

## Blue shows final answers

In this project, I am going to research [REDACTED] the symbol of the stock is [REDACTED]

```
> #Install package 'quantmod' and load it.
> install.packages("quantmod")
Installing package into 'C:\ntdata\Personal\HTtrinh\R\win-library\3.5'
(as 'lib' is unspecified)
trying URL 'https://cran.rstudio.com/bin/windows/contrib/3.5/quantmod_0.4-14.zip'
Content type 'application/zip' length 960802 bytes (938 KB)
downloaded 938 KB
```

package 'quantmod' successfully unpacked and MD5 sums checked

The downloaded binary packages are in

C:\Users\httrinh\AppData\Local\Temp\RtmpCqTW3i\downloaded\_packages

```
> library(quantmod)
```

Loading required package: xts

Loading required package: zoo

Attaching package: 'zoo'

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

as.Date, as.Date.numeric

Loading required package: TTR

Version 0.4-0 included new data defaults. See ?getSymbols.

Learn from a quantmod author: <https://www.datacamp.com/courses/importing-and-managing-financial-data-in-r>

```
> #Download daily prices of stock [REDACTED]
> getSymbols("[REDACTED]src="yahoo", from=[REDACTED])
'getSymbols' currently uses auto.assign=TRUE by default, but will
use auto.assign=FALSE in 0.5-0. You will still be able to use
'loadSymbols' to automatically load data. getOption("getSymbols.env")
and getOption("getSymbols.auto.assign") will still be checked for
alternate defaults.
```

This message is shown once per session and may be disabled by setting `options("getSymbols.warning4.0"=FALSE)`. See ?getSymbols for details.

```
[1] "[REDACTED]"
> #Obtain monthly log stock returns
> [REDACTED]n <- monthlyReturn([REDACTED]subset=NULL, type='log', leading=TRUE)
> #Change the data type to ts
> rtn <- ts([REDACTED], frequency=[REDACTED])
> # Install and import axtsa package
> install.packages("axtsa")
Installing package into 'C:\ntdata\Personal\HTtrinh\R\win-library\3.5'
(as 'lib' is unspecified)
trying URL 'https://cran.rstudio.com/bin/windows/contrib/3.5/axtsa_1.8.zip'
Content type 'application/zip' length 770254 bytes (752 KB)
downloaded 752 KB
```

package 'axtsa' successfully unpacked and MD5 sums checked

The downloaded binary packages are in

C:\Users\ [REDACTED] \AppData\Local\Temp\RtmpCqTW3i\downloaded\_packages

> library(astsa)

> # Fit an ARMA

> BA\_fit <-

initial value

iter 2 value

iter 3 value

iter 4 value

iter 4 value

iter 4 value

final value

converged

initial value

iter 2 value

iter 3 value

iter 3 value

iter 3 value

final value

converged

> BA\_fit

\$fit

Call:

stats::arima

Q), per

'

REPORT =

Coefficients

ar1

0.0925

s.e. 0.0678

sigma^2 esti

od = 233.41, aic = -460.81

\$degrees\_of

[1] 214

\$ttable

Estima

ar1 0.09

xmean 0.01

\$AIC

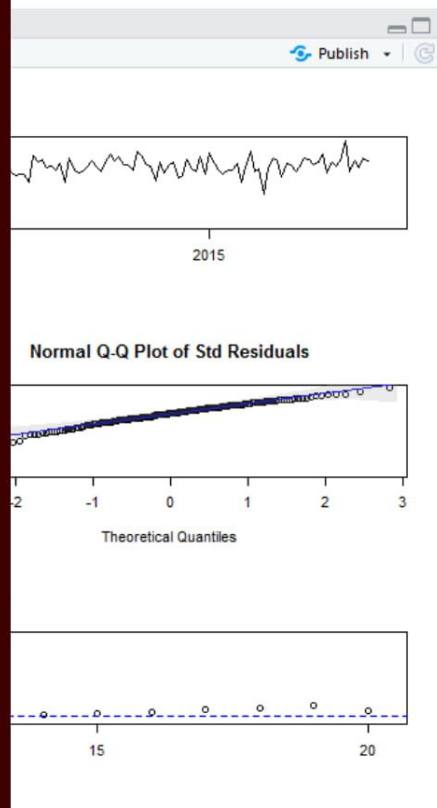
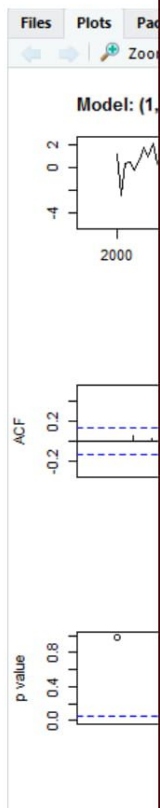
[1] -3.98056

