

Applying Support Vector Machine (SVM) Algorithm to identify S&P500 trend

[Code ▾](#)[Hide](#)

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

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

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

```
[1] 1886    6
```

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```
str(dataset)
```

```
'data.frame':   1886 obs. of  6 variables:
 $ GSPC.Open    : num  1258 1273 1269 1276 1274 ...
 $ GSPC.High    : num  1276 1274 1278 1278 1277 ...
 $ GSPC.Low     : num  1258 1263 1265 1270 1262 ...
 $ GSPC.Close   : num  1272 1270 1277 1274 1272 ...
 $ GSPC.Volume  : num  4.29e+09 4.80e+09 4.76e+09 4.84e+09 4.96e+09 ...
 $ GSPC.Adjusted: num  1272 1270 1277 1274 1272 ...
```

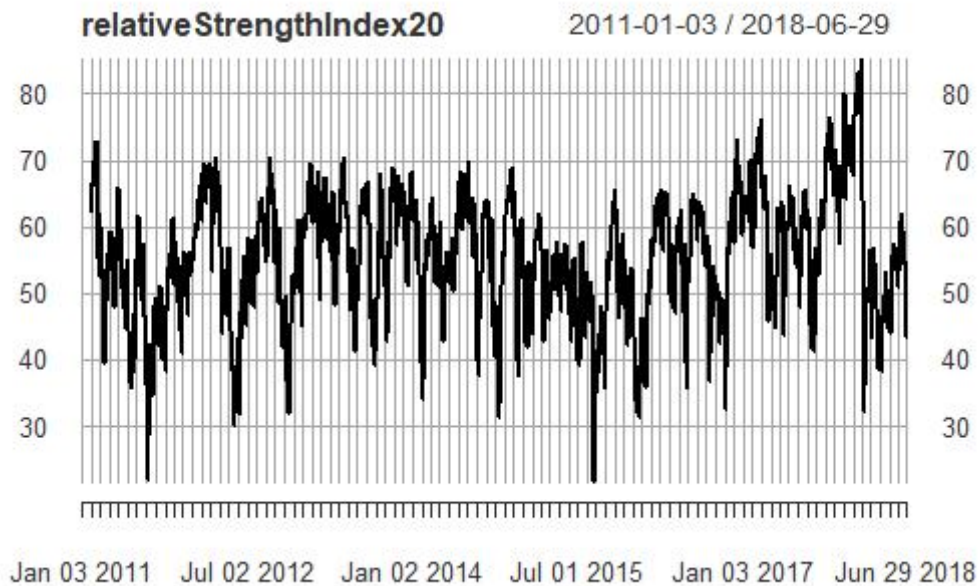
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```
#RSI indicator
relativeStrengthIndex20=RSI(Op(GSPC),n=20)
summary(relativeStrengthIndex20)
```

Index	rsi
Min. :2011-01-03	Min. :21.67
1st Qu.:2012-11-15	1st Qu.:49.28
Median :2014-10-01	Median :56.03
Mean :2014-10-01	Mean :55.47
3rd Qu.:2016-08-15	3rd Qu.:62.19
Max. :2018-06-29	Max. :85.12
	NA's :20

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```
plot(relativeStrengthIndex20)
```



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```
# Exponential Moving Average Indicator
exponentialMovingAverage20=EMA(Op(GSPC),n=20)
head(exponentialMovingAverage20)
```

