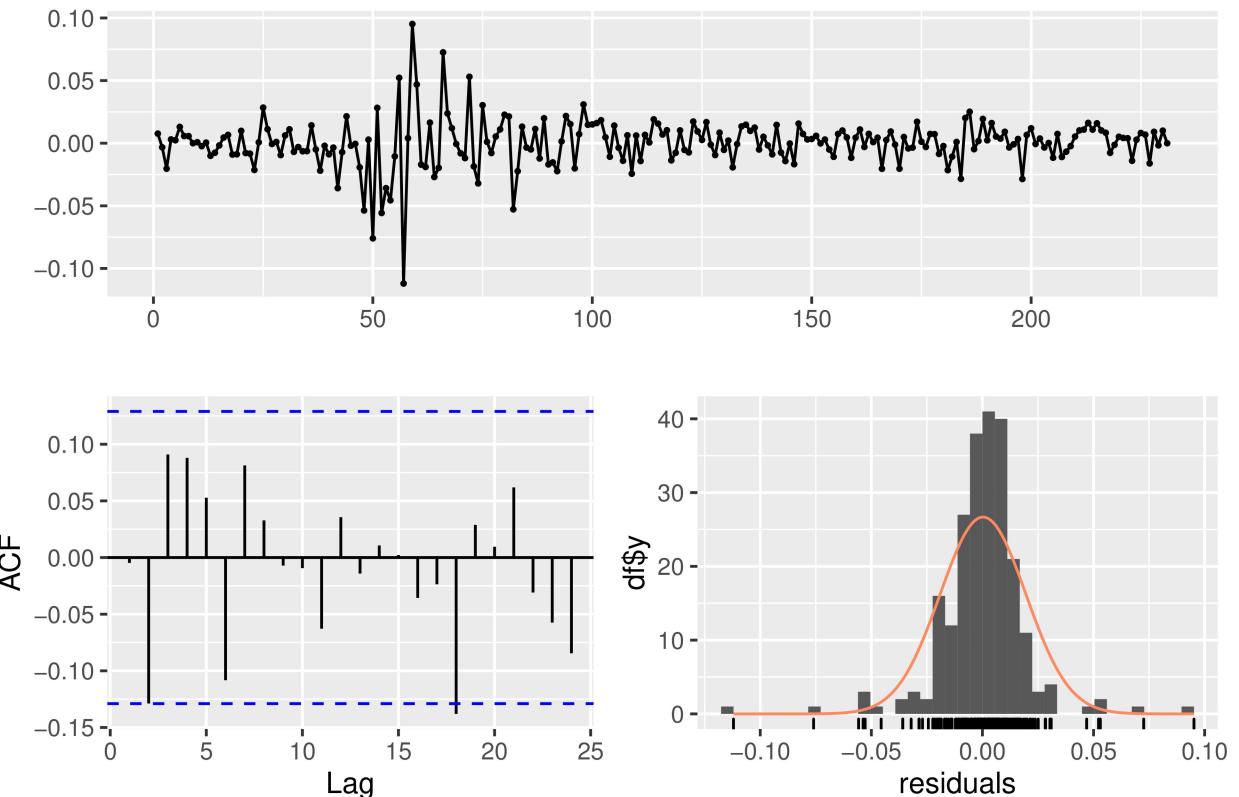


```
plot(y_test$Date,y_test$return,type='l',col='blue')
```



```
checkresiduals(aic_model)
```

Residuals from ARIMA(3,0,3) with zero mean



```
##  
## Ljung-Box test  
##  
## data: Residuals from ARIMA(3,0,3) with zero mean  
## Q* = 13.049, df = 4, p-value = 0.01104  
##  
## Model df: 6. Total lags used: 10
```

####Calculating MAPE

```
p1 = fcst$mean  
obs1 = y_test$return  
MSE_forecast1 = mean(abs(p1 - obs1))  
MSE_forecast1
```

```
## [1] 0.006625704
```