\$AICc

[1] -3.97078

\$BIC

[1] -4.9020



```
> # Fit an ARIMA model
> BA_fit <- arima(
initial val
iter 2 v
iter 3 v
iter 4 v
iter 5 v
iter 5 v
iter 5 v
final val
converged
initial val
iter 2 v
iter 3 v
iter 4 v
iter 4 v
iter 4 v
final val
converged
> BA_fit
$fit
```

```
Call:
stats::ari
Q), pe
>
REPORT
```

```
Coefficien
a
0.09
s.e. 0.06
sigma^2 es
```

```
seasonal = list(order = c(P, D,
n = FALSE, optim.control = list(trace = trc
```

```
= 233.41, aic = -458.82
```

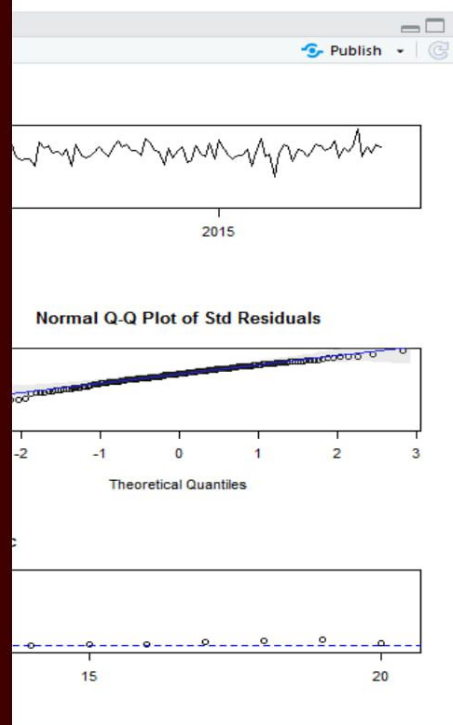
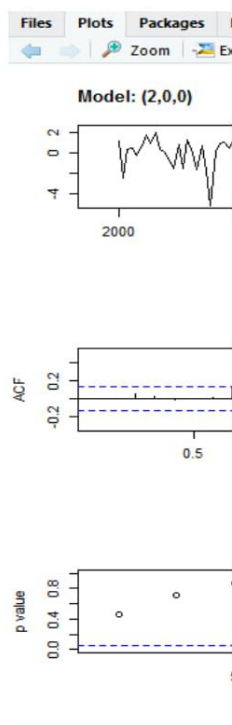
```
$degrees_of_freedom
[1] 213
```

```
$table
      Estimate
ar1      0.0921
ar2      0.0048
xmean    0.0111
```

```
$AIC
[1] -3.971329
```

```
$AICc
[1] -3.961192
```

```
$BIC
[1] -4.92445
```



```
> # Fit an MA(3) model
> BA_fit <- sar
initial value
iter 2 value
iter 3 value
iter 4 value
iter 4 value
iter 4 value
final value -2
converged
initial value
iter 2 value
iter 3 value
iter 3 value
iter 3 value
final value -2
converged
> BA_fit
$fit
```

```
Call:
stats::arima(x = x, order = c(P, D, Q), period = 1,
              REPORT = 1, ...)

Coefficients:
          ma1
0.0891 0.0096
s.e. 0.0681 0.0096

sigma^2 estimate = 233.65, aic = -457.3

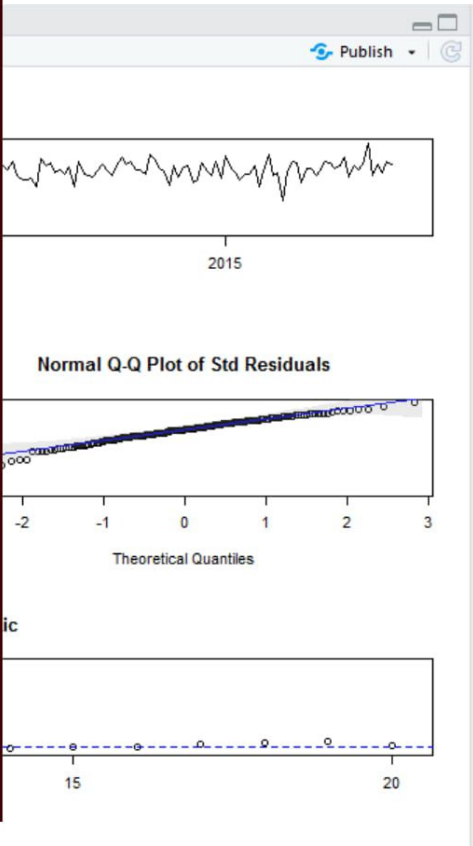
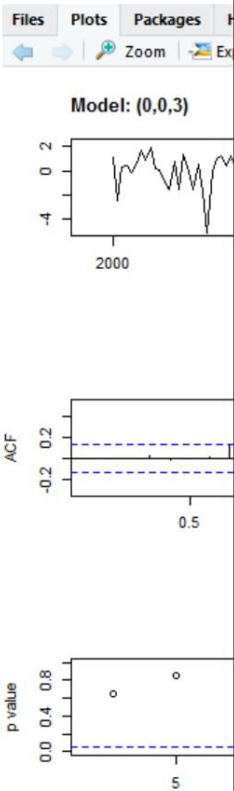
$degrees_of_freedom
[1] 212

$table
      Estimate
ma1      0.0891
ma2      0.0096
ma3      0.0491
xmean    0.0111

$AIC
[1] -3.964339

$AICc
[1] -3.953757

$BIC
[1] -4.901834
```



```

> # Fit an MA(
> A_fit <- sar
initial value
iter 2 value
iter 3 value
iter 4 value
iter 5 value
iter 5 value
iter 5 value
final value -
converged
initial value
iter 2 value
iter 3 value
iter 3 value
iter 3 value
final value -
converged
> # Fit an MA(
> BA_fit <- sa
initial value
iter 2 value
iter 3 value
iter 4 value
iter 5 value
iter 5 value
iter 5 value
final value -
converged
initial value
iter 2 value
iter 3 value
iter 3 value
iter 3 value
final value -
converged
> BA_fit
$fit

Call:
stats::arima(x
  Q), period
'
REPORT = 1

Coefficients:
      ma1
0.0921
s.e. 0.0684

sigma^2 estimate = 233.74, aic = -455.48

$degrees_of_freedom
[1] 211

$table
      Estimate
ma1      0.0921
ma2      0.0099
ma3      0.0515
ma4      0.0260
xmean    0.0112
$AIC

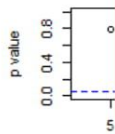
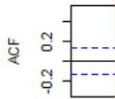
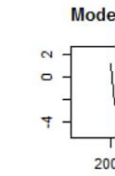
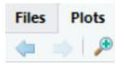
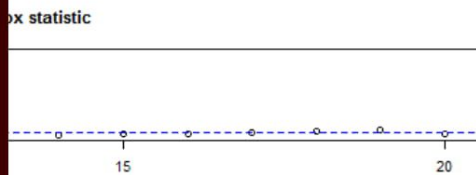
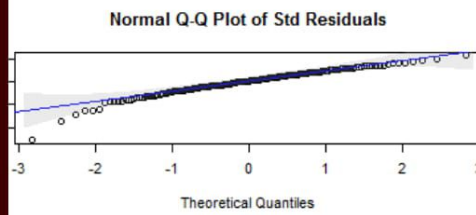
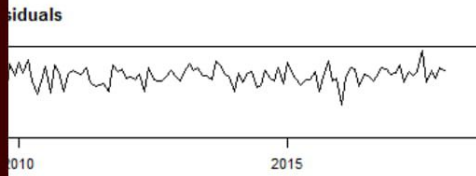
```

