

wqu-econometrics-group-6-A-w7

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Final Report

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3.3.1 Algorithmic Trading

Fenner (2012) says Design your own algorithmic trading strategy in R.

KBank SCB

pair trading by using Thai Stocks in Banking Sector, KBANK and SCB. You can search more details by using ticker KBANK.BK and SCB.BK in Yahoo Finance.

ANN model incorporate with GARCH model and Google trend input. References: 1. Volatility Forecast Based on the Hybrid Artificial Neural Network and GARCH-type Models (<https://www.sciencedirect.com/science/article/pii/S1877050916313382>) 2. GARCH based artificial neural networks in forecasting conditional variance of stock returns (<https://hrcak.srce.hr/file/197475>)

```
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
## Registered S3 method overwritten by 'xts':
##   method      from
##   as.zoo.xts zoo
## Loading required package: TTR
```

```

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

## Version 0.4-0 included new data defaults. See ?getSymbols.
getSymbols("SPY", scr="yahoo", from = as.Date("2007-01-04"), to = as.Date("2019-10-18"), warnings=FALSE)

## '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] "SPY"
SPY500<- SPY[, "SPY.Close"]
head(SPY500)

##           SPY.Close
## 2007-01-04    141.67
## 2007-01-05    140.54
## 2007-01-08    141.19
## 2007-01-09    141.07
## 2007-01-10    141.54
## 2007-01-11    142.16
#fill NA with previous non-NA value
library(zoo)
library(dplyr)

##
## Attaching package: 'dplyr'

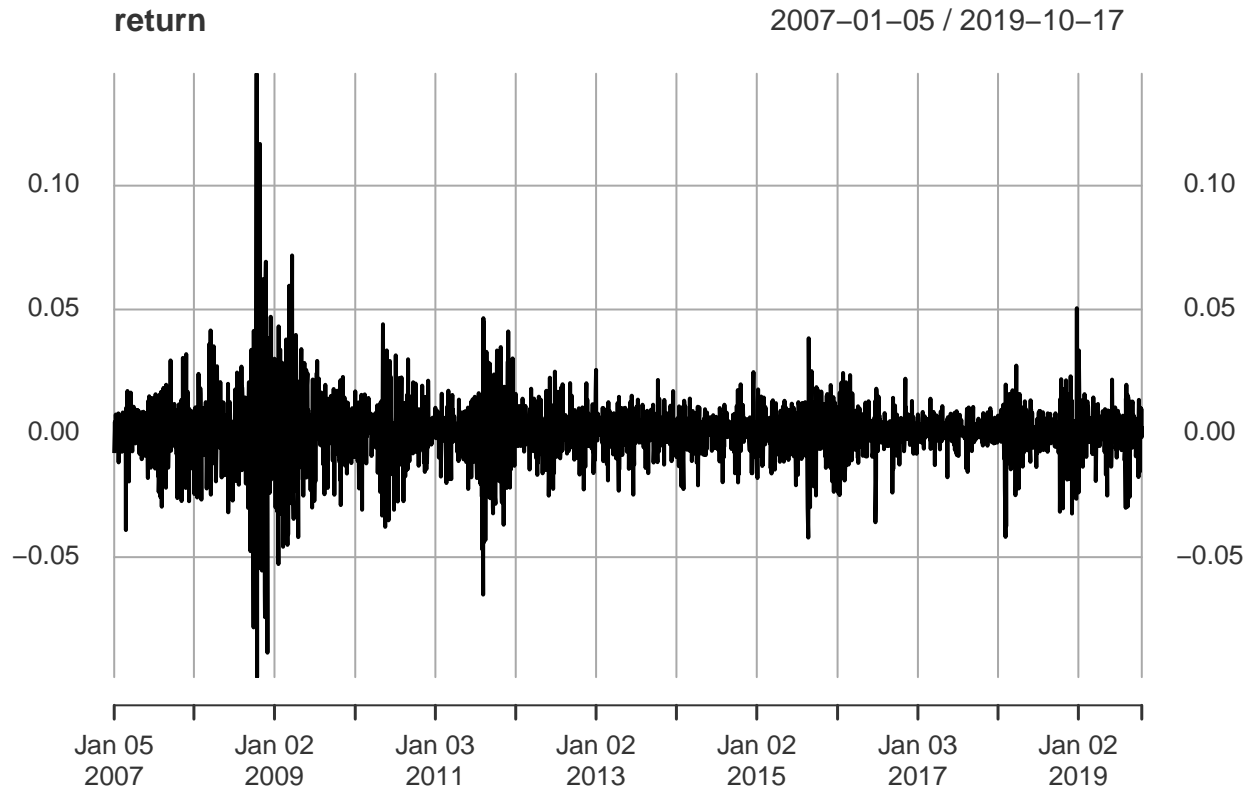
## The following objects are masked from 'package:xts':
##
##   first, last

## The following objects are masked from 'package:stats':
##
##   filter, lag

## The following objects are masked from 'package:base':
##
##   intersect, setdiff, setequal, union

SPY500 <- na.locf(SPY500)
return <- Delt(SPY500)
rows = nrow(return)
return <- return[2:rows]
plot(return)

```



```
#technical analysis indicators
average10<- rollapply(SPY500, 10, mean)
average20<-rollapply(SPY500, 20, mean)
std10<- rollapply(SPY500, 10, sd)
std20<- rollapply(SPY500, 20, sd)
rsi5<- RSI(SPY500,5,"SMA")
rsi14<- RSI(SPY500, 14, "SMA")
macd12269<- MACD(SPY500, 12, 26, 9, "SMA")
macd7205<- MACD(SPY500, 7, 20, 5, "SMA")
bollinger_bands<-BBands(SPY500,20,"SMA",2)
direction<- data.frame(matrix(NA,dim(SPY500)[1],1))
lagreturn<- (SPY500 - Lag(SPY500, 20))/Lag(SPY500, 20)
direction[lagreturn>0.02] <- "Up"
direction[lagreturn< -0.02] <- "Down"
direction[lagreturn< 0.02 &lagreturn> -0.02] <- "NoWhere"
```

```
#GARCH Model
require(stats)
require(tseries)
```

```
## Loading required package: tseries
require(forecast)
```

```
## Loading required package: forecast
## Registered S3 methods overwritten by 'forecast':
##   method          from
##   fitted.fracdiff  fracdiff
##   residuals.fracdiff fracdiff
```