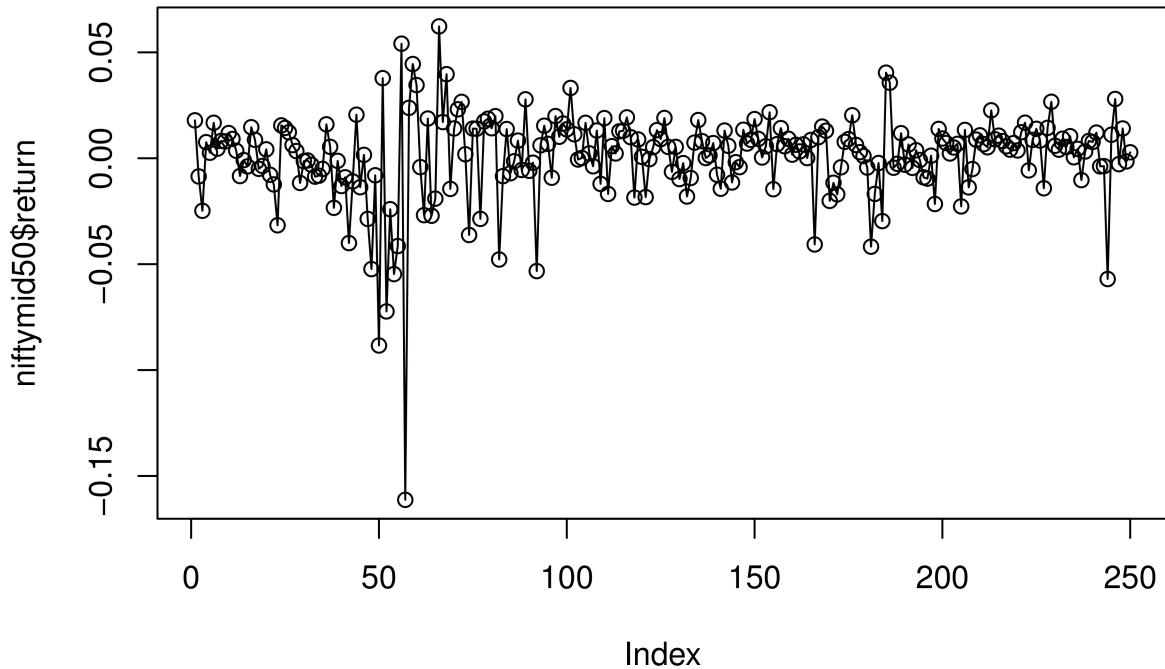
####ARIMA for Nifty midcap50

```
rm(list=ls())  
niftymid50=read.csv('nifty_midcap50.csv')  
niftymid50>Date <- as.Date(niftymid50>Date)  
niftymid50$Close=as.numeric(as.character(niftymid50$Close))  
price1=niftymid50$Close
```

```

log_return1=diff(log(price1),lag=1)
niftymid50=niftymid50[-1,]
niftymid50['return']=log_return1
plot(niftymid50$return,type="o")

```



```
adf.test(niftymid50$return, alternative = "stationary")
```

```

## Warning in adf.test(niftymid50$return, alternative = "stationary"): p-value
## smaller than printed p-value

```

```

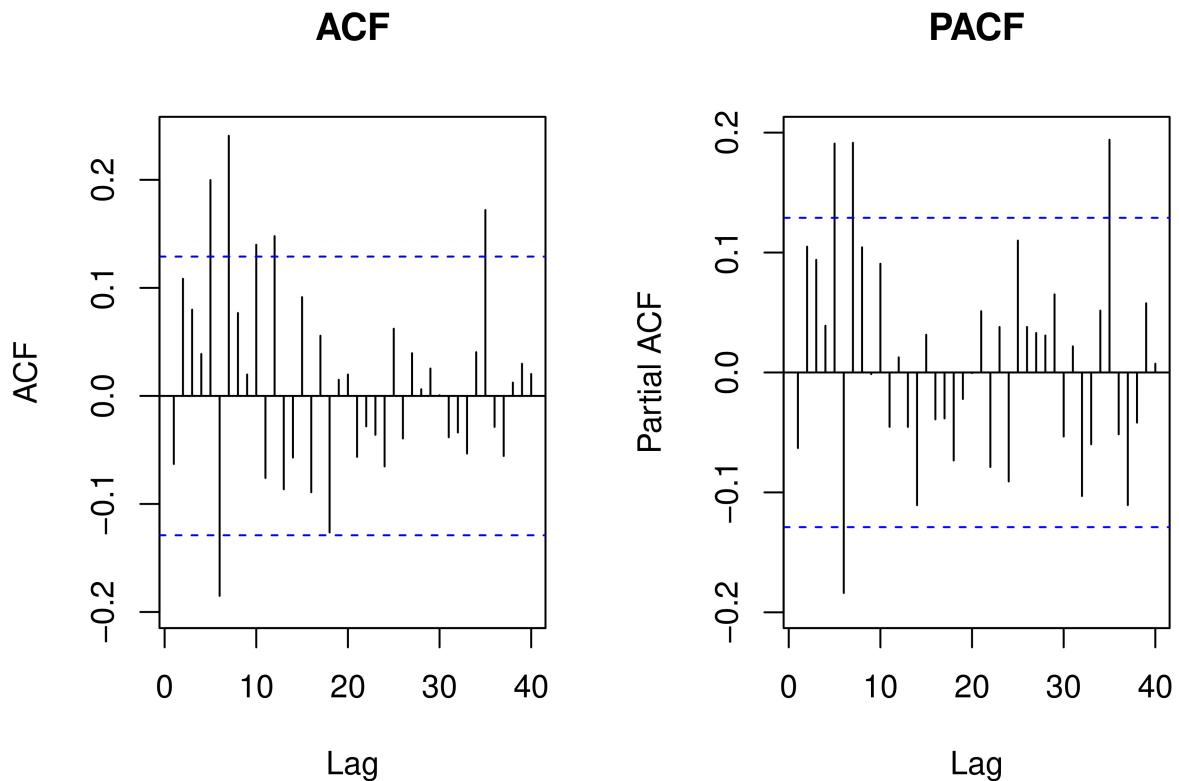
##
## Augmented Dickey-Fuller Test
##
## data: niftymid50$return
## Dickey-Fuller = -4.5885, Lag order = 6, p-value = 0.01
## alternative hypothesis: stationary

