

SVM S&P500 Monthly 01-01-2012 to 06-30-2018

[Code ▾](#)[Hide](#)

```
#install.packages("quantmod")
#install.packages("e1071")
library(quantmod)
```

Loading required package: xts
Loading required package: zoo

Attaching package: <U+393C><U+3E31>zoo<U+393C><U+3E32>

The following objects are masked from <U+393C><U+3E31>package:base<U+393C><U+3E32>:

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>

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```
library(e1071)
# Importing the dataset
startDate = as.Date("2011-01-01")
endDate = as.Date("2018-06-30")
getSymbols("^GSPC",src="yahoo",from=startDate,to=endDate)
```

<U+393C><U+3E31>getSymbols<U+393C><U+3E32> currently uses auto.assign=TRUE by default, but will use auto.assign=FALSE in 0.5-0. You will still be able to use <U+393C><U+3E31>loadSymbols<U+393C><U+3E32> 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.

WARNING: There have been significant changes to Yahoo Finance data.
Please see the Warning section of <U+393C><U+3E31>?getSymbols.yahoo<U+393C><U+3E32> for details.

This message is shown once per session and may be disabled by setting options("getSymbols.yahoo.warning"=FALSE).

```
[1] "GSPC"
```

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```
dataset=data.frame(to.monthly(GSPC))  
dim(dataset)
```

```
[1] 90  6
```

[Hide](#)

```
head(dataset)
```

	GSPC.Open <dbl>	GSPC.High <dbl>	GSPC.L... <dbl>	GSPC.Close <dbl>	GSPC.Volume <dbl>	GSPC.Adjusted <dbl>
Jan 2011	1257.62	1302.67	1257.62	1286.12	92164940000	1286.12
Feb 2011	1289.14	1344.07	1289.14	1327.22	59223660000	1327.22
Mar 2011	1328.64	1332.28	1249.05	1325.83	89507640000	1325.83
Apr 2011	1329.48	1364.56	1294.70	1363.61	77364810000	1363.61
May 2011	1365.21	1370.58	1311.80	1345.20	81708980000	1345.20
Jun 2011	1345.20	1345.20	1258.07	1320.64	86122730000	1320.64
6 rows						

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```
tail(dataset)
```

	GSPC.Open <dbl>	GSPC.High <dbl>	GSPC.L... <dbl>	GSPC.Close <dbl>	GSPC.Volume <dbl>	GSPC.Adjusted <dbl>
Jan 2018	2683.73	2872.87	2682.36	2823.81	76860120000	2823.81
Feb 2018	2816.45	2835.96	2532.69	2713.83	79579410000	2713.83
Mar 2018	2715.22	2801.90	2585.89	2640.87	76369800000	2640.87
Apr 2018	2633.45	2717.49	2553.80	2648.05	69648590000	2648.05
May 2018	2642.96	2742.24	2594.62	2705.27	75617280000	2705.27
Jun 2018	2718.70	2791.47	2691.99	2718.37	77439710000	2718.37
6 rows						

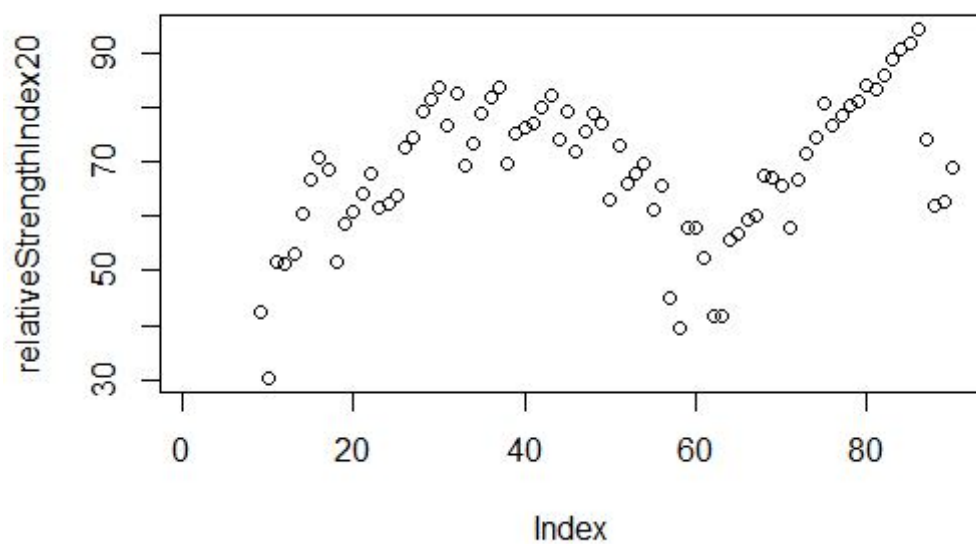
[Hide](#)

