

DL_A4

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Name -Ashish Ramesh Walke || RollNo.- 4272 || Batch-B8

Assignment No 4: : Recurrent neural network (RNN) - Use the Google stock prices dataset and design a time series analysis and prediction system using RNN.

```
[1]: import pandas as pd
import numpy as np
```

```
[2]: train_df = pd.read_csv(r'Google_Stock_Price_Train.csv') #Path where the CSV_
    ↪file is stored.
```

```
[3]: train_df
```

```
[3]:
```

	Date	Open	High	Low	Close	Volume
0	1/3/2012	325.25	332.83	324.97	663.59	7,380,500
1	1/4/2012	331.27	333.87	329.08	666.45	5,749,400
2	1/5/2012	329.83	330.75	326.89	657.21	6,590,300
3	1/6/2012	328.34	328.77	323.68	648.24	5,405,900
4	1/9/2012	322.04	322.29	309.46	620.76	11,688,800
...
1253	12/23/2016	790.90	792.74	787.28	789.91	623,400
1254	12/27/2016	790.68	797.86	787.66	791.55	789,100
1255	12/28/2016	793.70	794.23	783.20	785.05	1,153,800
1256	12/29/2016	783.33	785.93	778.92	782.79	744,300
1257	12/30/2016	782.75	782.78	770.41	771.82	1,770,000

[1258 rows x 6 columns]

```
[4]: test_df = pd.read_csv(r'Google_Stock_Price_Test.csv') #Path where the CSV file_
    ↪is stored.
```

```
[5]: test_df
```

```
[5]:
```

	Date	Open	High	Low	Close	Volume
0	1/3/2017	778.81	789.63	775.80	786.14	1,657,300
1	1/4/2017	788.36	791.34	783.16	786.90	1,073,000
2	1/5/2017	786.08	794.48	785.02	794.02	1,335,200
3	1/6/2017	795.26	807.90	792.20	806.15	1,640,200
4	1/9/2017	806.40	809.97	802.83	806.65	1,272,400

5	1/10/2017	807.86	809.13	803.51	804.79	1,176,800
6	1/11/2017	805.00	808.15	801.37	807.91	1,065,900
7	1/12/2017	807.14	807.39	799.17	806.36	1,353,100
8	1/13/2017	807.48	811.22	806.69	807.88	1,099,200
9	1/17/2017	807.08	807.14	800.37	804.61	1,362,100
10	1/18/2017	805.81	806.21	800.99	806.07	1,294,400
11	1/19/2017	805.12	809.48	801.80	802.17	919,300
12	1/20/2017	806.91	806.91	801.69	805.02	1,670,000
13	1/23/2017	807.25	820.87	803.74	819.31	1,963,600
14	1/24/2017	822.30	825.90	817.82	823.87	1,474,000
15	1/25/2017	829.62	835.77	825.06	835.67	1,494,500
16	1/26/2017	837.81	838.00	827.01	832.15	2,973,900
17	1/27/2017	834.71	841.95	820.44	823.31	2,965,800
18	1/30/2017	814.66	815.84	799.80	802.32	3,246,600
19	1/31/2017	796.86	801.25	790.52	796.79	2,160,600

```
[6]: test_df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 20 entries, 0 to 19
Data columns (total 6 columns):
#   Column   Non-Null Count  Dtype
---  -
0   Date     20 non-null    object
1   Open     20 non-null    float64
2   High     20 non-null    float64
3   Low      20 non-null    float64
4   Close    20 non-null    float64
5   Volume   20 non-null    object
dtypes: float64(4), object(2)
memory usage: 1.1+ KB
```