```

```

par(mfrow=c(1,2))
y_trainmid=niftymid50[1:231,]
y_testmid=niftymid50[231:250,]
Acf(y_trainmid$return, lag.max = 40, main='ACF')
Pacf(y_trainmid$return, lag.max = 40, main = 'PACF')

```



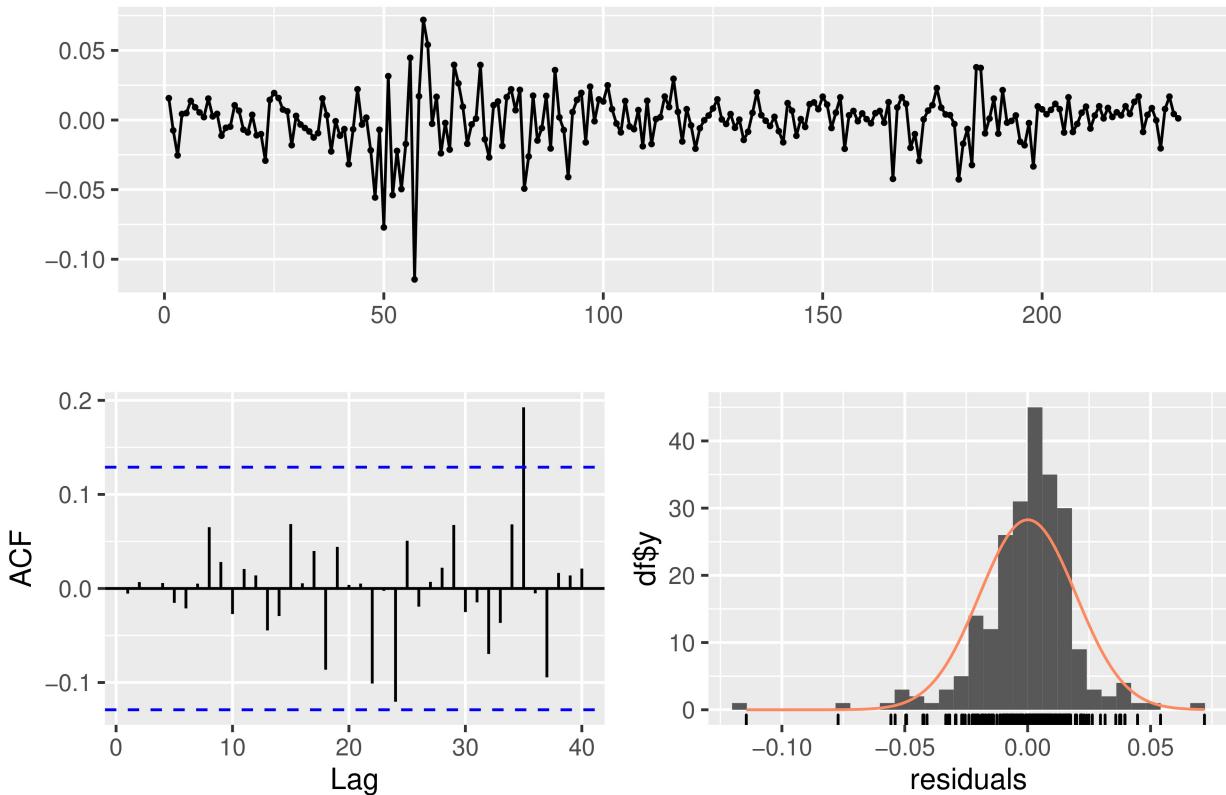
```

best.order = c(0, 0, 0)
best.aic = Inf
for (q in 1:10) for (p in 1:10) {
  fit.model= arima(y_trainmid$return, order = c(p, 0, q), optim.control = list(maxit = 2000) ,method='ML')
  fit.aic = fit.model$aic
  if (fit.aic < best.aic) {
    best.order = c(p, 0, q)
    best.arima = fit.model
    best.aic = fit.aic
  }
}
## Warning in log(s2): NaNs produced
best.order; best.aic
## [1] 7 0 4
## [1] -1132.937
model.fit = best.arima