[1] -3.

\$AICc  
[1] -3.

\$BIC

[1] -4.

[illegible]

```
data rtn
= 3)
```



```
iter 23 value -2.513510
iter 24 value -2.513518
iter 25 value -2.513532
iter 26 value -2.513552
iter 27 value -2.513569
iter 28 value -2.513570
iter 29 value -2.513571
iter 30 value -2.513571
iter 31 value -2.513572
iter 32 value -2.513572
iter 33 value -2.513572
iter 34 value -2.513572
iter 34 value -2.513572
iter 34 value -2.513572
final value -2.513572
converged
initial value -2.500212
iter 2 value -2.500348
iter 3 value -2.500420
iter 4 value -2.500425
iter 5 value -2.500433
iter 6 value -2.500444
iter 7 value -2.500474
iter 8 value -2.500534
iter 9 value -2.500663
iter 10 value -2.500820
iter 11 value -2.501016
iter 12 value -2.501214
iter 13 value -2.501223
iter 14 value -2.501229
iter 15 value -2.501232
iter 16 value -2.501253
iter 17 value -2.501400
iter 18 value -2.501812
iter 19 value -2.501980
iter 20 value -2.502053
iter 21 value -2.502170
iter 22 value -2.502205
iter 23 value -2.502248
iter 24 value -2.502521
iter 25 value -2.502802
iter 26 value -2.503400
iter 27 value -2.504166
iter 28 value -2.506247
iter 29 value -2.506583
iter 30 value -2.506946
iter 31 value -2.507693
iter 32 value -2.507716
iter 33 value -2.508068
iter 34 value -2.508405
iter 35 value -2.508595
iter 36 value -2.508766
iter 37 value -2.509012
iter 38 value -2.509212
iter 39 value -2.509852
iter 40 value -2.512441
iter 41 value -2.513771
iter 42 value -2.515239
iter 43 value -2.515856
iter 44 value -2.516091
iter 45 value -2.516395
iter 46 value -2.516587
iter 47 value -2.516689
iter 48 value -2.516707
```



```
iter 49 value -2.516711
iter 50 value -2.516713
iter
iter
iter
final
conver
> BA_f
$fit

Call:
stats::arima.model(aic = 237.12, seasonal = list(order = c(P, D,
Q), include.mean = FALSE, optim.control = list(trace = trc
, RE

Coeffi
          ma3    xmean
s.e.      0.1321 0.0112
          0.0778 0.0064
sigma^2    likelihood = 237.12, aic = -460.24

$degree
[1] 21

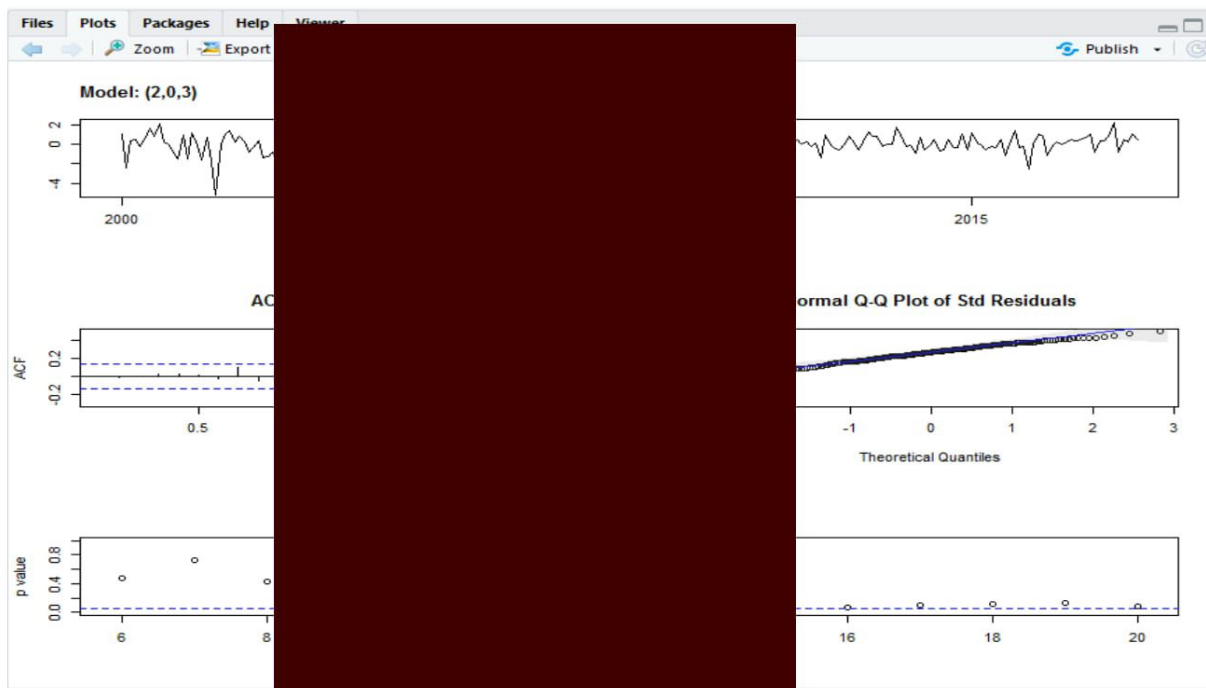
$ttabl

ar1
ar2
ma1
ma2
ma3
xmean

$AIC
[1] -3

$AICc
[1] -3

$BIC
[1] -4
```



```