```
#RSI indicator
relativeStrengthIndex20=RSI(Op(dataset),n=8)
summary(relativeStrengthIndex20)
```

Min.	1st Qu.	Median	Mean	3rd Qu.	Max.	NA's
30.38	60.81	69.47	68.64	78.78	94.41	8

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```
plot(relativeStrengthIndex20)
```



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```
tail(relativeStrengthIndex20)
```

```
[1] 91.86270 94.41305 74.15401 61.89350 62.71293 68.81637
```

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```
dim(relativeStrengthIndex20)
```

```
NULL
```

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```
ncol(relativeStrengthIndex20)
```

```
NULL
```

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```
nrow(relativeStrengthIndex20)
```

```
NULL
```

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```
# Exponential Moving Average Indicator  
exponentialMovingAverage20=EMA(Op(dataset),n=8)  
head(exponentialMovingAverage20)
```

```
[1] NA NA NA NA NA NA
```

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```
tail(exponentialMovingAverage20)
```

```
[1] 2540.936 2602.161 2627.285 2628.655 2631.834 2651.138
```

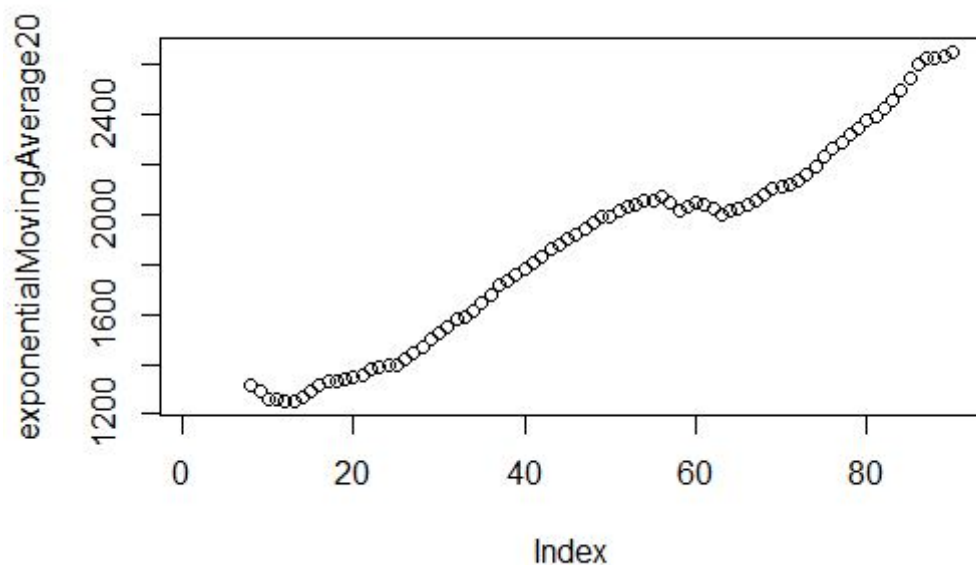
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```
summary(exponentialMovingAverage20)
```

Min.	1st Qu.	Median	Mean	3rd Qu.	Max.	NA's
1254	1485	1989	1875	2107	2651	7

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```
plot(exponentialMovingAverage20)
```

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```
# Difference in Exponential Moving Average
exponentialMovingAverageDiff = (Op(dataset) - exponentialMovingAverage20)
head(exponentialMovingAverageDiff)
```