Data Preprocessing

```
[7]: from sklearn.preprocessing import MinMaxScaler
```

```
[8]: # Convert 'Close' column to string type and remove commas
train_df['Close'] = train_df['Close'].astype(str).str.replace(',', '').
    ↪astype(float)
test_df['Close'] = test_df['Close'].astype(str).str.replace(',', '').
    ↪astype(float)
```

```
[9]: # Normalize the training and testing data separately
train_scaler = MinMaxScaler()
train_df['Normalized Close'] = train_scaler.fit_transform(train_df['Close'].
    ↪values.reshape(-1, 1))
test_scaler = MinMaxScaler()
```

```
test_df['Normalized Close'] = test_scaler.fit_transform(test_df['Close'].values.
↳reshape(-1, 1))
```

```
[10]: # Convert the data to the appropriate format for RNN
x_train = train_df['Normalized Close'].values[:-1].reshape(-1, 1, 1)
y_train = train_df['Normalized Close'].values[1:].reshape(-1, 1, 1)
x_test = test_df['Normalized Close'].values[:-1].reshape(-1, 1, 1)
y_test = test_df['Normalized Close'].values[1:].reshape(-1, 1, 1)
```

```
[11]: print("x_train shape: ",x_train.shape)
print("y_train shape: ",y_train.shape)
print("x_test shape: ",x_test.shape)
print("y_test shape: ",y_test.shape)
```

```
x_train shape: (1257, 1, 1)
y_train shape: (1257, 1, 1)
x_test shape: (19, 1, 1)
y_test shape: (19, 1, 1)
```

```
[12]: test_df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 20 entries, 0 to 19
Data columns (total 7 columns):
#   Column                Non-Null Count  Dtype
---  -
0   Date                  20 non-null    object
1   Open                  20 non-null    float64
2   High                  20 non-null    float64
3   Low                   20 non-null    float64
4   Close                 20 non-null    float64
5   Volume                20 non-null    object
6   Normalized Close     20 non-null    float64
dtypes: float64(5), object(2)
memory usage: 1.2+ KB
```

Building our Model

```
[13]: from keras.models import Sequential
from keras.layers import LSTM, Dense
```

```
WARNING:tensorflow:From C:\Users\Ashish\anaconda3\Lib\site-
packages\keras\src\losses.py:2976: The name
tf.losses.sparse_softmax_cross_entropy is deprecated. Please use
tf.compat.v1.losses.sparse_softmax_cross_entropy instead.
```

```
[14]: model = Sequential()
model.add(LSTM(4, input_shape=(1, 1)))
```

```
model.add(Dense(1))
model.compile(loss='mean_squared_error', optimizer='adam')
model.summary()
```

WARNING:tensorflow:From C:\Users\Ashish\anaconda3\Lib\site-packages\keras\src\backend.py:873: The name tf.get_default_graph is deprecated. Please use tf.compat.v1.get_default_graph instead.

WARNING:tensorflow:From C:\Users\Ashish\anaconda3\Lib\site-packages\keras\src\optimizers_init_.py:309: The name tf.train.Optimizer is deprecated. Please use tf.compat.v1.train.Optimizer instead.

Model: "sequential"

Layer (type)	Output Shape	Param #
lstm (LSTM)	(None, 4)	96
dense (Dense)	(None, 1)	5

=====
 Total params: 101 (404.00 Byte)
 Trainable params: 101 (404.00 Byte)
 Non-trainable params: 0 (0.00 Byte)
 =====

Building our Modely()

```
[15]: model.fit(x_train, y_train, epochs=50, batch_size=1, verbose=1)
```

Epoch 1/50

WARNING:tensorflow:From C:\Users\Ashish\anaconda3\Lib\site-packages\keras\src\utils\tf_utils.py:492: The name tf.ragged.RaggedTensorValue is deprecated. Please use tf.compat.v1.ragged.RaggedTensorValue instead.

