```
import numpy as np
import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt

df=pd.read_csv("/content/drive/MyDrive/datasets/bitcoin.csv")
df
```

	Date	Price
0	5/23/2019	7881.846680
1	5/24/2019	7987.371582
2	5/25/2019	8052.543945
3	5/26/2019	8673.215820
4	5/27/2019	8805.778320
362	5/19/2020	9729.038086
363	5/20/2020	9522.981445
364	5/21/2020	9081.761719
365	5/22/2020	9182.577148
366	5/23/2020	9180.045898
367 rows × 2 columns		

## df.head()

	Date	Price
0	5/23/2019	7881.846680
1	5/24/2019	7987.371582
2	5/25/2019	8052.543945
3	5/26/2019	8673.215820
4	5/27/2019	8805.778320

## df.tail()

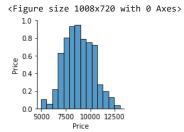
	Date	Price
362	5/19/2020	9729.038086
363	5/20/2020	9522.981445
364	5/21/2020	9081.761719
365	5/22/2020	9182.577148
366	5/23/2020	9180.045898

## df.isna().sum()

Date 0 Price 0 dtype: int64

## df.info()

```
plt.figure(figsize=(14,10))
sns.pairplot(data=df)
plt.show()
```



df.drop(['Date'],axis=1,inplace=True)
df

	Price	
0	7881.846680	
1	7987.371582	
2	8052.543945	
3	8673.215820	
4	8805.778320	
	•••	
362	9729.038086	
363	9522.981445	
364	9081.761719	
365	9182.577148	
366	9180.045898	
367 rows × 1 columns		

pday = 31
df['prediction']=df[['Price']].shift(-pday)
df

	Price	prediction	
0	7881.846680	10855.37109	
1	7987.371582	11011.10254	
2	8052.543945	11790.91699	
3	8673.215820	13016.23145	
4	8805.778320	11182.80664	
362	9729.038086	NaN	
363	9522.981445	NaN	
364	9081.761719	NaN	
365	9182.577148	NaN	
366	9180.045898	NaN	
367 rows × 2 columns			