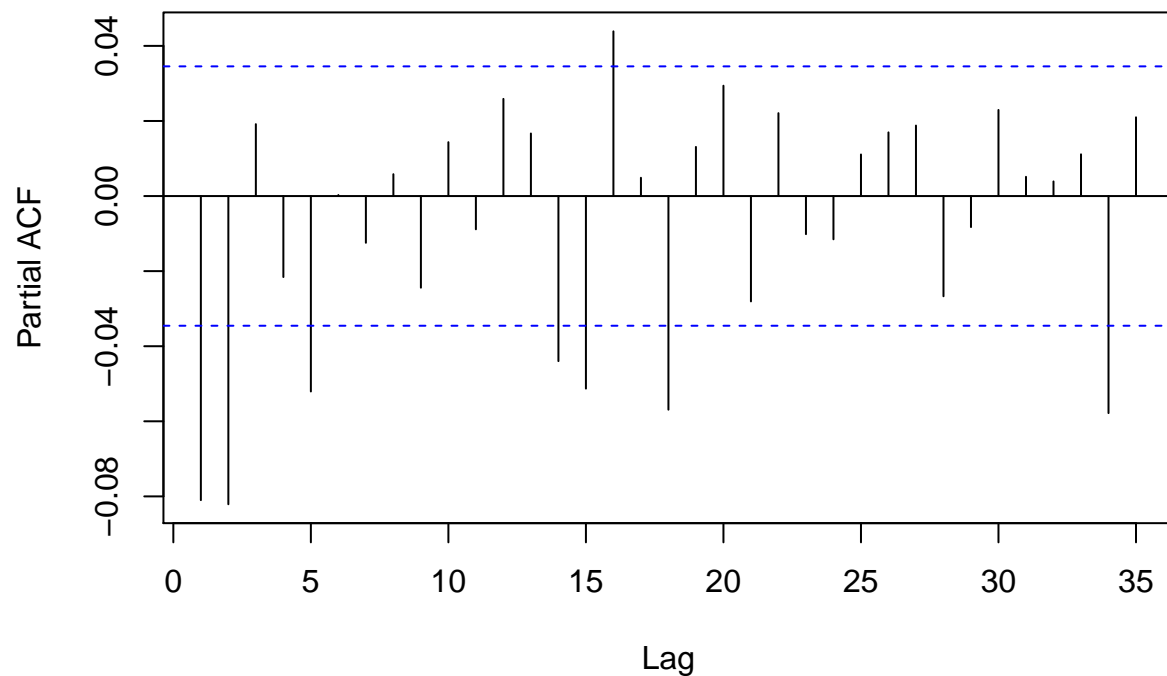
```
#adf test suggesting stationarity  
adf.test(return)
```

```
## Warning in adf.test(return): p-value smaller than printed p-value
```

```
##  
## Augmented Dickey-Fuller Test  
##  
## data: return  
## Dickey-Fuller = -16.192, Lag order = 14, p-value = 0.01  
## alternative hypothesis: stationary
```

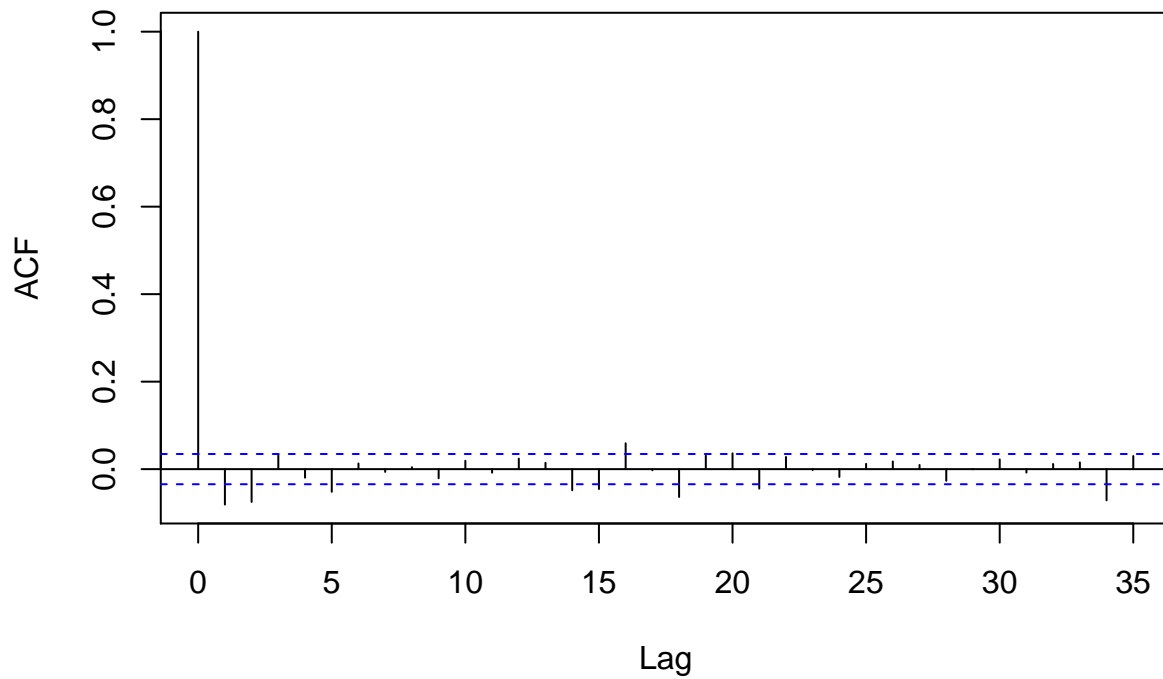
```
#PACF plot suggests significant spike through lag 2.  
pacf(return)
```

Series return



```
#ACF plot shows exponential decay. Thus, it can be deduced AR(2) model.  
acf(return)
```