```

```
checkresiduals(model.fit, lag.max = 40)
```

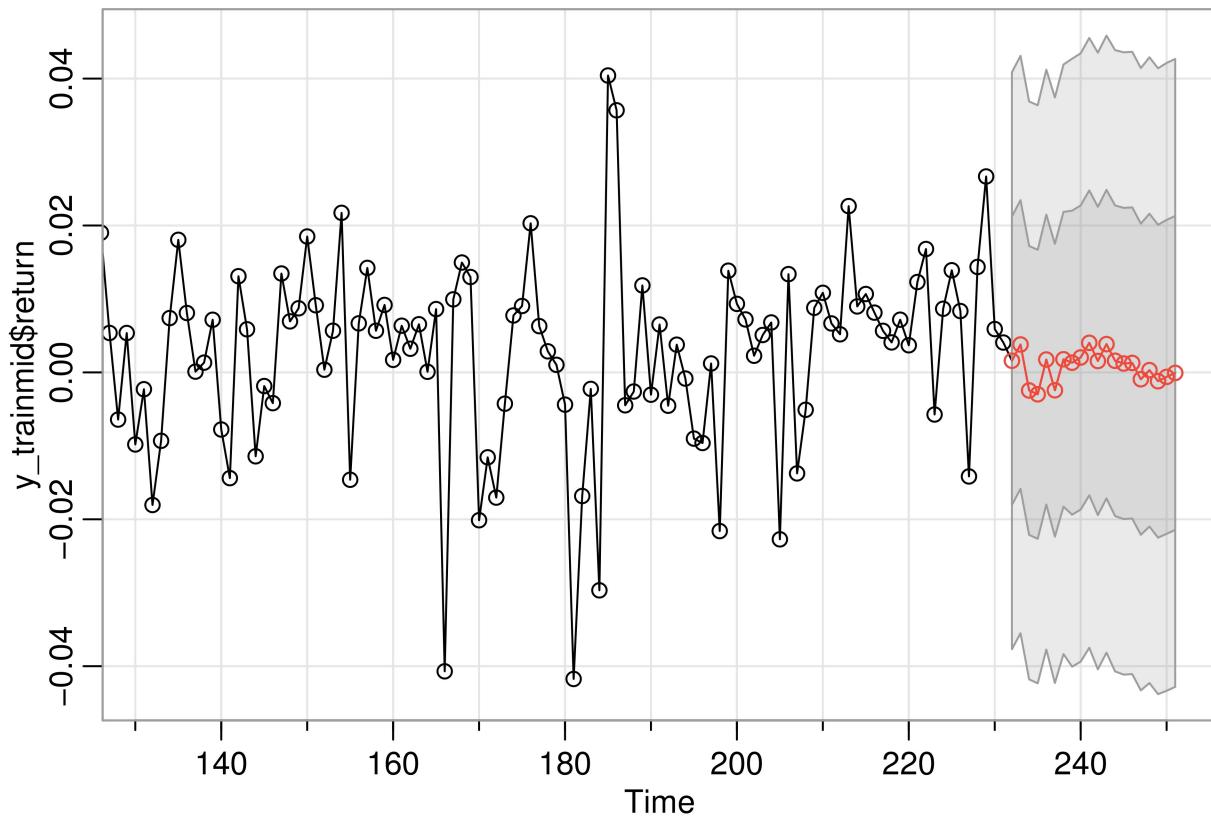
Residuals from ARIMA(7,0,4) with non-zero mean



```
##  
## Ljung-Box test  
##  
## data: Residuals from ARIMA(7,0,4) with non-zero mean  
## Q* = 3.6149, df = 3, p-value = 0.3062  
##  
## Model df: 12. Total lags used: 15
```

Violations of assumption here so ARCH can be fit here

```
forecast2 = sarima.for(y_trainmid$return, n.ahead = length(y_testmid$return), p=7, d=0, q=4)  
  
## Warning in log(s2): NaNs produced  
## Warning in log(s2): NaNs produced  
## Warning in log(s2): NaNs produced
```



