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212221230065
import numpy as np
import matplotlib.pyplot as plt
import pandas as pd
from sklearn.preprocessing import MinMaxScaler
from keras import layers
from keras.models import Sequential
dataset_train = pd.read_csv('trainset.csv')
dataset_train.columns
     Index(['Date', 'Open', 'High', 'Low', 'Close', 'Adj Close', 'Volume'], dtype='object')
dataset train.head()
             Date
                        0pen
                                    High
                                                Low
                                                         Close Adj Close
                                                                           Volume
                                                                                     \blacksquare
      0 2013-01-02 357.385559 361.151062 355.959839 359.288177 359.288177
                                                                           5115500
                                                                                     ıl.
      1 2013-01-03 360.122742 363.600128 358.031342 359.496826 359.496826
                                                                          4666500
      2 2013-01-04 362.313507 368.339294 361.488861 366.600616 366.600616 5562800
      3 2013-01-07 365.348755 367.301056 362.929504 365.001007 365.001007
                                                                          3332900
      4 2013-01-08 365.393463 365.771027 359.874359 364.280701 364.280701 3373900
 Next steps:
             Generate code with dataset train
                                               View recommended plots
train_set = dataset_train.iloc[:,1:2].values
type(train_set)
     numpy.ndarray
train_set.shape
     (1259, 1)
sc = MinMaxScaler(feature_range=(0,1))
training_set_scaled = sc.fit_transform(train_set)
training_set_scaled.shape
     (1259, 1)
X_train_array = []
y_train_array = []
for i in range(60, 1259):
 X_train_array.append(training_set_scaled[i-60:i,0])
  y_train_array.append(training_set_scaled[i,0])
X_train, y_train = np.array(X_train_array), np.array(y_train_array)
X_train1 = X_train.reshape((X_train.shape[0], X_train.shape[1],1))
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X_train.shape
```

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(1199, 60)
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length = 60
n_features = 1

```
model = Sequential()
model.add(layers.SimpleRNN(45,input_shape=(length,n_features)))
model.add(layers.Dense(1))
model.compile(optimizer='adam', loss='mse')
```

print("Name:S.Mohanraj")

print("Register Number: 212221230065")

model.summary()

Name:S.Mohanraj

Register Number: 212221230065

Model: "sequential"

Layer (type)	Output Shape	Param #
<pre>simple_rnn (SimpleRNN)</pre>	(None, 45)	2115
dense (Dense)	(None, 1)	46
delise (belise)	(None; 1)	
Total params: 2161 (8.44 KB)		
Trainable params: 2161 (8.44 KB)		
Non-trainable params: 0 (0.00 Byte)		

model.fit(X_train1,y_train,epochs=100, batch_size=32)

```
Epoch 1/100
38/38 [============ ] - 2s 11ms/step - loss: 0.0510
Epoch 2/100
Epoch 3/100
38/38 [===========] - 0s 11ms/step - loss: 0.0019
Epoch 4/100
38/38 [============] - 0s 11ms/step - loss: 0.0016
Epoch 5/100
38/38 [=============] - 0s 10ms/step - loss: 0.0015
Epoch 6/100
38/38 [============== ] - 0s 10ms/step - loss: 0.0013
Epoch 7/100
38/38 [=============] - 0s 10ms/step - loss: 0.0012
Epoch 8/100
38/38 [=============] - 0s 10ms/step - loss: 0.0011
Epoch 9/100
38/38 [============ ] - 0s 11ms/step - loss: 0.0010
Epoch 10/100
38/38 [============= ] - 0s 10ms/step - loss: 9.7325e-04
Epoch 11/100
38/38 [============ ] - 1s 18ms/step - loss: 9.1160e-04
Epoch 12/100
Epoch 13/100
Epoch 14/100
Epoch 15/100
38/38 [=============] - 0s 11ms/step - loss: 7.5042e-04
Epoch 16/100
38/38 [============ ] - 0s 11ms/step - loss: 7.2336e-04
Epoch 17/100
Epoch 18/100
38/38 [============== ] - 0s 11ms/step - loss: 6.9184e-04
Epoch 19/100
Epoch 20/100
Epoch 21/100
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Epoch 22/100
    38/38 [============] - 0s 10ms/step - loss: 5.9943e-04
    Epoch 23/100
    38/38 [=========== ] - 0s 10ms/step - loss: 6.6371e-04
    Epoch 24/100
    Epoch 25/100
    38/38 [============] - 0s 10ms/step - loss: 6.3415e-04
    Epoch 26/100
    Epoch 27/100
    38/38 [============= ] - 0s 11ms/step - loss: 6.4649e-04
    Epoch 28/100
    38/38 [=========== ] - 0s 11ms/step - loss: 5.3382e-04
    Epoch 29/100
    dataset_test = pd.read_csv('testset.csv')
test_set = dataset_test.iloc[:,1:2].values
test_set.shape
    (125, 1)
dataset_total = pd.concat((dataset_train['Open'],dataset_test['Open']),axis=0)
inputs = dataset_total.values
inputs = inputs.reshape(-1,1)
inputs_scaled=sc.transform(inputs)
X_{test} = []
for i in range(60,1384):
 X_test.append(inputs_scaled[i-60:i,0])
X_test = np.array(X_test)
X_test = np.reshape(X_test,(X_test.shape[0], X_test.shape[1],1))
X_test.shape
    (1324, 60, 1)
predicted_stock_price_scaled = model.predict(X_test)
predicted_stock_price = sc.inverse_transform(predicted_stock_price_scaled)
    42/42 [============= ] - 0s 4ms/step
print("Name:S.Mohanraj
                         Register Number:212221230065
plt.plot(np.arange(0,1384),inputs, color='red', label = 'Test(Real) Google stock price')
plt.plot(np.arange(60,1384),predicted_stock_price, color='blue', label = 'Predicted Google stock price')
plt.title('Google Stock Price Prediction')
plt.xlabel('Time')
plt.ylabel('Google Stock Price')
plt.legend()
plt.show()
```

Google Stock Price Prediction

