```
!pip install matplotlib-venn
     Requirement already satisfied: matplotlib-venn in /usr/local/lib/python3.10/dist-packages (0.11.9)
     Requirement already satisfied: matplotlib in /usr/local/lib/python3.10/dist-packages (from matplotlib-venn) (3.7.1)
     Requirement already satisfied: numpy in /usr/local/lib/python3.10/dist-packages (from matplotlib-venn) (1.22.4)
     Requirement already satisfied: scipy in /usr/local/lib/python3.10/dist-packages (from matplotlib-venn) (1.10.1)
     Requirement already satisfied: contourpy>=1.0.1 in /usr/local/lib/python3.10/dist-packages (from matplotlib->matplotlib-venn) (1.1.0)
     Requirement already satisfied: cycler>=0.10 in /usr/local/lib/python3.10/dist-packages (from matplotlib->matplotlib->enn) (0.11.0)
     Requirement already satisfied: fonttools>=4.22.0 in /usr/local/lib/python3.10/dist-packages (from matplotlib->matplotlib-venn) (4.40.0)
     Requirement already satisfied: kiwisolver>=1.0.1 in /usr/local/lib/python3.10/dist-packages (from matplotlib->matplotlib-venn) (1.4.4)
     Requirement already satisfied: packaging>=20.0 in /usr/local/lib/python3.10/dist-packages (from matplotlib->matplotlib-venn) (23.1)
     Requirement already satisfied: pillow>=6.2.0 in /usr/local/lib/python3.10/dist-packages (from matplotlib->matplotlib-venn) (8.4.0)
    Requirement already satisfied: pyparsing>=2.3.1 in /usr/local/lib/python3.10/dist-packages (from matplotlib->matplotlib->enn) (3.1.0)
     Requirement already satisfied: python-dateutil>=2.7 in /usr/local/lib/python3.10/dist-packages (from matplotlib->matplotlib-venn) (2.8.
     Requirement already satisfied: six>=1.5 in /usr/local/lib/python3.10/dist-packages (from python-dateutil>=2.7->matplotlib->matplotlib->
!apt-get -qq install -y libfluidsynth1
    E: Package 'libfluidsynth1' has no installation candidate
# https://pypi.python.org/pypi/libarchive
!apt-get -qq install -y libarchive-dev && pip install -U libarchive
import libarchive
    g previously unselected package libarchive-dev:amd64.
    database ... 123105 files and directories currently installed.)
    g to unpack .../libarchive-dev_3.4.0-2ubuntu1.2_amd64.deb ...
    g libarchive-dev:amd64 (3.4.0-2ubuntu1.2) ...
    up libarchive-dev:amd64 (3.4.0-2ubuntu1.2) ...
    ng triggers for man-db (2.9.1-1) ...
    ng libarchive
    ading libarchive-0.4.7.tar.gz (23 kB)
    ing metadata (setup.py) ... done
    ng nose (from libarchive)
    ading nose-1.3.7-py3-none-any.whl (154 kB)
                                       - 154.7/154.7 kB 13.9 MB/s eta 0:00:00
    wheels for collected packages: libarchive
    ng wheel for libarchive (setup.py) ... done
    d wheel for libarchive: filename=libarchive-0.4.7-py3-none-any.whl size=31629 sha256=47c6eeb5f1408c2f53b883d31109cd25d5932021a7b27e219b9
    in directory: /root/.cache/pip/wheels/3a/94/d0/6cd83c8a80a4236fd4cb2a1fd846ecf72ab1e0ac238c5951c0
    ully built libarchive
    ng collected packages: nose, libarchive
    ully installed libarchive-0.4.7 nose-1.3.7
import pandas as pd
from sklearn.model_selection import train_test_split
from sklearn.metrics import accuracy_score, precision_score, recall_score, f1_score
from keras.preprocessing.text import Tokenizer
from keras.models import Sequential
from keras.layers import Embedding, LSTM, GRU, Bidirectional, Dense, Dropout
from keras.callbacks import EarlyStopping
from keras.utils import pad_sequences
data = pd.read_csv("urdu-sentiment-corpus-v1.tsv", sep="\t")
data.head()
```

```
Tweet Class
data.info()
data.describe()
   <class 'pandas.core.frame.DataFrame'>
   RangeIndex: 1000 entries, 0 to 999
   Data columns (total 2 columns):
    #
      Column Non-Null Count Dtype
    0 Tweet 1000 non-null object
             999 non-null
    1
      Class
                         object
   dtypes: object(2)
   memory usage: 15.8+ KB
                                         1
                                              th
                              Tweet Class
    count
                               1000
                                     999
    unique
                                999
                                      3
          ...الله جانے وے ماہی تیرا پیار کی اے دل دی اوداسی
                                      Ν
     top
     freq
                                 2
                                     499
import warnings
warnings.filterwarnings('ignore')
data.isna().values.any()
   True
data.drop(data[data['Class']=='0'].index,inplace = True)
from sklearn.preprocessing import LabelEncoder
data['Class']=LabelEncoder().fit_transform(data['Class'])
piece = Tokenizer()
piece.fit_on_texts(data['Tweet'])
seq = piece.texts_to_sequences(data['Tweet'])
vocablen = len(piece.word_index) + 1
max_len = max(len(s) for s in seq)
x = pad_sequences(seq, maxlen=max_len, padding='post')
y = data['Class']
xtrain,xtest,ytrain,ytest = train_test_split(x,y,test_size=0.25,random_state=101)
from tensorflow.keras.layers import Dense, Dropout,Flatten,LSTM,Bidirectional,SimpleRNN
dummy1 = Sequential()
em = Embedding(vocablen,100,input_length=max_len,trainable = True)
dummv1.add(em)
dummy1.add(SimpleRNN(100,return_sequences=True ,dropout=0.3))
dummy1.add(Dense(512,activation='tanh'))
dummy1.add(Flatten())
dummy1.add(Dense(1,activation='sigmoid'))
dummy1.compile(loss ="binary_crossentropy", optimizer = 'adam', metrics=["accuracy"])
dummy1.fit(xtrain,ytrain,batch_size=20,epochs = 10)
   Epoch 2/10
   Epoch 3/10
   Epoch 4/10
   Epoch 5/10
   Epoch 6/10
   Epoch 7/10
```

```
Epoch 8/10
    37/37 [============= - - 2s 41ms/step - loss: 0.0077 - accuracy: 0.9864
    Epoch 9/10
    Epoch 10/10
    37/37 [=========== - 1s 36ms/step - loss: 0.0061 - accuracy: 0.9864
    <keras.callbacks.History at 0x7ff9c43669b0>
prediction1 = dummy1.predict(xtest)
    8/8 [======] - 0s 10ms/step
import numpy as np
prediction2 = np.argmax(prediction1,axis=1)
ac = accuracy_score(ytest,prediction2)
pres = precision_score(ytest,prediction2,average = 'macro')
f1 = f1_score(ytest,prediction2, average = 'macro')
rec = recall_score(ytest,prediction2, average = 'macro')
dataset = pd.DataFrame({
 'precision':[pres],
'recall' :[rec],
 'accuracy':[ac],
 'F1_score':[f1]
})
dataset
       precision recall accuracy F1_score
                                            ıl.