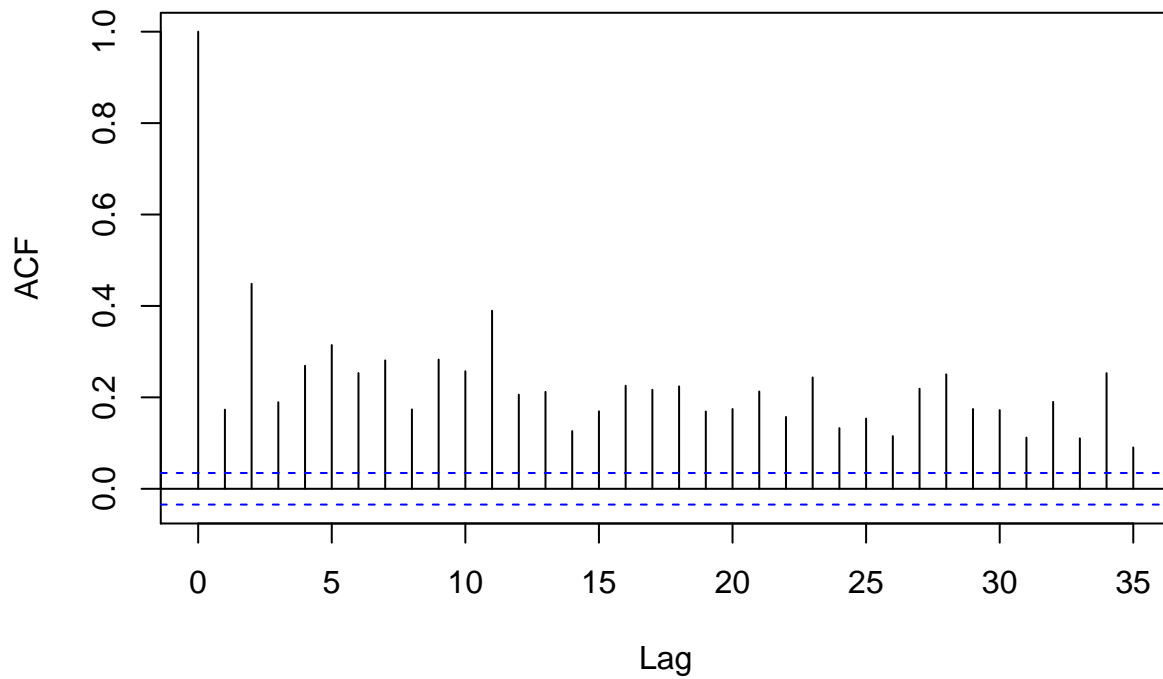
Series return



```
lengthOfReturns<-length(return)
timeseries <- ts(return)
```

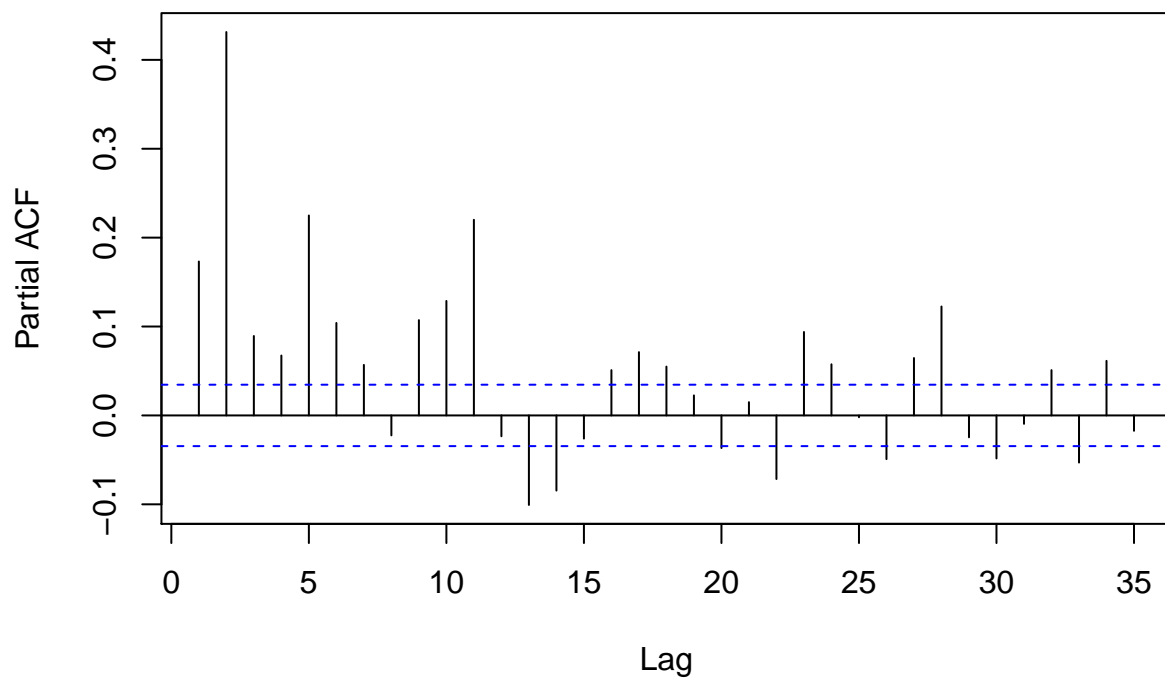
```
ARIMA_Model <- arima(window(timeseries,1,lengthOfReturns), order=c(2,0,0), method = "ML")
acf((ARIMA_Model$residuals)^2)
```