```
[1] NA NA NA NA NA NA
```

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```
tail(exponentialMovingAverageDiff)
```

```
[1] 142.794117 214.288735 87.934588 4.794664 11.125857
[6] 67.562326
```

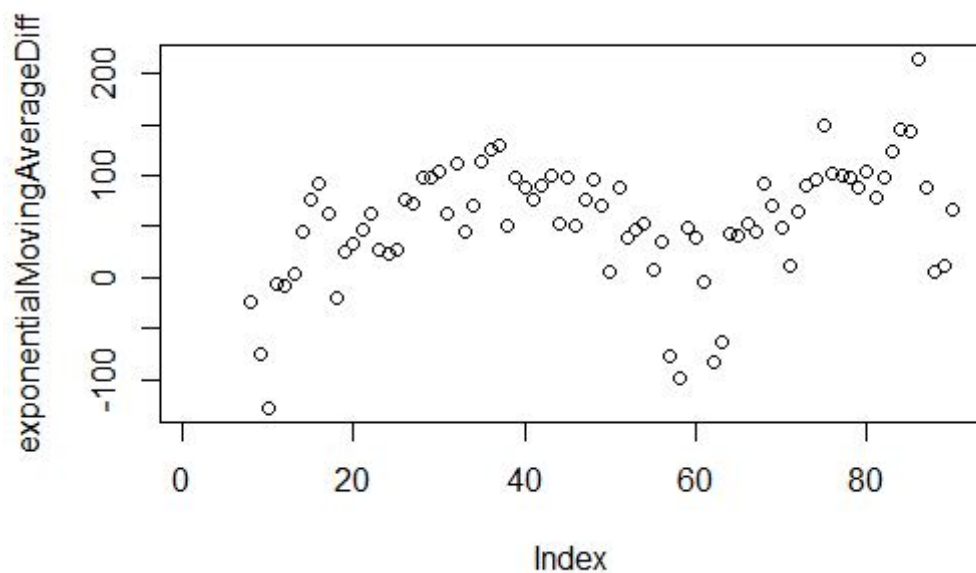
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```
summary(exponentialMovingAverageDiff)
```

Min.	1st Qu.	Median	Mean	3rd Qu.	Max.	NA's
-127.02	30.22	63.51	56.02	96.41	214.29	7

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```
plot(exponentialMovingAverageDiff)
```



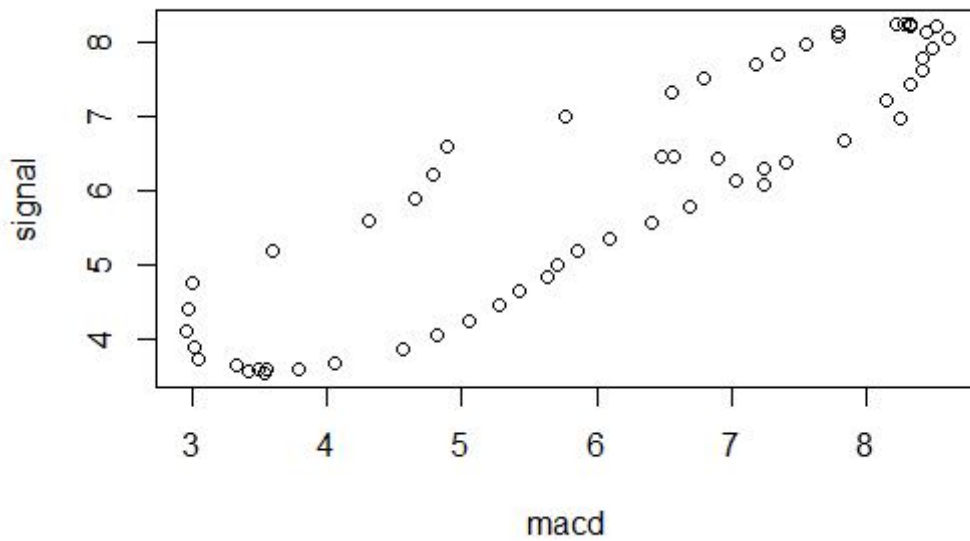
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```
# MACD Indicator
MACD <- MACD(Op(dataset),fast = 3, slow = 8, signal = 5, type = "EMA", histogram = TRUE)
head(MACD)
```