1257/1257 [=====] - 8s 3ms/step - loss: 0.0492

Epoch 2/50

1257/1257 [=====] - 4s 3ms/step - loss: 0.0061

Epoch 3/50

1257/1257 [=====] - 4s 3ms/step - loss: 7.9633e-04

Epoch 4/50

1257/1257 [=====] - 3s 3ms/step - loss: 7.6611e-04

Epoch 5/50

1257/1257 [=====] - 4s 3ms/step - loss: 7.6682e-04

Epoch 6/50

1257/1257 [=====] - 4s 3ms/step - loss: 7.6955e-04

Epoch 7/50

1257/1257 [=====] - 4s 3ms/step - loss: 7.6812e-04

Epoch 8/50

1257/1257 [=====] - 4s 3ms/step - loss: 7.5853e-04
Epoch 9/50
1257/1257 [=====] - 4s 3ms/step - loss: 7.6718e-04
Epoch 10/50
1257/1257 [=====] - 4s 3ms/step - loss: 7.6423e-04
Epoch 11/50
1257/1257 [=====] - 4s 3ms/step - loss: 7.4985e-04
Epoch 12/50
1257/1257 [=====] - 4s 3ms/step - loss: 7.5008e-04
Epoch 13/50
1257/1257 [=====] - 3s 3ms/step - loss: 7.6696e-04
Epoch 14/50
1257/1257 [=====] - 4s 3ms/step - loss: 7.6529e-04
Epoch 15/50
1257/1257 [=====] - 4s 3ms/step - loss: 7.4453e-04
Epoch 16/50
1257/1257 [=====] - 4s 3ms/step - loss: 7.5847e-04
Epoch 17/50
1257/1257 [=====] - 3s 3ms/step - loss: 7.6124e-04
Epoch 18/50
1257/1257 [=====] - 4s 3ms/step - loss: 7.6739e-04
Epoch 19/50
1257/1257 [=====] - 4s 3ms/step - loss: 7.6197e-04
Epoch 20/50
1257/1257 [=====] - 4s 3ms/step - loss: 7.5157e-04
Epoch 21/50
1257/1257 [=====] - 4s 3ms/step - loss: 7.6060e-04
Epoch 22/50
1257/1257 [=====] - 4s 3ms/step - loss: 7.5561e-04
Epoch 23/50
1257/1257 [=====] - 4s 3ms/step - loss: 7.5242e-04
Epoch 24/50
1257/1257 [=====] - 4s 3ms/step - loss: 7.6012e-04
Epoch 25/50
1257/1257 [=====] - 4s 3ms/step - loss: 7.5640e-04
Epoch 26/50
1257/1257 [=====] - 4s 3ms/step - loss: 7.5184e-04
Epoch 27/50
1257/1257 [=====] - 4s 3ms/step - loss: 7.5468e-04
Epoch 28/50
1257/1257 [=====] - 4s 3ms/step - loss: 7.5712e-04
Epoch 29/50
1257/1257 [=====] - 4s 3ms/step - loss: 7.5842e-04
Epoch 30/50
1257/1257 [=====] - 4s 3ms/step - loss: 7.5292e-04
Epoch 31/50
1257/1257 [=====] - 4s 3ms/step - loss: 7.6508e-04
Epoch 32/50

```

1257/1257 [=====] - 4s 3ms/step - loss: 7.5352e-04
Epoch 33/50
1257/1257 [=====] - 4s 3ms/step - loss: 7.5178e-04
Epoch 34/50
1257/1257 [=====] - 4s 3ms/step - loss: 7.6670e-04
Epoch 35/50
1257/1257 [=====] - 4s 3ms/step - loss: 7.4873e-04
Epoch 36/50
1257/1257 [=====] - 4s 3ms/step - loss: 7.7639e-04
Epoch 37/50
1257/1257 [=====] - 4s 3ms/step - loss: 7.4424e-04
Epoch 38/50
1257/1257 [=====] - 4s 3ms/step - loss: 7.4913e-04
Epoch 39/50
1257/1257 [=====] - 4s 3ms/step - loss: 7.5499e-04
Epoch 40/50
1257/1257 [=====] - 4s 3ms/step - loss: 7.5923e-04
Epoch 41/50
1257/1257 [=====] - 4s 3ms/step - loss: 7.5799e-04
Epoch 42/50
1257/1257 [=====] - 4s 3ms/step - loss: 7.5670e-04
Epoch 43/50
1257/1257 [=====] - 4s 3ms/step - loss: 7.5790e-04
Epoch 44/50
1257/1257 [=====] - 4s 3ms/step - loss: 7.6846e-04
Epoch 45/50
1257/1257 [=====] - 4s 3ms/step - loss: 7.4961e-04
Epoch 46/50
1257/1257 [=====] - 4s 3ms/step - loss: 7.6863e-04
Epoch 47/50
1257/1257 [=====] - 4s 3ms/step - loss: 7.4557e-04
Epoch 48/50
1257/1257 [=====] - 4s 3ms/step - loss: 7.5276e-04
Epoch 49/50
1257/1257 [=====] - 4s 3ms/step - loss: 7.4168e-04
Epoch 50/50
1257/1257 [=====] - 4s 3ms/step - loss: 7.4874e-04

```

[15]: <keras.src.callbacks.History at 0x183a2c4de10>

Evaluating our Model

```

[16]: test_loss = model.evaluate(x_test, y_test)
      print('Testing loss: ', test_loss)

```

```

1/1 [=====] - 1s 1s/step - loss: 0.0250
Testing loss: 0.025006726384162903

```

Testing our Model

```
[17]: y_pred = model.predict(x_test)
```

```
1/1 [=====] - 1s 850ms/step
```

```
[18]: # Inverse transform the normalized values to get the actual values  
y_test_actual = test_scaler.inverse_transform(y_test.reshape(-1, 1))  
y_pred_actual = test_scaler.inverse_transform(y_pred.reshape(-1, 1))
```

```
[19]: i=1
```

```
[20]: print("Actual value: {:.2f}".format(y_test_actual[i][0]))  
      print("Predicted value: {:.2f}".format(y_pred_actual[i][0]))
```

```
Actual value: 794.02
```

```
Predicted value: 787.29
```