Series (ARIMA_Model\$residuals)^2



```
pacf((ARIMA_Model$residuals)^2)
```

Series (ARIMA_Model\$residuals)^2



```
require(fGarch)
```

```
## Loading required package: fGarch
```

```

## Loading required package: timeDate
## Loading required package: timeSeries
##
## Attaching package: 'timeSeries'
## The following object is masked from 'package:zoo':
##
##     time<-
## Loading required package: fBasics
##
## Attaching package: 'fBasics'
## The following object is masked from 'package:TTR':
##
##     volatility
model <- garchFit(formula = ~ arma(2,0) + garch(11,0) , data = timeseries, trace = F)
summary(model)

##
## Title:
## GARCH Modelling
##
## Call:
## garchFit(formula = ~arma(2, 0) + garch(11, 0), data = timeseries,
##     trace = F)
##
## Mean and Variance Equation:
## data ~ arma(2, 0) + garch(11, 0)
## <environment: 0xbdd4a40>
## [data = timeseries]
##
## Conditional Distribution:
## norm
##
## Coefficient(s):
##      mu      ar1      ar2      omega      alpha1
## 8.1551e-04 -6.5797e-02 -2.4323e-02 1.8515e-05 7.6807e-02
##      alpha2      alpha3      alpha4      alpha5      alpha6
## 1.5374e-01 9.8210e-02 1.3536e-01 6.3122e-02 5.9403e-02
##      alpha7      alpha8      alpha9      alpha10      alpha11
## 5.4562e-02 5.9989e-02 5.9964e-02 6.7037e-02 4.2801e-02
##
## Std. Errors:
## based on Hessian
##
## Error Analysis:
##      Estimate Std. Error t value Pr(>|t|)
## mu      8.155e-04 1.360e-04 5.997 2.01e-09 ***
## ar1     -6.580e-02 1.836e-02 -3.583 0.000339 ***
## ar2     -2.432e-02 1.936e-02 -1.256 0.209051
## omega    1.852e-05 1.747e-06 10.595 < 2e-16 ***
## alpha1   7.681e-02 1.794e-02 4.280 1.87e-05 ***
## alpha2   1.537e-01 2.475e-02 6.212 5.22e-10 ***

```

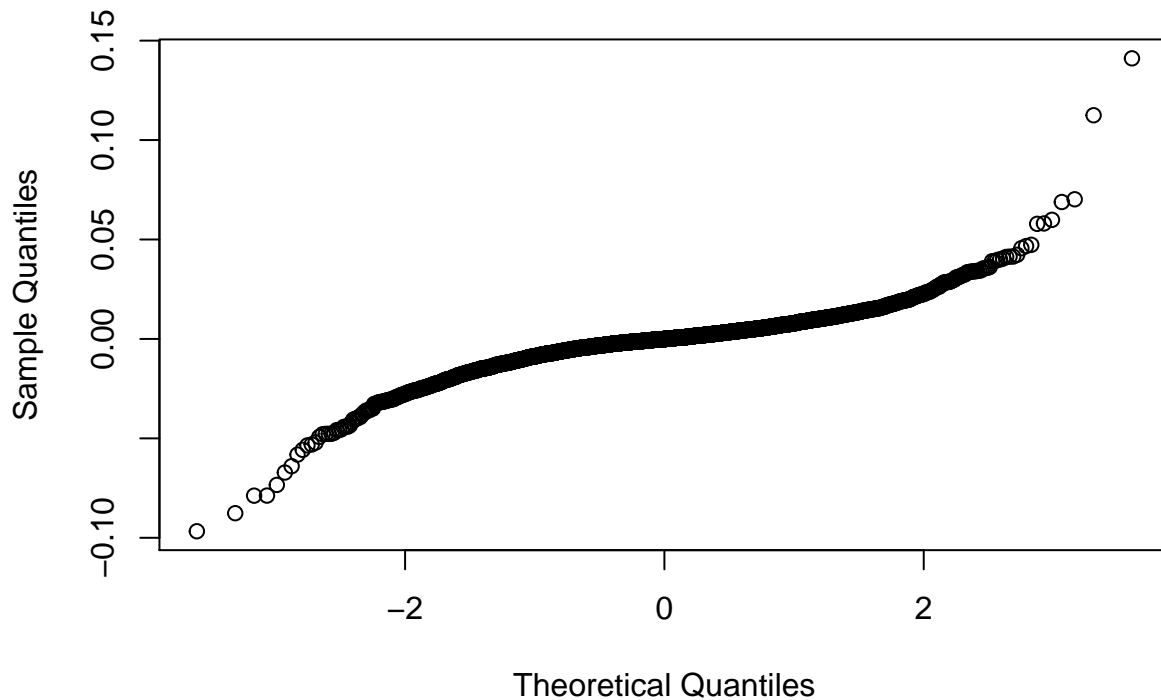
```

## alpha3 9.821e-02 2.234e-02 4.396 1.11e-05 ***
## alpha4 1.354e-01 2.603e-02 5.200 2.00e-07 ***
## alpha5 6.312e-02 1.846e-02 3.419 0.000628 ***
## alpha6 5.940e-02 1.861e-02 3.191 0.001416 **
## alpha7 5.456e-02 1.725e-02 3.162 0.001565 **
## alpha8 5.999e-02 1.885e-02 3.183 0.001460 **
## alpha9 5.996e-02 2.092e-02 2.867 0.004149 **
## alpha10 6.704e-02 1.907e-02 3.516 0.000438 ***
## alpha11 4.280e-02 1.716e-02 2.494 0.012641 *
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Log Likelihood:
## 10521.29 normalized: 3.268496
##
## Description:
## Tue Oct 22 10:11:35 2019 by user:
##
##
## Standardised Residuals Tests:
##
## Statistic p-Value
## Jarque-Bera Test R Chi^2 766.633 0
## Shapiro-Wilk Test R W 0.9749061 0
## Ljung-Box Test R Q(10) 12.21964 0.2706257
## Ljung-Box Test R Q(15) 23.61286 0.07196621
## Ljung-Box Test R Q(20) 27.28147 0.1275291
## Ljung-Box Test R^2 Q(10) 3.15208 0.977615
## Ljung-Box Test R^2 Q(15) 8.047753 0.9218539
## Ljung-Box Test R^2 Q(20) 10.68428 0.9540028
## LM Arch Test R TR^2 5.127157 0.9535963
##
## Information Criterion Statistics:
## AIC BIC SIC HQIC
## -6.527673 -6.499356 -6.527716 -6.517524
res = residuals(model)

#qq-plot of residual between GARCH model and actual data
qqnorm(res)

```


Normal Q-Q Plot



```
library("rugarch")
```

```
## Loading required package: parallel
```

```
##
```

```
## Attaching package: 'rugarch'
```

```
## The following object is masked from 'package:stats':
```

```
##
```

```
##      sigma
```

```
garch11_spec <- ugarchspec(variance.model = list(garchOrder = c(11, 0)), mean.model = list(armaOrder = c(2, 0)),
garch11_fit<-ugarchfit(spec=garch11_spec, data=timeseries)
```

```
## Warning in .sgarchfit(spec = spec, data = data, out.sample = out.sample, :
```

```
## ugarchfit-->warning: solver failed to converge.
```

```
garch11_fit
```

```
##
```

```
## *-----*
```

```
## *          GARCH Model Fit          *
```

```
## *-----*
```

```
##
```

```
## Conditional Variance Dynamics
```

```
## -----
```

```
## GARCH Model   : sGARCH(11,0)
```

```
## Mean Model    : ARFIMA(2,0,0)
```

```
## Distribution   : norm
```

```
##
```

```
## Convergence Problem:
```

```
## Solver Message:
```