	EMA
2011-01-03	NA
2011-01-04	NA
2011-01-05	NA
2011-01-06	NA
2011-01-07	NA
2011-01-10	NA

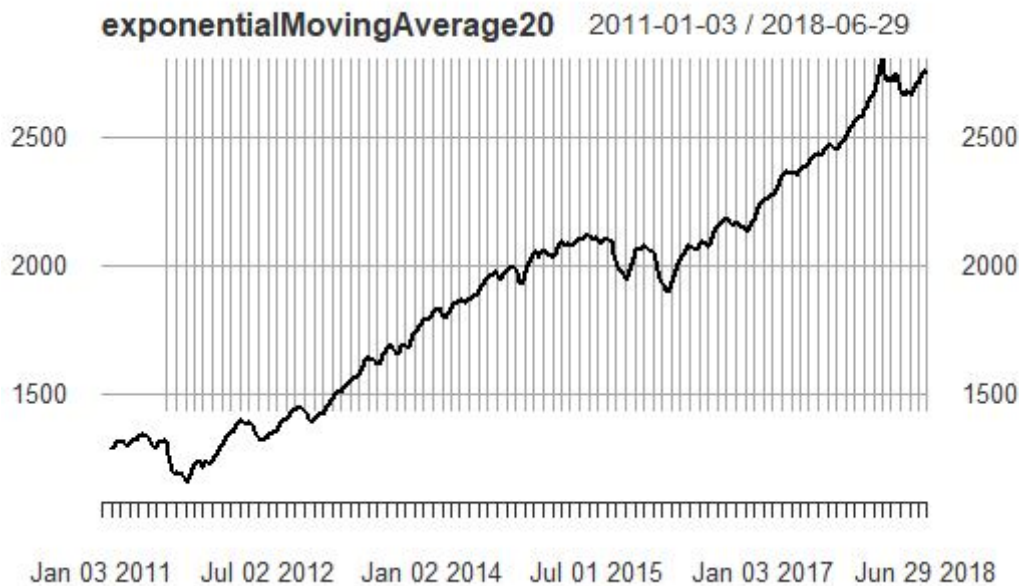
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```
summary(exponentialMovingAverage20)
```

Index	EMA
Min. :2011-01-03	Min. :1158
1st Qu.:2012-11-15	1st Qu.:1438
Median :2014-10-01	Median :1957
Mean :2014-10-01	Mean :1890
3rd Qu.:2016-08-15	3rd Qu.:2147
Max. :2018-06-29	Max. :2798
	NA's :19

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```
plot(exponentialMovingAverage20)
```



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```
# Difference in Exponential Moving Average
exponentialMovingAverageDiff <- Op(GSPC) - exponentialMovingAverage20
head(exponentialMovingAverageDiff)
```

	GSPC.Open
2011-01-03	NA
2011-01-04	NA
2011-01-05	NA
2011-01-06	NA
2011-01-07	NA
2011-01-10	NA

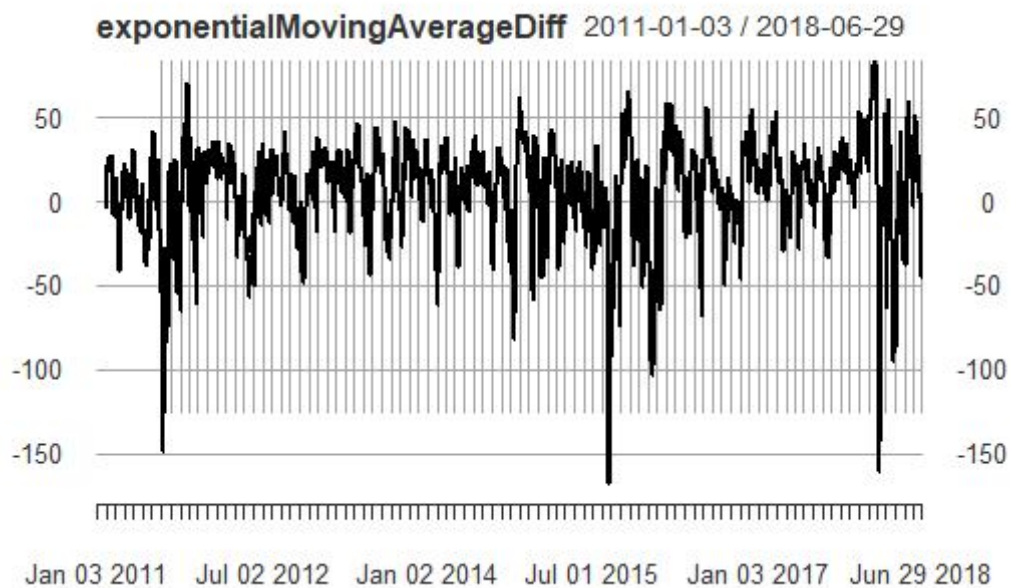
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```
summary(exponentialMovingAverageDiff)
```