> # Fit an ARMA(3
> BA_fit <- sarim
initial value -2
iter 2 value -2
iter 3 value -2
iter 4 value -2
iter 5 value -2
iter 6 value -2
iter 7 value -2
iter 8 value -2
iter 9 value -2
iter 10 value -2
iter 11 value -2
iter 12 value -2
iter 13 value -2
iter 14 value -2
iter 15 value -2
iter 16 value -2
iter 17 value -2
iter 18 value -2
iter 19 value -2
iter 20 value -2
iter 20 value -2
iter 20 value -2
final value -2.5
converged
initial value -2
iter 2 value -2
iter 3 value -2
iter 4 value -2
iter 5 value -2
iter 6 value -2
iter 7 value -2
iter 8 value -2
iter 9 value -2
iter 10 value -2
iter 11 value -2
iter 12 value -2
iter 13 value -2
```

```
iter 14 value -2.501518
iter 15 value -2.501543
iter 16 value -2.501579
iter 17 value -2.501631
iter 18 value -2.501799
iter 19 value -2.501854
iter 20 value -2.501888
iter 21 value -2.501950
iter 22 value -2.502035
iter 23 value -2.502268
iter 24 value -2.502874
iter 25 value -2.502885
iter 26 value -2.504055
iter 27 value -2.504526
iter 28 value -2.504543
iter 29 value -2.504992
iter 30 value -2.505170
iter 31 value -2.505391
iter 32 value -2.505616
iter 33 value -2.506018
iter 34 value -2.506365
iter 35 value -2.507189
iter 36 value -2.507338
iter 37 value -2.507345
iter 38 value -2.507773
iter 39 value -2.508413
iter 40 value -2.509474
iter 41 value -2.510496
iter 42 value -2.511216
iter 43 value -2.511868
iter 44 value -2.513127
iter 45 value -2.513212
iter 46 value -2.513984
iter 47 value -2.515371
iter 48 value -2.515542
iter 49 value -2.516381
iter 50 value -2.516608
iter 51 value -2.516827
iter 52 value -2.516842
iter 53 value -2.516847
iter 54 value -2.516848
iter 55 value -2.516848
iter 55 value -2.516848
final
converge
> BA_fit
$fit

Call:
stats::arima(, seasonal = list(order = c(P, D,
Q),
,
REPO

Coefficients:
0
s.e. 2
sigma^2
likelihood = 237.15, aic = -456.3

$degrees
[1] 208
```

```
$ttable
```

```
      Estimate  
ar1      0.3096 2.11  
ar2     -0.8520 0.8  
ar3     -0.1016 1.9  
ma1     -0.2285 2.1  
ma2      0.8836 0.6  
ma3      0.2420 1.9  
ma4     -0.0037 0.3  
xmean    0.0112 0.0
```

```
$AIC
```