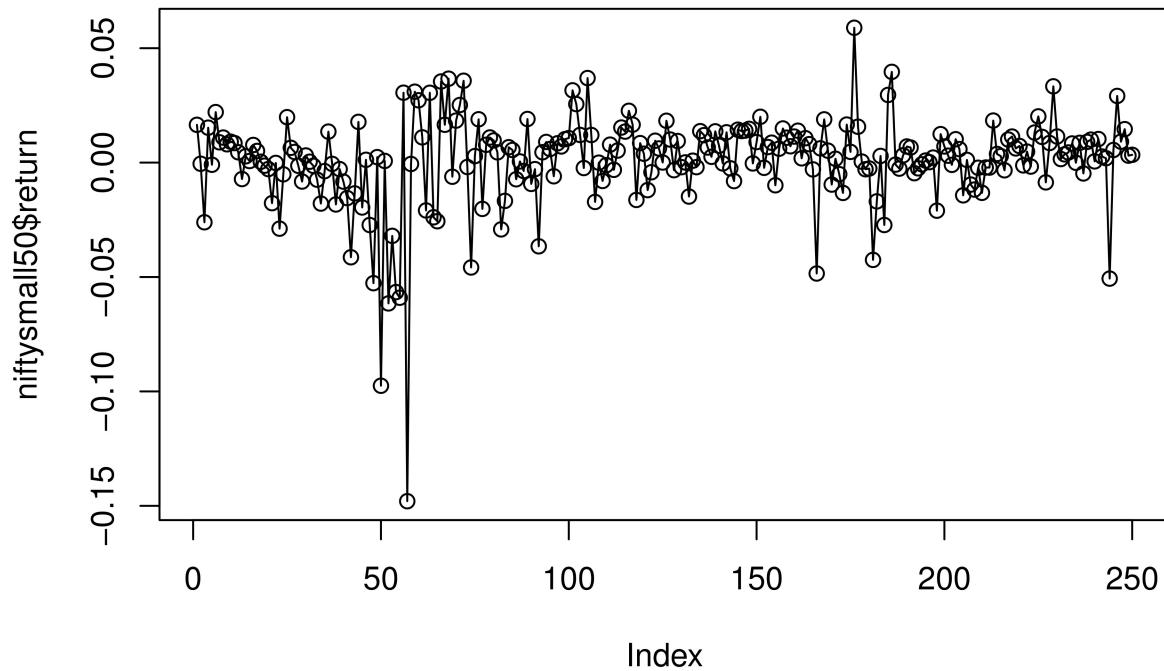
```
p2 = forecast2$pred
obs2 = y_testmid$return
MSE_forecast2 = mean(abs((p2 - obs2)))
MSE_forecast2
```

```
## [1] 0.01018547
```

```
####
```

```
###For nifty small cap
```

```
rm(list=ls())
niftysmall150=read.csv('nifty_smallcap50.csv')
niftysmall150>Date <- as.Date(niftysmall150>Date)
niftysmall150$Close=as.numeric(as.character(niftysmall150$Close))
price1=niftysmall150$Close
log_return2=diff(log(price1),lag=1)
niftysmall150=niftysmall150[-1,]
niftysmall150['return']=log_return2
plot(niftysmall150$return,type="o")
```



```

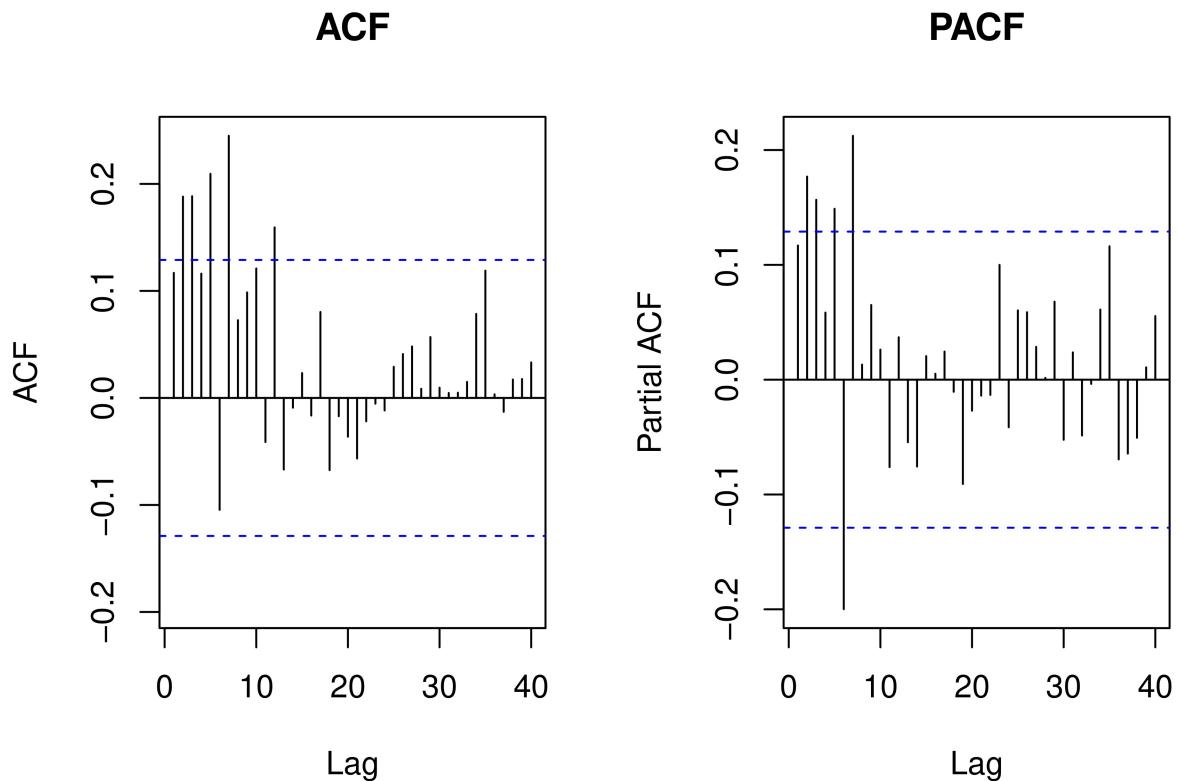
adf.test(niftysmall150$return, alternative = "stationary")

## Warning in adf.test(niftysmall150$return, alternative = "stationary"): p-value
## smaller than printed p-value

##
## Augmented Dickey-Fuller Test
##
## data: niftysmall150$return
## Dickey-Fuller = -4.2713, Lag order = 6, p-value = 0.01
## alternative hypothesis: stationary

par(mfrow=c(1,2))
y_trainsmall=niftysmall150[1:231,]
y_testsma...