df.tail()

```
Price prediction
       362 9729.038086
                                 NaN
       363 9522.981445
                                 NaN
x=np.array(df.drop(['prediction'],axis =1))
x=x[:len(df)-pday]
             [ אסיים, אסל ],
             [ 9477.677734],
              [ 9552.860352],
              [ 9519.145508],
              [ 9607.423828],
              [10085.62793],
              [10399.66895],
              [10518.17481],
              [10821.72656 ],
              [10970.18457],
              [11805.65332],
              [11478.16895],
              [11941.96875],
              [11966.40723],
              [11862.93652],
              [11354.02441 ],
              [11523.5791],
              [11382.61621],
              [10895.83008],
              [10051.7041 ],
              [10311.5459],
              [10374.33887],
              [10231.74414 ],
              [10345.81055],
              [10916.05371],
              [10763.23242],
              [10138.04981],
              [10131.05566],
              [10407.96484],
              [10159.96094],
              [10138.51758],
              [10370.82031],
              [10185.5
              [ 9754.422852],
             [ 9510.200195],
               9598.173828],
              [ 9630.664063],
              [ 9757.970703],
              [10346.76074],
              [10623.54004],
              [10594.49316],
              [10575.5332
              [10353.30273],
              [10517.25488],
              [10441.27637],
              [10334.97461],
              [10115.97559],
              [10178.37207],
              [10410.12695],
              [10360.54688],
              [10358.04883 ],
              [10347.71289],
              [10276.79395],
              [10241.27246],
              [10198.24805],
              [10266.41504],
              [10181.6416 ],
              [10019.7168 ],
              [10070.39258],
y=np.array(df['prediction'])
y=y[:-pday]
у
     array([10855.37109 , 11011.10254 , 11790.91699 , 13016.23145 ,
             11182.80664 , 12407.33203 , 11959.37109 , 10817.15527 , 10583.13477 , 10801.67773 , 11961.26953 , 11215.4375 ,
             10978.45996 , 11208.55078 , 11450.84668 , 12285.95801 ,
             12573.8125 , 12156.5127 , 11358.66211 , 11815.98633 , 11392.37891 , 10256.05859 , 10895.08984 , 9477.641602,
              9693.802734,\ 10666.48242\ ,\ 10530.73242\ ,\ 10767.13965\ ,
             10599.10547 , 10343.10645 , 9900.767578 , 9811.925781, 9911.841797 , 9870.303711 , 9477.677734 , 9552.860352 ,
```

```
9519.145508, 9607.423828, 10085.62793 , 10399.66895
             10518.17481 , 10821.72656 , 10970.18457 , 11805.65332 ,
             11478.16895 , 11941.96875 , 11966.40723 , 11862.93652 ,
             11354.02441 , 11523.5791 , 11382.61621 , 10895.83008 , 10051.7041 , 10311.5459 , 10374.33887 , 10231.74414 ,
             10345.81055 , 10916.05371 , 10763.23242 , 10138.04981 ,
             10131.05566 , 10407.96484 , 10159.96094 , 10138.51758 ,
             10370.82031 , 10185.5 , 9754.422852, 9510.200195,
              9598.173828, 9630.664063, 9757.970703, 10346.76074,
             10623.54004 , 10594.49316 , 10575.5332 , 10353.30273 ,
             10517.25488 , 10441.27637 , 10334.97461 , 10115.97559 ,
             10178.37207 , 10410.12695 , 10360.54688 , 10358.04883 ,
             10347.71289 , 10276.79395 , 10241.27246 , 10198.24805 ,
             10266.41504 , 10181.6416 , 10019.7168 , 10070.39258 ,
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              8343.276367, 8393.041992, 8259.992188, 8205.939453,
              8151.500488, 7988.155762, 8245.623047, 8228.783203,
              8595.740234, 8586.473633, 8321.756836, 8336.555664,
              8321.005859, 8374.686523, 8205.369141, 8047.526855,
              8103.911133, 7973.20752, 7988.560547, 8222.078125, 8243.720703, 8078.203125, 7514.671875, 7493.48877,
              8660.700195, 9244.972656, 9551.714844, 9256.148438,
              9427.6875 ,
                             9205.726563, 9199.584961,
                                                            9261.104492,
              9324.717773, 9235.354492, 9412.612305, 9342.527344,
              9360.879883, 9267.561523, 8804.880859, 8813.582031,
              9055.526367, 8757.788086, 8815.662109, 8808.262695,
              8708.094727, 8491.992188, 8550.760742, 8577.975586,
              8309.286133, 8206.145508, 8027.268066, 7642.75
              7296.577637, 7397.796875, 7047.916992, 7146.133789,
              7218.371094, 7531.663574, 7463.105957, 7761.243652,
              7569.629883, 7424.29248, 7321.988281, 7320.145508, 7252.034668, 7448.307617, 7546.996582, 7556.237793,
              7564.345215, 7400.899414, 7278.119629, 7217.427246,
              7243.134277, 7269.68457, 7124.673828, 7152.301758, 6932.480469, 6640.515137, 7276.802734, 7202.844238,
              7218.816406, 7191.158691, 7511.588867, 7355.628418,
              7322.532227, 7275.155762, 7238.966797, 7290.088379, 7317.990234, 7422.652832, 7292.995117, 7193.599121,
              7200.174316, 6985.470215, 7344.884277, 7410.656738,
                             7769.219238, 8163.692383, 8079.862793,
              7411.317383,
              7879.071289, 8166.554199, 8037.537598, 8192.494141,
              8144.194336, 8827.764648, 8807.010742, 8723.786133, 8929.038086, 8942.808594, 8706.245117, 8657.642578,
              8745.894531, 8680.875977, 8406.515625, 8445.43457,
              8367.847656, 8596.830078, 8909.819336, 9358.589844,
              9316.629883,
                             9508.993164, 9350.529297,
                                                            9392.875