```
      macd signal
[1,]   NA    NA
[2,]   NA    NA
[3,]   NA    NA
[4,]   NA    NA
[5,]   NA    NA
[6,]   NA    NA
```

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```
plot(MACD)
```



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```
tail(MACD)
```

```
      macd  signal
[85,] 6.684577 5.796747
[86,] 7.231540 6.083705
[87,] 7.236006 6.314165
[88,] 6.893389 6.430010
[89,] 6.572852 6.458579
[90,] 6.476851 6.462233
```

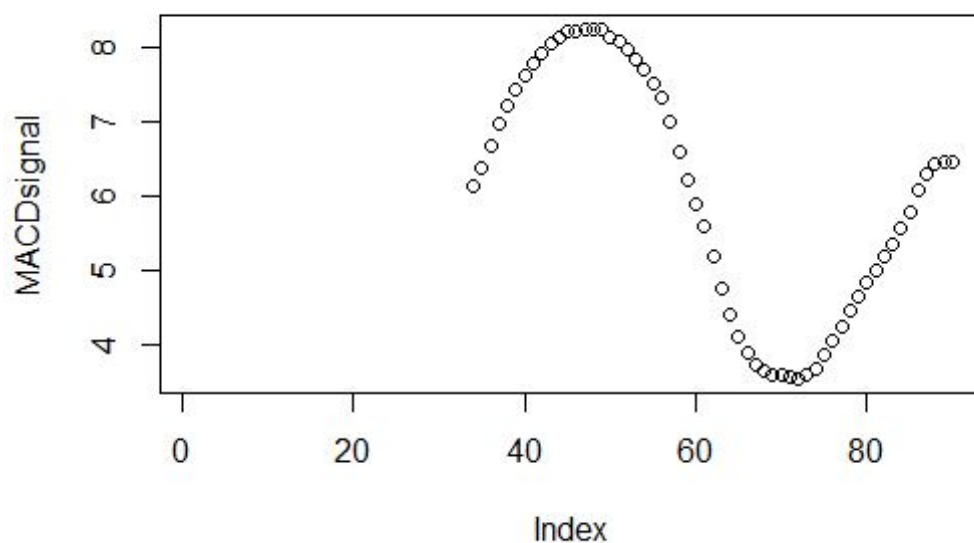
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```
summary(MACD)
```

	macd	signal
Min.	:2.966	Min. :3.560
1st Qu.:	4.780	1st Qu.:4.461
Median :	6.405	Median :6.231
Mean :	6.065	Mean :6.032
3rd Qu.:	7.544	3rd Qu.:7.633
Max. :	8.601	Max. :8.258
NA's :	25	NA's :33

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```
MACDsignal <- MACD[,2]
plot(MACDsignal)
```



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```
# Bollinger Band indicator
BollingerBands <- BBands(Op(dataset),n=8,sd=2)
head(BollingerBands)
```

	dn	mavg	up	pctB
[1,]	NA	NA	NA	NA
[2,]	NA	NA	NA	NA
[3,]	NA	NA	NA	NA
[4,]	NA	NA	NA	NA
[5,]	NA	NA	NA	NA
[6,]	NA	NA	NA	NA

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```
tail(BollingerBands)
```