```



```
aic_model1=auto.arima(y_trainsmall$return,trace=TRUE,d=0,max.p=10,max.q=10,ic="aic",approximation=FALSE)
```

```
##  
## ARIMA(2,0,2) with non-zero mean : -1144.999  
## ARIMA(0,0,0) with non-zero mean : -1134.417  
## ARIMA(1,0,0) with non-zero mean : -1135.591  
## ARIMA(0,0,1) with non-zero mean : -1134.792  
## ARIMA(0,0,0) with zero mean : -1136.184  
## ARIMA(1,0,2) with non-zero mean : -1146.998  
## ARIMA(0,0,2) with non-zero mean : -1138.085  
## ARIMA(1,0,1) with non-zero mean : -1146.928  
## ARIMA(1,0,3) with non-zero mean : -1145.002  
## ARIMA(0,0,3) with non-zero mean : -1143.255  
## ARIMA(2,0,1) with non-zero mean : -1146.915  
## ARIMA(2,0,3) with non-zero mean : -1149.802  
## ARIMA(3,0,3) with non-zero mean : -1156.435  
## ARIMA(3,0,2) with non-zero mean : -1150.845  
## ARIMA(4,0,3) with non-zero mean : -1154.838  
## ARIMA(3,0,4) with non-zero mean : -1155.178  
## ARIMA(2,0,4) with non-zero mean : -1148.629  
## ARIMA(4,0,2) with non-zero mean : -1149.722  
## ARIMA(4,0,4) with non-zero mean : -1144.885  
## ARIMA(3,0,3) with zero mean : -1158.332  
## ARIMA(2,0,3) with zero mean : -1151.702  
## ARIMA(3,0,2) with zero mean : -1152.745
```

```

##  ARIMA(4,0,3) with zero mean      : Inf
##  ARIMA(3,0,4) with zero mean      : -1157.07
##  ARIMA(2,0,2) with zero mean      : -1146.893
##  ARIMA(2,0,4) with zero mean      : -1150.526
##  ARIMA(4,0,2) with zero mean      : -1151.62
##  ARIMA(4,0,4) with zero mean      : -1146.776
##
##  Best model: ARIMA(3,0,3) with zero mean

```

```
aic_model1
```

```

## Series: y_trainsmall$return
## ARIMA(3,0,3) with zero mean
##
## Coefficients:
##          ar1     ar2     ar3     ma1     ma2     ma3
##        -0.7801  0.5784  0.7519  0.8740 -0.3704 -0.5178
##  s.e.    0.0985  0.1324  0.0894  0.1268  0.1840  0.1137
##
## sigma^2 estimated as 0.0003748: log likelihood=586.17
## AIC=-1158.33   AICc=-1157.83   BIC=-1134.23