                             9293.521484, 9180.962891, 9613.423828,
              9344.365234,
from sklearn.model_selection import train_test_split
xtrain,xtest,ytrain,ytest=train_test_split(x,y,test_size=.30,random_state=2)
pd_array=np.array(df.drop(['prediction'],axis=1))[:31]
pd array
     array([[ 7881.84668 ],
               7987.3715821.
               8052.543945],
               8673.21582 ],
             [ 8805.77832 ],
               8719.9619141.
               8659.487305],
               8319.472656],
               8574.5019531.
               8564.016602],
               8742.9580081.
               8208.995117],
               7707.7709961,
               7824.231445],
               7822.023438],
               8043.9511721.
               7954.12793 ],
               7688.077148],
               8000.32959 1.
               7927.714355],
               8145.857422],
               8230.923828],
               8693.8330081,
               8838.375
               8994.4882811.
               9320.3525391.
               9081.762695],
               9273.521484],
             [ 9527.160156],
```

```
[10144.55664]
             [10701.69141 ]])
from sklearn.svm import SVR
rbf=SVR(kernel='rbf', C=1e3, gamma=0.00001)
rbf.fit(xtrain,ytrain)
     SVR(C=1000.0, gamma=1e-05)
ypred=rbf.predict(xtest)
ypred
ray([ 8192.68472846, 8394.09790238, 7369.81300927, 9095.76431084,
              8409.47221331, 8384.46510418, 9200.08780352, 9418.33676949,
              9399.931077 , 8120.0289368 , 8210.48215429, 8498.55306989,
              9199.82415879, 9384.01519834, 8642.83029044, 8360.46251757,
              9424.60265631, 8381.06252515, 7957.9066053, 9286.17102072, 7373.11879255, 10107.09648525, 9076.72703499, 8532.58736185,
              8704.24392319, 8259.45785205, 8700.51987179, 8933.0962491 ,
             8904.67185472, 8913.79929938, 9222.38598162, 8228.64120208, 10251.93540835, 7816.66292128, 8595.29990347, 7786.05199406,
              8369.15389203, 9246.96270491, 8231.95043282, 8245.21210878,
              8356.90801522, 8179.95206842, 8368.30839932, 9077.79815736,
              7501.01763913, \quad 8329.99806907, \quad 7411.01752626, \quad 10930.27523917,
              8522.90442279, 9080.81508635, 8338.07740311, 8790.02204398,
              8008.4035956, 8662.60263294, 8002.32750099, 7768.05959887,
              7707.68764805, 7753.11947953, 8943.6226229 , 10570.05927785,
              8811.45067426, 9227.10977073, 8284.52086868, 8099.33466261, 8942.98548174, 9037.48880063, 7587.13641072, 8627.39745095,
              8334.71168594, 9231.47635783, 9232.31908547, 8223.93155281,
              9275.17203098, 10559.51467224, 9265.25856721, 9402.59171154,
             10360.12084234, 7453.8117267, 8293.18700591, 9204.83514423,
              8539.88248909, 8917.50457973, 9425.2677042, 7489.54608326, 9366.55675956, 7660.58614662, 10155.56379047, 9049.22938557,
              8120.65442868, \quad 8326.62531273, \quad 8030.70543127, \quad 7862.42169167,
              8621.43389648, 9238.10062117, 7383.39913664, 7403.77264991,
              8215.34171229, 9202.9699312, 8730.8021278, 8245.5840324,
              9012.85299391])
next30=rbf.predict(pd_array)
next30
     array([ 8952.77943964, 9359.38372038, 9425.2677042 , 8700.51987179,
              8014.40330924, 8460.34972772, 8766.4404633 , 9200.08780352,
              9090.64529823, 9118.37118648, 8338.07740311, 9265.25856721,
              7816.66292128, 8595.29990347, 8580.41822052, 9423.72593653, 9272.14235069, 7707.68764805, 9383.10084659, 9175.37512053, 9350.09741267, 9240.89299242, 8596.93254982, 7862.42169167,
              7371.37725653, 7768.05959887, 7271.51889762, 7585.65690956,
              8400.73965237, 8231.95043282, 10107.09648525])
print(next30)
print()
print(df.tail(pday))
     [ 8952.77943964 9359.38372038 9425.2677042 8700.51987179
       8014.40330924 8460.34972772 8766.4404633
                                                         9200.08780352
       9090.64529823 9118.37118648 8338.07740311 9265.25856721
       7816.66292128 8595.29990347 8580.41822052 9423.72593653
       9272.14235069 7707.68764805 9383.10084659 9175.37512053
       9350.09741267 9240.89299242 8596.93254982 7862.42169167
       7371.37725653 7768.05959887 7271.51889762 7585.65690956
       8400.73965237 8231.95043282 10107.09648525]
                 Price prediction
     336 7429.724609
                                 NaN
     337 7550.900879
                                 NaN
     338 7569.936035
                                 NaN
     339 7679.867188
                                 NaN
     340 7795.601074
                                 NaN
     341 7807.058594
                                 NaN
     342 8801.038086
                                 NaN
     343 8658.553711
                                 NaN
     344 8864.766602
                                 NaN
     345
           8988.596680
                                 NaN
     346 8897,468750
                                 NaN
     347
           8912.654297
                                 NaN
     348
           9003.070313
                                 NaN
          9268,761719
     349
                                 NaN
```

350	9951.518555	NaN
351	9842.666016	NaN
352	9593.896484	NaN
353	8756.430664	NaN
354	8601.795898	NaN
355	8804.477539	NaN
356	9269.987305	NaN
357	9733.721680	NaN
358	9328.197266	NaN
359	9377.013672	NaN
360	9670.739258	NaN
361	9726.575195	NaN
362	9729.038086	NaN
363	9522.981445	NaN
364	9081.761719	NaN
365	9182.577148	NaN
366	9180.045898	NaN

Colab paid products - Cancel contracts here