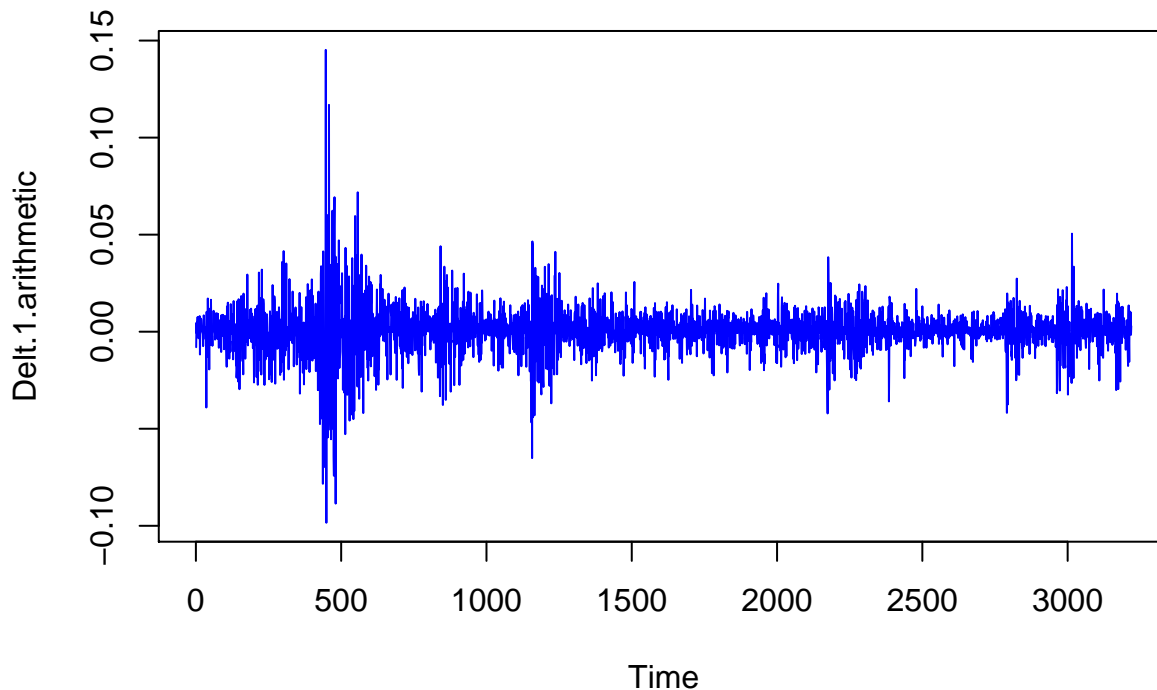
```
garch11_fit@fit$fitted.values
```

```
## NULL
```

```
#visualize how well GARCH itself fit the data
```

```
plot(timeseries, type="l", col="blue")
```

```
lines(garch11_fit@fit$fitted.values, col="green")
```



```
#binding closing price and technical analysis indicators into a variable SPY500
```

```
SPY500 <- cbind(SPY500[2:nrow(SPY500)], average10[2:nrow(average10)], average20[2:nrow(average20)], std
```

```
#integrate GARCH model rolling window prediction output into variable
```

```
SPY500 <- cbind(SPY500,garch11_fit@fit$fitted.values)
```

```
#Import Google trend data regarding trend of recession and expansion
```

```
recessiondata<-read.csv("Recession_gtrends.csv",header=F)$V2
```

```
expansiondata<-read.csv("Expansion_gtrends.csv",header=F)$V2
```

```
#integrate Google trend data into variable
```

```
SPY500 <- cbind(SPY500,recessiondata,expansiondata)
```

```
#indicate end and start dates for train, validating and testing period
```

```
train_sdate<- "2007-03-01"
```

```
train_edate<- "2017-03-01"
```

```
vali_sdate<- "2017-03-02"
```

```
vali_edate<- "2018-03-02"
```

```
test_sdate<- "2018-03-03"
```

```
test_edate<- "2019-10-18"
```

```
#constructing data ranges for the three datasets
```

```
trainrow<- which(index(SPY500) >= train_sdate& index(SPY500) <= train_edate)
```

```
valirow<- which(index(SPY500) >= vali_sdate& index(SPY500) <= vali_edate)
```

```
testrow<- which(index(SPY500) >= test_sdate& index(SPY500) <= test_edate)
```

```

#extract data fpr training, validating and testing periods
train<- SPY500[trainrow,]
vali<- SPY500[valirow,]
test<- SPY500[testrow,]
trainme<-apply(train,2,mean)
trainstd<-apply(train,2,sd)

#training, validating and testing data dimensions
trainidn<- (matrix(1,dim(train)[1],dim(train)[2]))
valiidn<- (matrix(1,dim(vali)[1],dim(vali)[2]))
testidn<- (matrix(1,dim(test)[1],dim(test)[2]))

#normalize the three datasets
norm_train<- (train-t(trainme*t(trainidn)))/t(trainstd*t(trainidn))
norm_vali<- (vali-t(trainme*t(valiidn)))/t(trainstd*t(valiidn))
norm_test<- (test-t(trainme*t(testidn)))/t(trainstd*t(testidn))

#define training, validating and testing period
traindir<- direction[trainrow,1]
validir<- direction[valirow,1]
testdir<- direction[testrow,1]

#implement ANN
library(nnet)
set.seed(1)
neural_network<- nnet(norm_train, class.ind(traindir), size=4, trace=T)

## # weights: 87
## initial value 2046.443499
## iter 10 value 856.398611
## iter 20 value 666.873545
## iter 30 value 630.913796
## iter 40 value 599.414176
## iter 50 value 562.914859
## iter 60 value 545.981441
## iter 70 value 538.057476
## iter 80 value 534.228500
## iter 90 value 529.830311
## iter 100 value 527.163171
## final value 527.163171
## stopped after 100 iterations

#obtain data dimension
dim(norm_train)

## [1] 2519 17

#make prediction
vali_pred<-predict(neural_network, norm_vali)
head(vali_pred)

##           Down      NoWhere      Up
## 2017-03-02 0.01051211 0.001628957 0.9978049
## 2017-03-03 0.01007747 0.001287799 0.9982364
## 2017-03-06 0.01317520 0.005836338 0.9928284