Index	GSPC.Open
Min. :2011-01-03	Min. :-167.908
1st Qu.:2012-11-15	1st Qu.: -6.288
Median :2014-10-01	Median : 12.592
Mean :2014-10-01	Mean : 7.432
3rd Qu.:2016-08-15	3rd Qu.: 25.688
Max. :2018-06-29	Max. : 83.325
	NA's :19

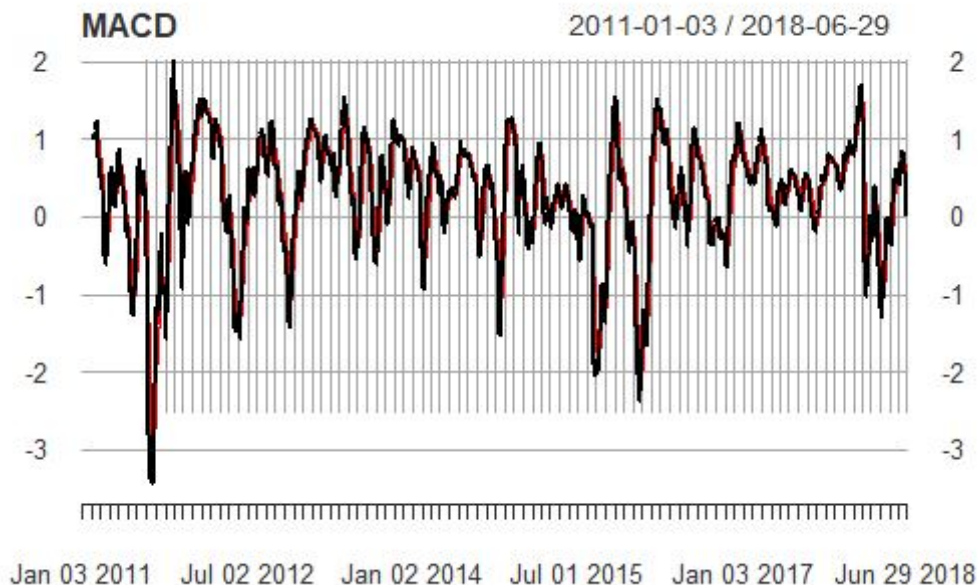
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```
plot(exponentialMovingAverageDiff)
```



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```
# MACD Indicator
MACD <- MACD(Op(GSPC),fast = 12, slow = 26, signal = 9, type = "EMA", histogram = TRUE)
plot(MACD)
```



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```
tail(MACD)
```

	macd	signal
2018-06-22	0.574743051	0.6812994
2018-06-25	0.471427824	0.6393251
2018-06-26	0.324827483	0.5764256
2018-06-27	0.224507694	0.5060420
2018-06-28	0.056802071	0.4161940
2018-06-29	0.007311846	0.3344176

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```
summary(MACD)
```