```

      dn      mavg      up      pctB
[85,] 2343.950 2528.975 2714.000 0.9181996
[86,] 2335.982 2579.075 2822.167 0.9882400
[87,] 2385.597 2614.554 2843.511 0.7198366
[88,] 2430.075 2634.097 2838.120 0.4984131
[89,] 2490.421 2655.165 2819.909 0.4629577
[90,] 2546.632 2679.852 2813.073 0.6458009

```

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```
summary(BollingerBands)
```

```

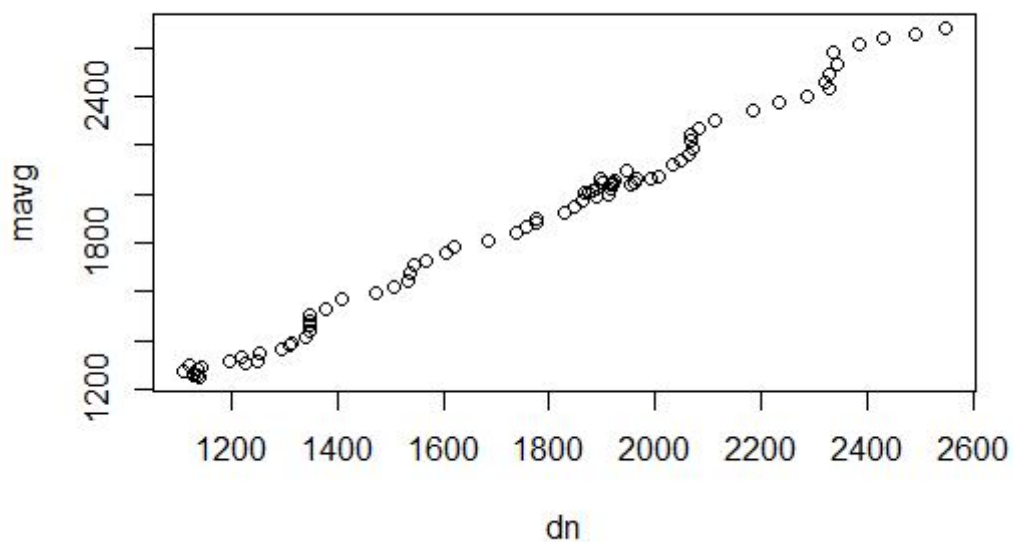
      dn      mavg      up
Min.   :1112   Min.   :1254   Min.   :1368
1st Qu.:1349   1st Qu.:1472   1st Qu.:1596
Median :1867   Median :1992   Median :2088
Mean   :1748   Mean   :1874   Mean   :2000
3rd Qu.:2005   3rd Qu.:2085   3rd Qu.:2216
Max.   :2547   Max.   :2680   Max.   :2844
NA's   :7      NA's   :7      NA's   :7

      pctB
Min.   :-0.0484
1st Qu.: 0.6486
Median : 0.7843
Mean    : 0.7189
3rd Qu.: 0.9027
Max.    : 1.0238
NA's    :7

```

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```
plot(BollingerBands)
```



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```
# % Change BB  
PercentageChngpctB <- BollingerBands[,4]  
head(PercentageChngpctB)
```

```
[1] NA NA NA NA NA NA
```

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```
tail(PercentageChngpctB)
```

```
[1] 0.9181996 0.9882400 0.7198366 0.4984131 0.4629577 0.6458009
```

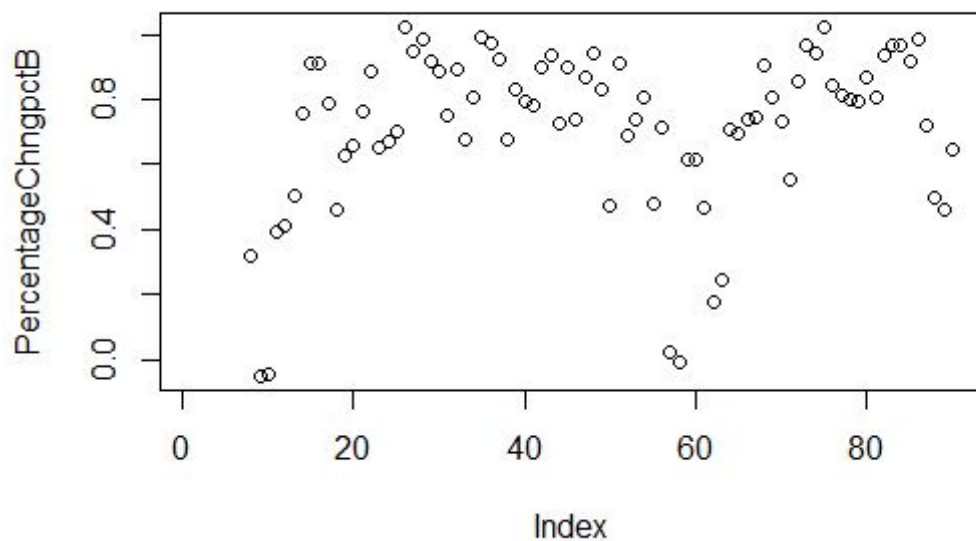
[Hide](#)