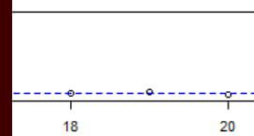
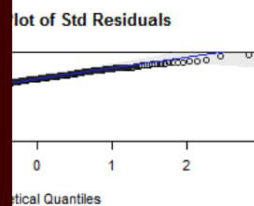
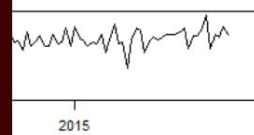
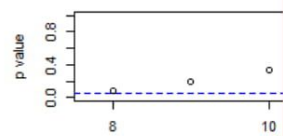
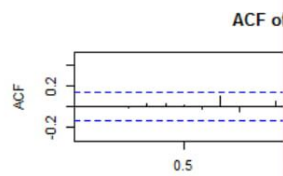
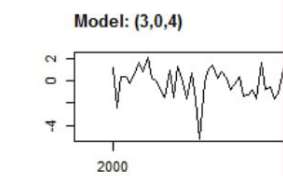
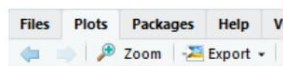
```
[1] -3.977641
```

```
$AICc
```

```
[1] -3.964336
```

```
$BIC
```

```
[1] -4.85263
```



```
> # Fit an ARIMA(1,0,4) model  
> BA_fit <- sarima(1,0,4)  
initial value -1.1  
iter 2 value -1.1  
iter 3 value -1.1  
iter 4 value -1.1  
iter 5 value -1.1  
iter 6 value -1.1  
iter 7 value -1.1  
iter 8 value -2.1  
iter 9 value -2.1  
iter 10 value -2.1
```

```

iter 11 value -2.185620
iter 12 value -2.236023
iter 13 value -2.244500
iter 14 value -2.251062
iter 15 value -2.254496
iter 16 value -2.257668
iter 17 value -2.262104
iter 18 value -2.265271
iter 19 value -2.267407
iter 20 value -2.268683
iter 21 value -2.269991
iter 22 value -2.270901
iter 23 value -2.271324
iter 24 value -2.271773
iter 25 value -2.272299
iter 26 value -2.272306
iter 27 value -2.272313
iter 28 value -2.272342
iter 29 value -2.272343
iter 30 value -2.272348
iter 31 value -2.272359
iter 32 value -2.273955
iter 33 value -2.274233
iter 34 value -2.274344
iter 34 value -2.274344
iter 34 value -2.274344
final value -2.274344