```

```
bic_model1=auto.arima(y_trainsmall$return,trace=TRUE,d=0,max.p=10,max.q=10,ic="bic",approximation=FALSE)
```

```

##
##  ARIMA(2,0,2) with non-zero mean : -1124.345
##  ARIMA(0,0,0) with non-zero mean : -1127.532
##  ARIMA(1,0,0) with non-zero mean : -1125.264
##  ARIMA(0,0,1) with non-zero mean : -1124.465
##  ARIMA(0,0,0) with zero mean    : -1132.741
##  ARIMA(1,0,1) with non-zero mean : -1133.158
##  ARIMA(2,0,1) with non-zero mean : -1129.703
##  ARIMA(1,0,2) with non-zero mean : -1129.786
##  ARIMA(0,0,2) with non-zero mean : -1124.316
##  ARIMA(2,0,0) with non-zero mean : -1127.129
##  ARIMA(1,0,1) with zero mean   : -1138.487
##  ARIMA(0,0,1) with zero mean   : -1129.705
##  ARIMA(1,0,0) with zero mean   : -1130.516
##  ARIMA(2,0,1) with zero mean   : -1135.038
##  ARIMA(1,0,2) with zero mean   : -1135.122
##  ARIMA(0,0,2) with zero mean   : -1129.583
##  ARIMA(2,0,0) with zero mean   : -1132.426
##  ARIMA(2,0,2) with zero mean   : -1129.681
##
##  Best model: ARIMA(1,0,1) with zero mean

```

```
bic_model1
```

```

## Series: y_trainsmall$return
## ARIMA(1,0,1) with zero mean
##
## Coefficients:

```

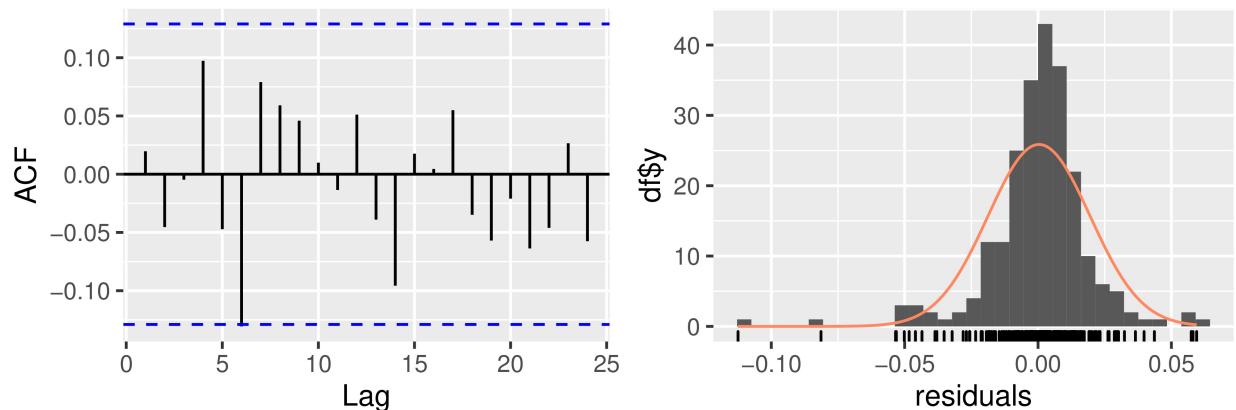
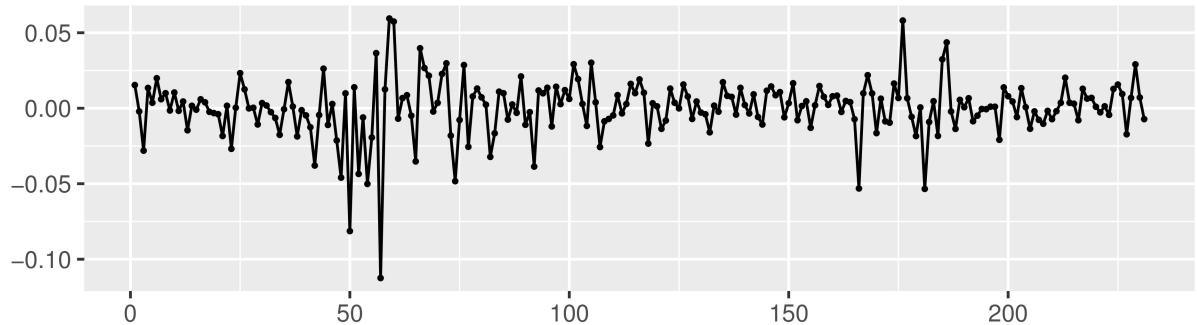
```

##          ar1      ma1
##     0.9019 -0.7846
## s.e.  0.0584  0.0806
##
## sigma^2 estimated as 0.000398:  log likelihood=577.41
## AIC=-1148.81   AICc=-1148.71   BIC=-1138.49