```

```

## 2017-03-07 0.01391204 0.007970571 0.9904222
## 2017-03-08 0.02628690 0.143936354 0.8426444
## 2017-03-09 0.02283733 0.081539201 0.9038932

#calculate the predicted direction using the information obtained above
vali_pred_class<- data.frame(matrix(NA,dim(vali_pred)[1],1))
vali_pred_class[vali_pred[, "Down"] > 0.5,1]<- "Down"
vali_pred_class[vali_pred[, "NoWhere"] > 0.5,1]<- "NoWhere"
vali_pred_class[vali_pred[, "Up"] > 0.5,1]<- "Up"
vali_pred_class[is.na(vali_pred_class)]<- "NoWhere"

#check forecast accuracy
library(caret)

## Loading required package: lattice

## Loading required package: ggplot2

u<- union(vali_pred_class[,1],validir)
t<-table(factor(vali_pred_class[,1],u),factor(validir,u))
confusionMatrix(t)

## Confusion Matrix and Statistics
##
##
##           Up NoWhere Down
## Up         67      25    0
## NoWhere    21     120    6
## Down        0       1   13
##
## Overall Statistics
##
##               Accuracy : 0.7905
##               95% CI : (0.7351, 0.839)
##       No Information Rate : 0.5771
##       P-Value [Acc > NIR] : 6.531e-13
##
##               Kappa : 0.6078
##
##  Mcnemar's Test P-Value : NA
##
## Statistics by Class:
##
##               Class: Up Class: NoWhere Class: Down
## Sensitivity          0.7614          0.8219    0.68421
## Specificity          0.8485          0.7477    0.99573
## Pos Pred Value       0.7283          0.8163    0.92857
## Neg Pred Value       0.8696          0.7547    0.97490
## Prevalence           0.3478          0.5771    0.07510
## Detection Rate       0.2648          0.4743    0.05138
## Detection Prevalence 0.3636          0.5810    0.05534
## Balanced Accuracy    0.8049          0.7848    0.83997

#check accuracy on testing data
test_pred<- predict(neural_network, norm_test)
test_pred

```

##		Down	NoWhere	Up
##	2018-03-05	0.183357235	6.564762e-01	0.0052489332
##	2018-03-06	0.050737861	9.135963e-01	0.0084945613
##	2018-03-07	0.034055294	9.370218e-01	0.0116446949
##	2018-03-08	0.017907481	8.331590e-01	0.1635788293
##	2018-03-09	0.010497377	9.714202e-04	0.9984643629
##	2018-03-12	0.004814221	1.913241e-05	0.9999643674
##	2018-03-13	0.007094653	1.096155e-04	0.9998056459
##	2018-03-14	0.013208037	4.370332e-03	0.9941480205
##	2018-03-15	0.018583365	2.997261e-02	0.9643509099
##	2018-03-16	0.043126981	3.707012e-01	0.5323167980
##	2018-03-19	0.030471852	8.664017e-01	0.0794152874
##	2018-03-20	0.027339338	9.448381e-01	0.0129365274
##	2018-03-21	0.775905848	3.592525e-03	0.0299051214
##	2018-03-22	0.976331756	4.492896e-04	0.0058513045
##	2018-03-23	0.992088965	1.669573e-04	0.0029006852
##	2018-03-26	0.930592766	1.216710e-03	0.0118878810
##	2018-03-27	0.991016491	1.870533e-04	0.0031465735
##	2018-03-28	0.994934275	1.115180e-04	0.0021906646
##	2018-03-29	0.987873139	2.447817e-04	0.0038133562
##	2018-04-02	0.995916750	9.187234e-05	0.0019114504
##	2018-04-03	0.995016983	1.099094e-04	0.0021675548
##	2018-04-04	0.919166154	1.435839e-03	0.0130055084
##	2018-04-05	0.624168024	7.949893e-03	0.0425586707
##	2018-04-06	0.979235022	4.021632e-04	0.0053466953
##	2018-04-09	0.992237551	2.140369e-04	0.0022326584
##	2018-04-10	0.846517896	1.783749e-01	0.0003380597
##	2018-04-11	0.750428031	2.793809e-01	0.0004789883
##	2018-04-12	0.828969383	2.006875e-01	0.0003562084
##	2018-04-13	0.709004675	3.158755e-01	0.0005517534
##	2018-04-16	0.632278710	3.772265e-01	0.0007071621
##	2018-04-17	0.270187078	6.756941e-01	0.0020534452
##	2018-04-18	0.022502534	9.171161e-01	0.0360721015
##	2018-04-19	0.064775126	5.552919e-01	0.2317686014
##	2018-04-20	0.037559906	8.793059e-01	0.0441559492
##	2018-04-23	0.038678063	9.074189e-01	0.0206972980
##	2018-04-24	0.116735450	6.604302e-01	0.0441357167
##	2018-04-25	0.100502192	5.758906e-01	0.1228459513
##	2018-04-26	0.028337540	8.481131e-01	0.0803418326
##	2018-04-27	0.035265637	6.196896e-01	0.2858593934
##	2018-04-30	0.036303010	7.197210e-01	0.1856448372
##	2018-05-01	0.008584614	8.621195e-01	0.1864802194
##	2018-05-02	0.009462422	9.211766e-01	0.1189477079
##	2018-05-03	0.009435433	9.733644e-01	0.0298672304
##	2018-05-04	0.005368766	8.862977e-01	0.1510582881
##	2018-05-07	0.003560896	6.014242e-01	0.4885796402
##	2018-05-08	0.006535963	8.440220e-01	0.1465852464
##	2018-05-09	0.003949110	4.077621e-01	0.5813350354
##	2018-05-10	0.002157956	3.605941e-02	0.9680018218
##	2018-05-11	0.006533204	5.459693e-04	0.9991677636
##	2018-05-14	0.006458653	6.740972e-03	0.9906979971
##	2018-05-15	0.026256644	8.204917e-02	0.7574333526
##	2018-05-16	0.023091888	1.169555e-01	0.8478507323
##	2018-05-17	0.036382909	3.525368e-01	0.5255604257

```

## 2018-05-18 0.035543242 6.654083e-01 0.2077066380
## 2018-05-21 0.028741874 1.959807e-01 0.7448400535
## 2018-05-22 0.014664298 5.487638e-01 0.4657046299
## 2018-05-23 0.015077632 3.280769e-01 0.6770242393
## 2018-05-24 0.022276041 9.205770e-02 0.9001257233
## 2018-05-25 0.020936823 7.195009e-02 0.9240678218
## 2018-05-29 0.037657823 5.575861e-01 0.4215320050
## 2018-05-30 0.018559052 3.595300e-02 0.9577900399
## 2018-05-31 0.021792064 2.426545e-01 0.7388608444
## 2018-06-01 0.010344592 6.096450e-03 0.9911535563
## 2018-06-04 0.008890774 1.411440e-03 0.9976000570
## 2018-06-05 0.004761002 6.271996e-04 0.9990783524
## 2018-06-06 0.005693197 3.460117e-05 0.9999250862
## 2018-06-07 0.004538612 1.300833e-05 0.9999731439
## 2018-06-08 0.004831588 2.051284e-05 0.9999589773
## 2018-06-11 0.005459971 3.553826e-05 0.9999316425
## 2018-06-12 0.007206131 3.083546e-04 0.9994876834
## 2018-06-13 0.016140788 2.661029e-02 0.9619073179
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## 2019-08-13 0.224899537 3.368145e-02 0.1364025865
## 2019-08-14 0.559596421 1.024700e-02 0.0494694376
## 2019-08-15 0.789323514 3.902842e-03 0.0257003879
## 2019-08-16 0.527725130 1.141847e-02 0.0536111432
## 2019-08-19 0.672967214 6.568683e-03 0.0374119859
## 2019-08-20 0.945930194 9.768483e-04 0.0099459477
## 2019-08-21 0.367187844 1.827738e-02 0.0970703573
## 2019-08-22 0.348546562 1.514060e-02 0.2028474305
## 2019-08-23 0.721122669 5.036574e-03 0.0381332607
## 2019-08-26 0.839445368 1.690649e-03 0.0673879042
## 2019-08-27 0.675011008 2.595765e-02 0.0133683098
## 2019-08-28 0.651690832 1.687519e-01 0.0072067833
## 2019-08-29 0.779030723 6.108464e-02 0.0160102035
## 2019-08-30 0.238488876 2.953715e-01 0.1104328095