```

converged

```

initial value -2.242270
iter 2 value -2.247272
iter 3 value -2.261307
iter 4 value -2.261601
iter 5 value -2.265804
iter 6 value -2.266020
iter 7 value -2.266322
iter 8 value -2.266381
iter 9 value -2.266383
iter 10 value -2.266386
iter 11 value -2.266391

```

```

iter 12 value
iter 12 value
iter 13 value
iter 14 value
iter 14 value
iter 14 value

```

```
final value -
```

converged

```
> BA_fit
```

```
$fit
```

Call:

```
stats::arima(
  Q), period
REPORT = 1

```

Coefficients:

```

arl
-0.4546
s.e. 0.0619

```

```
sigma^2 estima
```

```
$degrees_of_fre
```

```
[1] 210
```

```

= list(order = c(P, D,
tim.control = list(trace = trc,

```

```
.51, aic = -353.01
```

```
$ttable
```

```
Estimate
```

```
ar1 -0.4546 0.0000  
ma1 -1.9978 0.0000  
ma2 1.0000 0.0000
```

```
$AIC
```

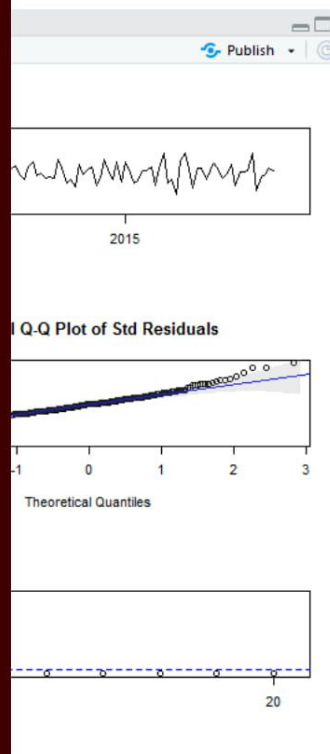
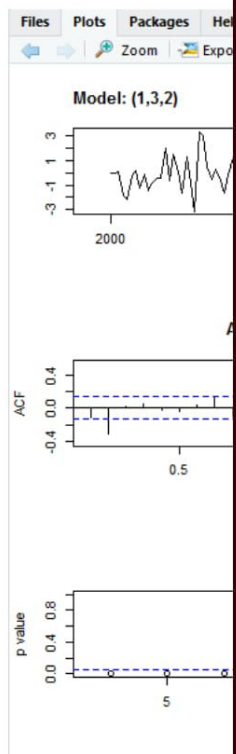
```
[1] -3.585728
```

```
$AICc
```

```
[1] -3.575591
```

```
$BIC
```

```
[1] -4.538849
```



```
> # Fit an ARIMA model  
> BA_fit <- sarima(1,3,2)  
initial value  
iter 2 value  
iter 3 value  
iter 4 value  
iter 5 value  
iter 6 value  
iter 7 value  
iter 8 value  
iter 9 value  
iter 10 value  
iter 11 value  
iter 12 value  
iter 13 value  
iter 14 value  
iter 15 value  
iter 16 value  
iter 17 value  
iter 18 value
```

```
return
```

```

iter 19 value -2.1
iter 20 value -2.1
iter 21 value -2.1
iter 22 value -2.1
iter 23 value -2.1
iter 24 value -2.1
iter 25 value -2.1
iter 26 value -2.1
iter 27 value -2.1
iter 28 value -2.1
iter 29 value -2.1
iter 29 value -2.1
final value -2.1
converged
initial value -2.1
iter 2 value -2.1
iter 3 value -2.1
iter 4 value -2.1
iter 5 value -2.1
iter 6 value -2.1
iter 7 value -2.1
iter 8 value -2.1
iter 9 value -2.1
iter 10 value -2.1
iter 10 value -2.1
final value -2.1
converged
> BA_fit
$fit

```

```

Call:
stats::arima(x =
  Q), period =
  REPORT = 1,

```

```

Coefficients:
      ar1
0.0565 -2.
s.e. 0.0939 0.