```

```
checkresiduals(aic_model1)
```

Residuals from ARIMA(3,0,3) with zero mean



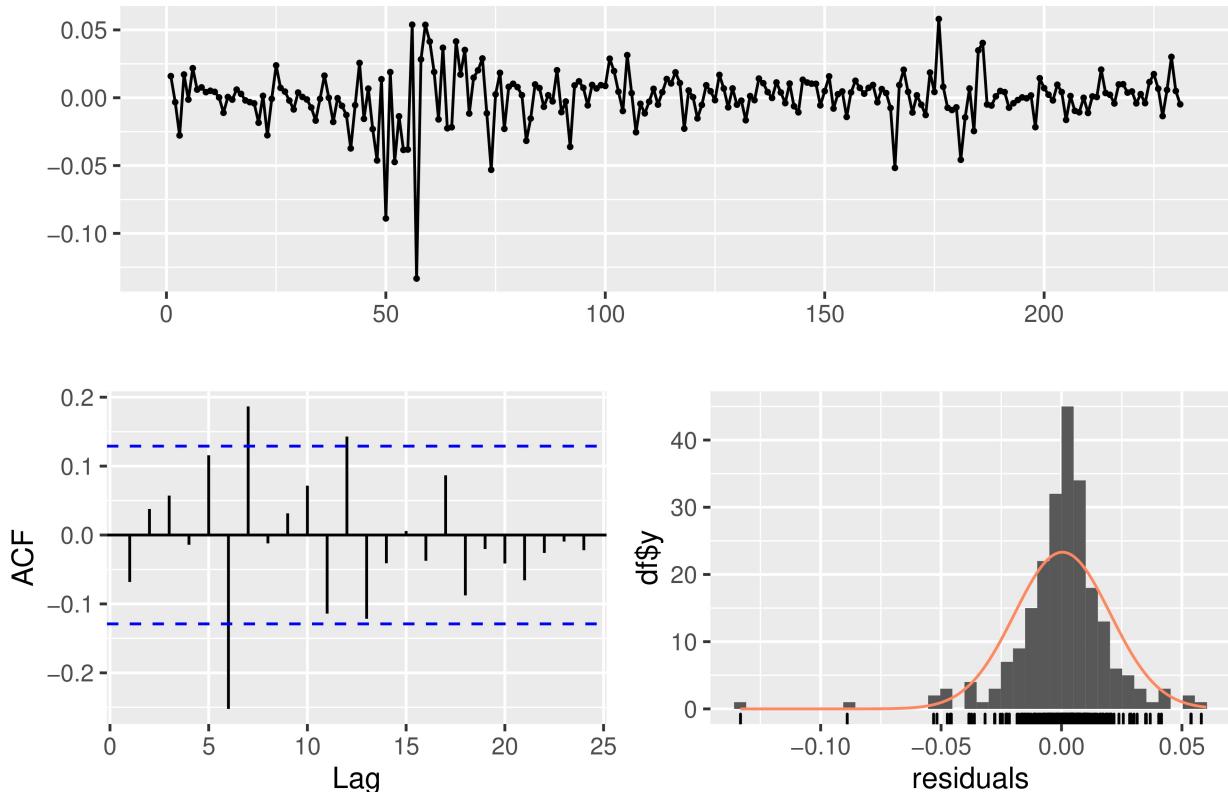
```

## 
## Ljung-Box test
## 
## data: Residuals from ARIMA(3,0,3) with zero mean
## Q* = 10.32, df = 4, p-value = 0.03536
## 
## Model df: 6. Total lags used: 10

```

```
checkresiduals(bic_model1)
```

Residuals from ARIMA(1,0,1) with zero mean



```
##  
## Ljung-Box test  
##  
## data: Residuals from ARIMA(1,0,1) with zero mean  
## Q* = 30.543, df = 8, p-value = 0.0001694  
##  
## Model df: 2. Total lags used: 10
```

Taking the aic model because bic model violates the assumptions

```
fcst1=forecast(aic_model1,h=20)  
p3 = fcst1$mean  
obs3 = y_testsmall$return  
MSE_forecast3 = mean(abs(p3 - obs3))  
MSE_forecast3
```

```
## [1] 0.008600537
```