```
summary(PercentageChngpctB)
```

Min.	1st Qu.	Median	Mean	3rd Qu.	Max.	NA's
-0.0484	0.6486	0.7843	0.7189	0.9027	1.0238	7

[Hide](#)

```
plot(PercentageChngpctB)
```

[Hide](#)

```
# Price (Closes above Open = 1, Closes below Open = 0)  
Price=ifelse(dataset[4]>dataset[1], 1,0)  
head(Price)
```

GSPC.Close

Jan 2011	1
Feb 2011	1
Mar 2011	0
Apr 2011	1
May 2011	0
Jun 2011	0

Hide

tail(Price)

GSPC.Close

Jan 2018	1
Feb 2018	0
Mar 2018	0
Apr 2018	1
May 2018	1
Jun 2018	0

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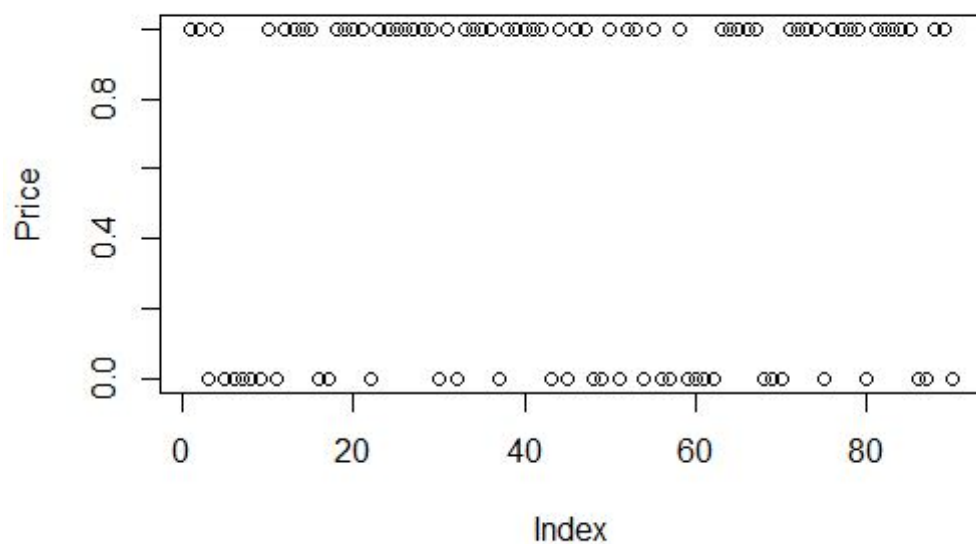
tail(dataset)

	GSPC.Open <dbl>	GSPC.High <dbl>	GSPC.L... <dbl>	GSPC.Close <dbl>	GSPC.Volume <dbl>	GSPC.Adjusted <dbl>
Jan 2018	2683.73	2872.87	2682.36	2823.81	76860120000	2823.81
Feb 2018	2816.45	2835.96	2532.69	2713.83	79579410000	2713.83
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May 2018	2642.96	2742.24	2594.62	2705.27	75617280000	2705.27
Jun 2018	2718.70	2791.47	2691.99	2718.37	77439710000	2718.37

6 rows

Hide

plot(Price)



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`dim(relativeStrengthIndex20)`

NULL

Hide

`dim(exponentialMovingAverage20)`

NULL

Hide

`dim(MACDsignal)`

NULL

Hide

`dim(PercentageChngpctB)`

NULL

Hide

`dim(Price)`

`[1] 90 1`

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```
dataset1 = data.frame(relativeStrengthIndex20, exponentialMovingAverage20, MACDsignal, PercentageChngpctB, Price)
# Size of Data
str(dataset1)
```

```
'data.frame':  90 obs. of  5 variables:
 $ relativeStrengthIndex20  : num  NA NA NA NA NA NA ...
 $ exponentialMovingAverage20: num  NA NA NA NA NA NA ...
 $ MACDsignal               : num  NA NA NA NA NA NA NA NA NA NA NA ...
 $ PercentageChngpctB       : num  NA NA NA NA NA NA ...
 $ GSPC.Close               : num  1 1 0 1 0 0 0 0 0 1 ...
```