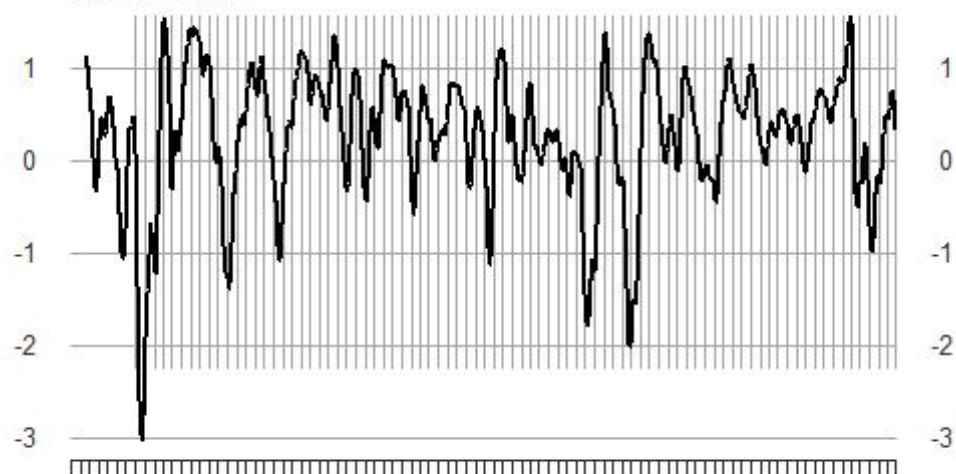
Index	macd	signal
Min. :2011-01-03	Min. : -3.44194	Min. : -3.0188
1st Qu.:2012-11-15	1st Qu.: -0.07795	1st Qu.: -0.0489
Median :2014-10-01	Median : 0.41028	Median : 0.3901
Mean :2014-10-01	Mean : 0.28198	Mean : 0.2801
3rd Qu.:2016-08-15	3rd Qu.: 0.78858	3rd Qu.: 0.7543
Max. :2018-06-29	Max. : 2.01064	Max. : 1.5568
	NA's :25	NA's :33

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

MACDsignal

2011-01-03 / 2018-06-29



Jan 03 2011 Jul 02 2012 Jan 02 2014 Jul 01 2015 Jan 03 2017 Jun 29 2018

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```
# Bollinger Band indicator
BollingerBands <- BBands(Op(GSPC),n=20,sd=2)
tail(BollingerBands)
```

	dn	mavg	up	pctB
2018-06-22	2701.463	2754.360	2807.256	0.56078445
2018-06-25	2704.032	2755.327	2806.621	0.37926017
2018-06-26	2707.758	2756.177	2804.596	0.14831230
2018-06-27	2713.733	2757.478	2801.223	0.16821581
2018-06-28	2708.057	2756.364	2804.670	-0.09695981
2018-06-29	2709.668	2756.785	2803.902	0.18530104

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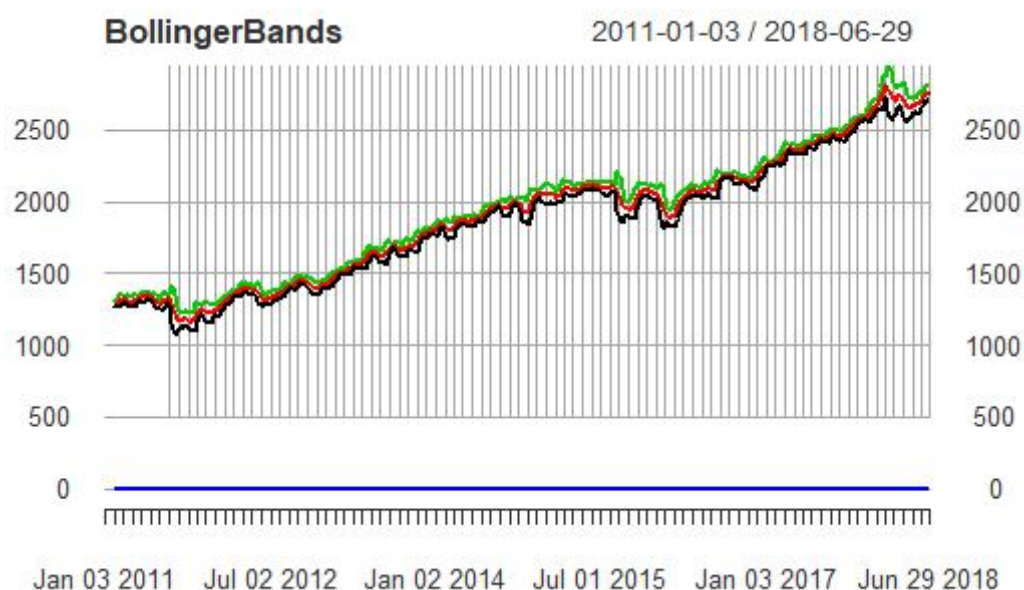
```
summary(BollingerBands)
```

Index	dn	mavg
Min. :2011-01-03	Min. :1075	Min. :1163
1st Qu.:2012-11-15	1st Qu.:1400	1st Qu.:1438
Median :2014-10-01	Median :1885	Median :1955
Mean :2014-10-01	Mean :1844	Mean :1890
3rd Qu.:2016-08-15	3rd Qu.:2094	3rd Qu.:2146
Max. :2018-06-29	Max. :2724	Max. :2802
	NA's :19	NA's :19

up	pctB
Min. :1222	Min. :-0.4976
1st Qu.:1471	1st Qu.: 0.3697
Median :2009	Median : 0.6752
Mean :1935	Mean : 0.5937
3rd Qu.:2183	3rd Qu.: 0.8456
Max. :2939	Max. : 1.2709
NA's :19	NA's :19

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```
plot(BollingerBands)
```