```

```
## 2019-09-03 0.054192580 8.989306e-01 0.0089461187
## 2019-09-04 0.025541597 6.314917e-01 0.2538999055
## 2019-09-05 0.017191394 8.472539e-03 0.9868490353
## 2019-09-06 0.161718946 1.517631e-06 0.9997819085
## 2019-09-09 0.167623330 5.481642e-06 0.9994070995
## 2019-09-10 0.176204974 5.620111e-06 0.9993576495
## 2019-09-11 0.045030433 1.120440e-05 0.9997648151
## 2019-09-12 0.009151266 5.370529e-05 0.9998522552
## 2019-09-13 0.129465428 4.099745e-07 0.9999376069
## 2019-09-16 0.177631904 3.541542e-06 0.9995449775
## 2019-09-17 0.016362789 2.858733e-03 0.9947905946
## 2019-09-18 0.015671706 1.607602e-02 0.9818740656
## 2019-09-19 0.015372496 1.574124e-02 0.9825713184
## 2019-09-20 0.028763619 3.439918e-01 0.6835090194
## 2019-09-23 0.031737561 3.524058e-01 0.6484671717
## 2019-09-24 0.045011699 5.517544e-01 0.3959062842
## 2019-09-25 0.027261353 1.495237e-01 0.8327506807
## 2019-09-26 0.027099256 1.326447e-01 0.8462088841
## 2019-09-27 0.031066937 2.674601e-01 0.7079455127
## 2019-09-30 0.023023934 1.443282e-01 0.8323979128
## 2019-10-01 0.009771510 9.064495e-01 0.1327475529
## 2019-10-02 0.159906866 6.919338e-02 0.1168137130
## 2019-10-03 0.174950529 6.057580e-02 0.1108302166
## 2019-10-04 0.021172344 9.320244e-01 0.0207789091
## 2019-10-07 0.040221857 9.347331e-01 0.0082300887
## 2019-10-08 0.792218563 3.836010e-03 0.0253200731
## 2019-10-09 0.147549934 8.198938e-01 0.0031121667
## 2019-10-10 0.109337868 8.591332e-01 0.0038659475
## 2019-10-11 0.040977380 9.255918e-01 0.0088102731
## 2019-10-14 0.024211640 9.513037e-01 0.0126735398
## 2019-10-15 0.009145773 9.280675e-01 0.0565279989
## 2019-10-16 0.029474955 9.181412e-01 0.0145177479
## 2019-10-17 0.021814083 9.205676e-01 0.0214635023
```

```
#indicate the classes for the testing data
test_pred_class<- data.frame(matrix(NA,dim(test_pred)[1],1))
test_pred_class[test_pred[, "Down"] > 0.5,1]<- "Down"
test_pred_class[test_pred[, "NoWhere"] > 0.5,1]<- "NoWhere"
test_pred_class[test_pred[, "Up"] > 0.5,1]<- "Up"
test_pred_class[is.na(test_pred_class)]<- "NoWhere"
```

```
#Check the accuracy of the forecasts
u<- union(test_pred_class[,1],testdir)
t<-table(factor(test_pred_class[,1],u),factor(testdir,u))
confusionMatrix(t)
```

```
## Confusion Matrix and Statistics
```

```
##
```

```
##
```

```
##           NoWhere  Up  Down
## NoWhere           93  18   16
## Up                36 147    2
## Down              16   0   82
```

```
##
```

```
## Overall Statistics
```

```

##
##          Accuracy : 0.7854
##          95% CI : (0.7424, 0.8241)
##    No Information Rate : 0.4024
##    P-Value [Acc > NIR] : < 2e-16
##
##          Kappa : 0.6701
##
## Mcnemar's Test P-Value : 0.04601
##
## Statistics by Class:
##
##          Class: NoWhere Class: Up Class: Down
## Sensitivity          0.6414    0.8909    0.8200
## Specificity          0.8717    0.8449    0.9484
## Pos Pred Value       0.7323    0.7946    0.8367
## Neg Pred Value       0.8163    0.9200    0.9423
## Prevalence           0.3537    0.4024    0.2439
## Detection Rate       0.2268    0.3585    0.2000
## Detection Prevalence 0.3098    0.4512    0.2390
## Balanced Accuracy     0.7565    0.8679    0.8842