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```
dim(dataset1)
```

```
[1] 90  5
```

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```
#Checking for missing data
d3=dataset1
for(i in 1:ncol(d3))
{
  print(colnames(d3[i]))
  print(sum(is.na(d3[i])))
}
```

```
[1] "relativeStrengthIndex20"
[1] 8
[1] "exponentialMovingAverage20"
[1] 7
[1] "MACDsignal"
[1] 33
[1] "PercentageChngpctB"
[1] 7
[1] "GSPC.Close"
[1] 0
```

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```
dataset1 = na.omit(dataset1)
#Checking for missing data again
dim(dataset1)
```

```
[1] 57  5
```

Hide

```
d3=dataset1
for(i in 1:ncol(d3))
{
  print(colnames(d3[i]))
  print(sum(is.na(d3[i])))
}
```

```
[1] "relativeStrengthIndex20"
[1] 0
[1] "exponentialMovingAverage20"
[1] 0
[1] "MACDsignal"
[1] 0
[1] "PercentageChngpctB"
[1] 0
[1] "GSPC.Close"
[1] 0
```

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```
colnames(dataset1)=c ("RSI20", "EMA20", "MACDsignal", "BB", "Price")
# Exploring the data set components
str(dataset1)
```

```
'data.frame':  57 obs. of  5 variables:
 $ RSI20      : num  73.5 79 81.7 83.7 69.8 ...
 $ EMA20      : num  1611 1644 1680 1717 1732 ...
 $ MACDsignal: num   6.14 6.39 6.68 6.99 7.22 ...
 $ BB         : num   0.806 0.993 0.97 0.925 0.678 ...
 $ Price      : num   1 1 1 0 1 1 1 1 1 0 ...
- attr(*, "na.action")=Class 'omit' Named int [1:33] 1 2 3 4 5 6 7 8 9 10 ...
.. ..- attr(*, "names")= chr [1:33] "Jan 2011" "Feb 2011" "Mar 2011" "Apr 2011" ...
```

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```
# Encoding the target feature as factor
dataset1$Price=as.factor(dataset1$Price)
# Splitting the dataset into the Training set and Test set
library(caTools)
set.seed(123)
split = sample.split(dataset1$Price, SplitRatio = 0.8)
training_set = subset(dataset1, split == TRUE)
test_set = subset(dataset1, split == FALSE)
# Feature Scaling (Normalization and dropping the predicted variable)
training_set[-5] = scale(training_set[-5])
test_set[-5] = scale(test_set[-5])
# Applying Kernel SVM Model on the Training set
library(e1071)
classifier = svm(formula = Price ~ .,
                 data = training_set,
                 type = 'C-classification',
                 kernel = 'radial')
# Predicting the Test set results
predict_val = predict(classifier, newdata = test_set[-5])
# Confusion Matrix
cm = table(test_set[, 5], predict_val)
print(cm)
```

```
predict_val
 0 1
0 0 4
1 1 6
```

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```
# Evaluating Model Accuracy on test data set using Confusion Matrix
Model_Accuracy=(cm[1,1] + cm[2,2])/ (cm[1,1] + cm[1,2] + cm[2,1] + cm[2,2])
print("Model Accuracy is")
```

```
[1] "Model Accuracy is"
```

Hide

```
print(Model_Accuracy)
```

```
[1] 0.5454545
```