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

	pctB
2018-06-22	0.56078445
2018-06-25	0.37926017
2018-06-26	0.14831230
2018-06-27	0.16821581
2018-06-28	-0.09695981
2018-06-29	0.18530104

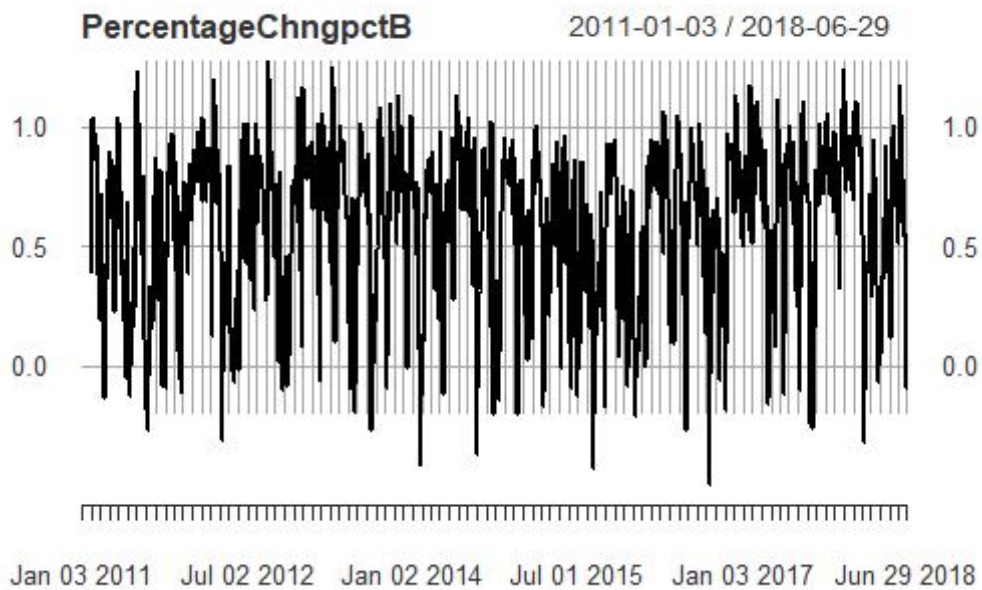
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```
summary(PercentageChngpctB)
```

Index	pctB
Min. :2011-01-03	Min. : -0.4976
1st Qu.:2012-11-15	1st Qu.: 0.3697
Median :2014-10-01	Median : 0.6752
Mean :2014-10-01	Mean : 0.5937
3rd Qu.:2016-08-15	3rd Qu.: 0.8456
Max. :2018-06-29	Max. : 1.2709
	NA's :19

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```
plot(PercentageChngpctB)
```



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

```
      GSPC.Close
2018-06-22      0
2018-06-25      0
2018-06-26      1
2018-06-27      0
2018-06-28      1
2018-06-29      0
```