```

```

sigma^2 estimated

```

```

$degrees_of_freedom
[1] 209

```

```

$table
      Estimate
ar1 0.0565 0.09
ma1 -2.7053 0.09
ma2 2.4234 0.18
ma3 -0.7177 0.09

```

```

$AIC
[1] -3.790261

```

```

$AICc
[1] -3.779679

```

```

$BIC
[1] -4.727756

```

```

al = list(order = c(P, D,
optim.control = list(trace = trc,

```

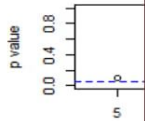
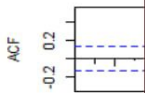
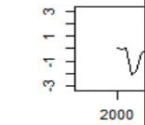
```

03.35, aic = -396.71

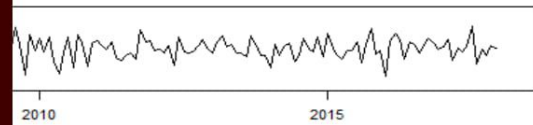
```



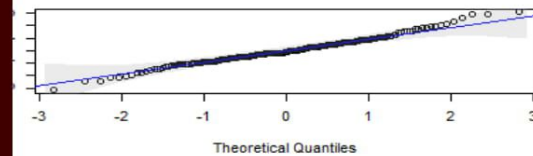
Model: (



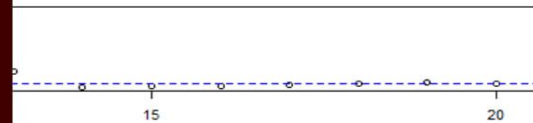
residuals



Normal Q-Q Plot of Std Residuals



Box statistic



&gt; # Fit an

&gt; BA\_fit &lt;-

initial va

iter 2 va

iter 3 va

iter 4 va

iter 5 va

iter 6 va

iter 7 va

iter 8 va

iter 9 va

iter 10 va

iter 11 va

iter 12 va

iter 13 va

iter 14 va

iter 15 va

iter 16 va

iter 17 va

iter 18 va

iter 19 va

iter 20 va

iter 21 va

iter 22 va

iter 23 va

iter 24 va

iter 25 va

iter 26 va

iter 27 va

iter 27 va

iter 27 va

final valu

converged

initial va

iter 2 va

ata rtn

```
iter 3 value -2.4000000
iter 4 value -2.4000000
iter 5 value -2.4000000
iter 6 value -2.4000000
iter 7 value -2.4000000
iter 8 value -2.4000000
iter 9 value -2.4000000
iter 10 value -2.4000000
iter 11 value -2.4000000
iter 12 value -2.4000000
iter 13 value -2.4000000
iter 14 value -2.4000000
iter 15 value -2.4000000
iter 16 value -2.4000000
iter 17 value -2.4000000
iter 18 value -2.4000000
iter 19 value -2.4000000
iter 20 value -2.4000000
iter 21 value -2.4000000
iter 22 value -2.4000000
iter 22 value -2.4000000
iter 22 value -2.4000000
final value -2.4000000
converged
> BA_fit
$fit
```

Call:

```
stats::arima(x =
  Q), period =
  REPORT = 1, r
```

Coefficients:

ar1	ma1
-0.6105	-1.2710
s.e. 0.1785	0.18

sigma^2 estimated

\$degrees\_of\_freedom

```
[1] 205
```

\$table

	Estimate
ar1	-0.6105 0.17
ma1	-1.2710 0.18
ma2	-0.3784 0.32
ma3	0.6466 0.15
ma4	0.0325 0.14
ma5	-0.1563 0.13
ma6	0.2142 0.16
ma7	-0.1488 0.14
ma8	0.0621 0.07