```

```

#generate trade signals using the same pattern as human psychology
signal<-ifelse(test_pred_class=="Up",1,ifelse(test_pred_class=="Down",-1, 0))
signal

```

```

##          matrix.NA..dim.test_pred..1...1.
##    [1,]                                0
##    [2,]                                0
##    [3,]                                0
##    [4,]                                0
##    [5,]                                1
##    [6,]                                1
##    [7,]                                1
##    [8,]                                1
##    [9,]                                1
##   [10,]                                1
##   [11,]                                0
##   [12,]                                0
##   [13,]                               -1
##   [14,]                               -1
##   [15,]                               -1
##   [16,]                               -1
##   [17,]                               -1
##   [18,]                               -1
##   [19,]                               -1
##   [20,]                               -1
##   [21,]                               -1
##   [22,]                               -1
##   [23,]                               -1
##   [24,]                               -1
##   [25,]                               -1
##   [26,]                               -1
##   [27,]                               -1
##   [28,]                               -1

```

##	[29,]	-1
##	[30,]	-1
##	[31,]	0
##	[32,]	0
##	[33,]	0
##	[34,]	0
##	[35,]	0
##	[36,]	0
##	[37,]	0
##	[38,]	0
##	[39,]	0
##	[40,]	0
##	[41,]	0
##	[42,]	0
##	[43,]	0
##	[44,]	0
##	[45,]	0
##	[46,]	0
##	[47,]	1
##	[48,]	1
##	[49,]	1
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##	[69,]	1
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##	[71,]	1
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##	[73,]	0
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## [106,]	1
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## [154,]	-1
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## [168,]	-1
## [169,]	-1
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## [201,]	-1
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## [398,]	0
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## [401,]	0
## [402,]	0
## [403,]	-1
## [404,]	0
## [405,]	0
## [406,]	0

```

## [407,] 0
## [408,] 0
## [409,] 0
## [410,] 0

test_return_SPY<- return[(index(return)>= test_sdate & index(return)<= test_edate), ]
test_return<- test_return_SPY*(signal)

library(PerformanceAnalytics)

##
## Attaching package: 'PerformanceAnalytics'
## The following objects are masked from 'package:timeDate':
##
##      kurtosis, skewness
## The following object is masked from 'package:graphics':
##
##      legend

#calculate cummulative return
cumm_return<- Return.cumulative(test_return)
cumm_return

##          Delt.1.arithmetic
## Cumulative Return      0.5698866

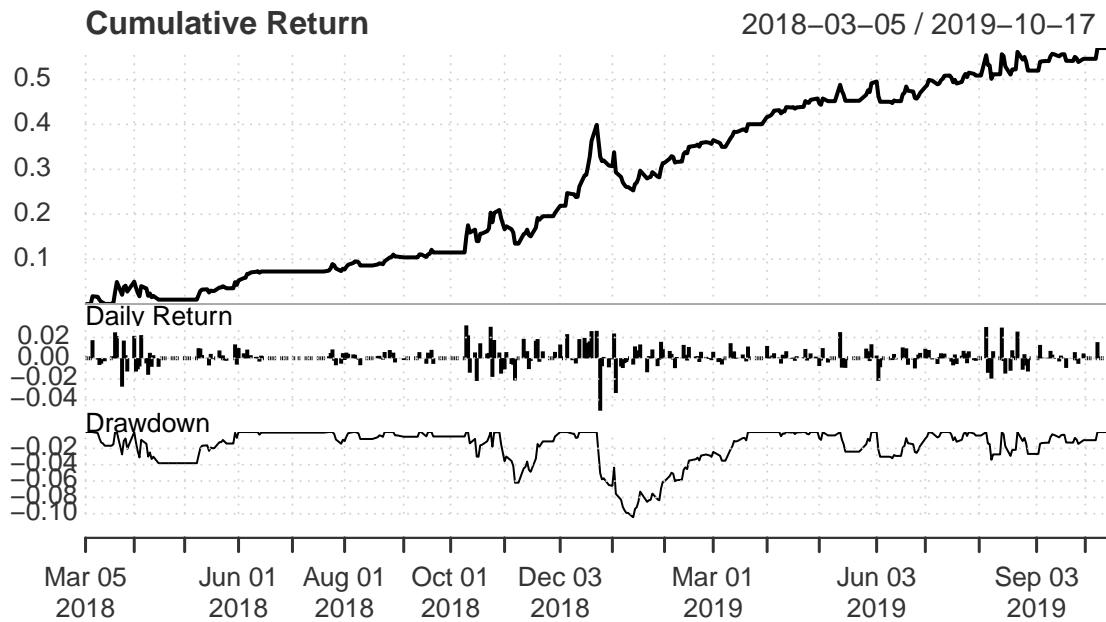
#calculate annual return
annual_return<- Return.annualized(test_return)
annual_return

##          Delt.1.arithmetic
## Annualized Return      0.319433

charts.PerformanceSummary(test_return)

```

Delt.1.arithmetic Performance



```
VaR(test_return, p=0.95)
```

```
##      Delt.1.arithmetic  
## VaR      -0.01182357
```

```
SharpeRatio(as.ts(test_return), Rf = 0, p=0.95, FUN = "StdDev")
```

```
##                                     [,1]  
## StdDev Sharpe (Rf=0%, p=95%): 0.1343047
```

```
SharpeRatio.annualized(test_return, Rf=0)
```

```
##                                     Delt.1.arithmetic  
## Annualized Sharpe Ratio (Rf=0%)      2.37838
```

References

Fenner, Martin. 2012. "One-Click Science Marketing." *Nature Materials* 11 (4). Nature Publishing Group: 261–63. <https://doi.org/10.1038/nmat3283>.