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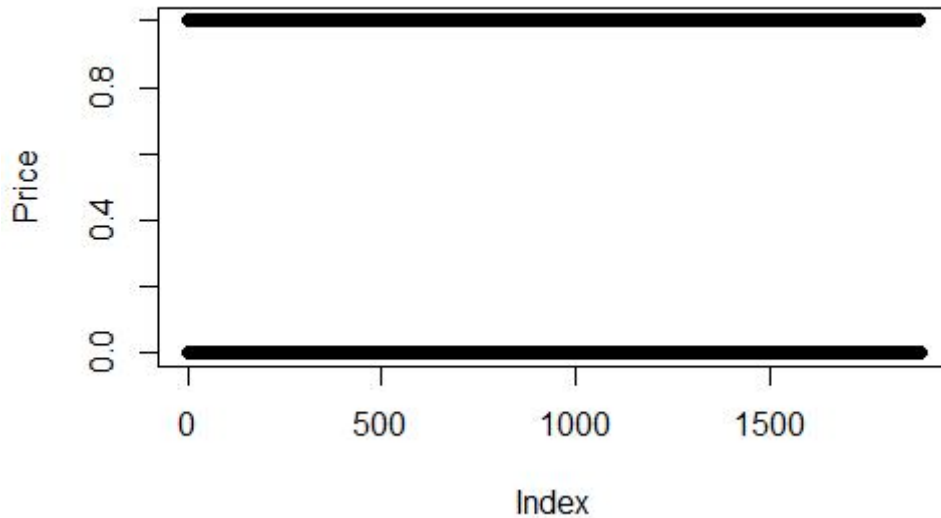
```
tail(dataset)
```

	GSPC.Op... <dbl>	GSPC.High <dbl>	GSPC.L... <dbl>	GSPC.Close <dbl>	GSPC.Volume <dbl>	GSPC.Adjusted <dbl>
2018-06-22	2760.79	2764.17	2752.68	2754.88	5450550000	2754.88
2018-06-25	2742.94	2742.94	2698.67	2717.07	3655080000	2717.07
2018-06-26	2722.12	2732.91	2715.60	2723.06	3555090000	2723.06
2018-06-27	2728.45	2746.09	2699.38	2699.63	3776090000	2699.63
2018-06-28	2698.69	2724.34	2691.99	2716.31	3428140000	2716.31
2018-06-29	2727.13	2743.26	2718.03	2718.37	3565620000	2718.37

6 rows

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```
plot(Price)
```



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```
dataset1<-data.frame(relativeStrengthIndex20, exponentialMovingAverage20, MACDsignal, Percentage
ChngpctB, Price)
# Size of Data
str(dataset1)
```

```
'data.frame':  1886 obs. of  5 variables:
 $ rsi      : num  NA NA NA NA NA NA NA NA NA NA NA ...
 $ EMA      : num  NA NA NA NA NA NA NA NA NA NA NA ...
 $ signal   : num  NA NA NA NA NA NA NA NA NA NA NA ...
 $ pctB     : num  NA NA NA NA NA NA NA NA NA NA NA ...
 $ GSPC.Close: num  1 0 1 0 0 0 1 1 0 1 ...
```

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```
dim(dataset1)
```

```
[1] 1886    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] "rsi"
[1] 20
[1] "EMA"
[1] 19
[1] "signal"
[1] 33
[1] "pctB"
[1] 19
[1] "GSPC.Close"
[1] 0
```

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

```
[1] 1853    5
```

[Hide](#)

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

```
[1] "rsi"
[1] 0
[1] "EMA"
[1] 0
[1] "signal"
[1] 0
[1] "pctB"
[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':  1853 obs. of  5 variables:
 $ RSI20      : num  72.9 71.9 58.9 55.1 55.2 ...
 $ EMA20      : num  1313 1316 1316 1315 1314 ...
 $ MACDsignal: num   1.12 1.14 1.13 1.08 1.02 ...
 $ BB         : num   0.926 0.867 0.522 0.385 0.378 ...
 $ Price      : num   1 0 0 0 1 1 0 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] "2011-01-03" "2011-01-04" "2011-01-05" "2011-01-06" ...
```

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```
# Encoding the target feature as factor
dataset1$Price=factor(dataset1$Price, levels = c(0, 1))
```

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```
# 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   3 169
1   2 197
```

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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"
```

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```
print(Model_Accuracy)
```

```
[1] 0.5390836
```