\$AIC

```
[1] -3.914096
```

\$AICc

```
[1] -3.899868
```

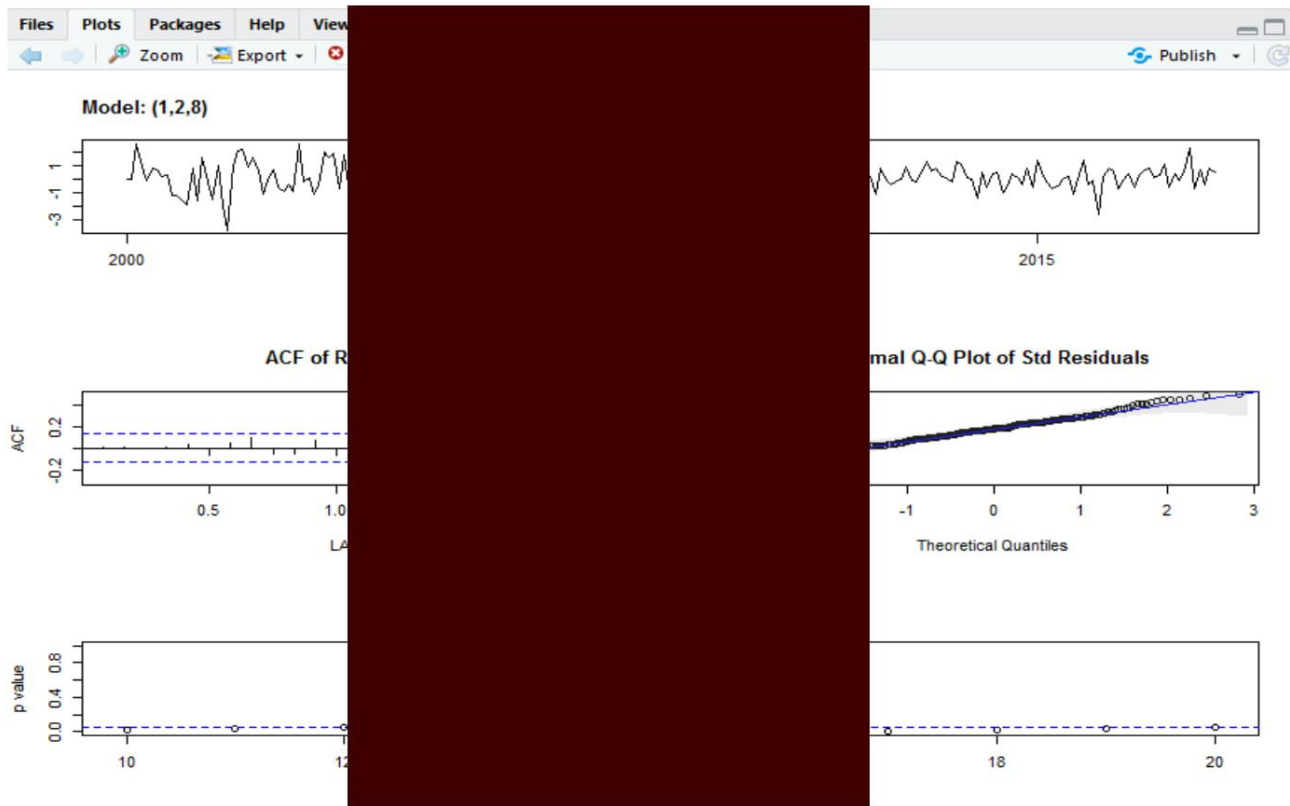
\$BIC

```
[1] -4.773459
```

```
al = list(order = c(P, D,
optim.control = list(trace = trc,
```

	ma6	ma7	ma8
	0.2142	-0.1488	0.0621
2	0.1673	0.1442	0.0790

```
22.93,aic = -425.86
```



The AICs and BICs from each run are summarized as below:

Model	
AR(1)	
AR(2)	
MA(3)	
MA(4)	
ARMA (2,3)	
ARMA(3,4)	
ARIMA(1,3,2)	
ARIMA (1,3,3)	
ARIMA(1,2,8)	

ARMA(2,3) has the lowest AIC and BIC values, indicating it is the best model for the data in this case.

Below is the summary of the diagnostic plots for the ARMA (2,3) model. The result shows ARMA (2,3) is a good candidate for the data. The residuals with some autocorrelations exist, Normal Q-Q plot shows the residuals are normally distributed. The autocorrelations of residuals are NOT significantly different from zero.

	AR(1)		AR(2)		MA(3)	
Standardized residuals	Some outliers exist		Some outliers exist		Lots of outliers exist	
ACF of residuals	Some spikes exist		Some spikes exist		Some spikes exist	
Normal Q-Q plot	Close to normal distribution		Close to normal distribution		Close to normal distribution	
p.values for Ljung-Box statistics	Autocorrelation residuals are different from 0		Autocorrelation residuals are different from 0		Autocorrelation residuals are different from 0	

	ARMA(1,1)		ARMA(2,2)		ARMA(3,4)	
Standardized residuals	Some outliers exist		Lots of outliers exist		Lots of outliers exist	
ACF of residuals	Some spikes exist		Some spikes exist		Some spikes exist	
Normal Q-Q plot	Close to normal distribution		Close to normal distribution		Close to normal distribution	
p.values for Ljung-Box statistics	Autocorrelation residuals are different from 0		Autocorrelation residuals are significantly different from 0		Autocorrelation residuals are different from 0	

	ARIMA(1,3,2)		ARIMA(1,2,2)		ARIMA(1,2,0)	
Standardized residuals	Some outliers exist		Lots of outliers exist		Lots of outliers exist	
ACF of residuals	Some spikes exist		Some spikes exist		Some spikes exist	
Normal Q-Q plot	Close to normal distribution		Close to normal distribution		Close to normal distribution	
p.values for Ljung-Box statistics	Autocorrelation residuals are different from 0		Autocorrelation residuals are significantly different from 0		Autocorrelation residuals are significantly different from 0	