       0.244898
                  0.5 0.489796 0.328767
#BiLSTM
dummv2 = Sequential()
em = Embedding(vocablen,100,input_length=max_len,trainable = True)
dummy2.add(em)
dummy2.add(Bidirectional(LSTM(100,return sequences=True)))
dummy2.add(Dropout(0.3))
dummy2.add(Dense(2,activation='relu'))
dummy2.add(Flatten())
dummy2.add(Dense(1,activation='sigmoid'))
dummy2.compile(loss ="binary_crossentropy", optimizer = 'sgd', metrics=["accuracy"])
dummy2.summary()
   Model: "sequential_3"
    Layer (type)
                           Output Shape
                                                 Param #
    _____
     embedding_2 (Embedding)
                           (None, 37, 100)
                                                 549800
    bidirectional (Bidirectiona (None, 37, 200)
                                                 160800
    dropout (Dropout)
                           (None, 37, 200)
                                                 0
    dense_2 (Dense)
                           (None, 37, 2)
                                                 402
    flatten_1 (Flatten)
                           (None, 74)
                                                 0
    dense_3 (Dense)
                           (None, 1)
    ______
    Total params: 711,077
    Trainable params: 711,077
    Non-trainable params: 0
```

dummy2.fit(xtrain,ytrain,batch_size=30,epochs = 10)

```
Epoch 1/10
   25/25 [============== ] - 7s 81ms/step - loss: 0.6935 - accuracy: 0.4707
   Epoch 2/10
   Epoch 3/10
   Epoch 4/10
   25/25 [=============== ] - 2s 87ms/step - loss: 0.6929 - accuracy: 0.5197
   Epoch 5/10
   Epoch 6/10
   Epoch 7/10
   Epoch 8/10
   Epoch 9/10
   Epoch 10/10
   <keras.callbacks.History at 0x7ff9b3e71300>
prediction1 = dummy2.predict(xtest)
   8/8 [======] - 1s 21ms/step
prediction2 = np.argmax(prediction1,axis=1)
ac = accuracy score(ytest,prediction2)
pres = precision_score(ytest,prediction2,average = 'macro')
f1 = f1_score(ytest,prediction2, average = 'macro')
rec = recall score(ytest,prediction2, average = 'macro')
dataset2 = pd.DataFrame({
'precision':[pres],
'recall' :[rec],
'accuracy':[ac],
'F1_score':[f1]
dataset2
     precision recall accuracy F1_score
      0.244898
              0.5 0.489796 0.328767
#GRU
dummy3 = Sequential()
em = Embedding(vocablen,100,input_length=max_len,trainable = True)
dummy3.add(em)
dummy3.add(GRU(100, return_sequences=True))
dummy3.add(Dense(2,activation='relu'))
dummy3.add(Flatten())
dummy3.add(Dense(1,activation='sigmoid'))
dummy3.compile(loss ="binary_crossentropy", optimizer = 'sgd', metrics=["accuracy"])
dummy3.summary()
   Model: "sequential_4"
   Layer (type)
                     Output Shape
                                      Param #
   _____
   embedding_3 (Embedding)
                     (None, 37, 100)
                                      549800
   gru (GRU)
                                      60600
                     (None, 37, 100)
   dense_4 (Dense)
                     (None, 37, 2)
                                      202
   flatten_2 (Flatten)
                     (None, 74)
   dense_5 (Dense)
                     (None, 1)
```

Total params: 610,677 Trainable params: 610,677

```
Non-trainable params: 0
```

```
dummy3.fit(xtrain,ytrain,batch_size=20,epochs = 10)
   Epoch 1/10
   Epoch 2/10
   Epoch 3/10
  Epoch 4/10
  Epoch 5/10
   Epoch 6/10
   Epoch 7/10
  Epoch 8/10
  Epoch 9/10
  37/37 [============== - - 2s 66ms/step - loss: 0.6929 - accuracy: 0.5156
  Epoch 10/10
  <keras.callbacks.History at 0x7ff9be673a30>
prediction1 = dummy3.predict(xtest)
   8/8 [======] - 1s 10ms/step
prediction2 = np.argmax(prediction1,axis=1)
ac = accuracy_score(ytest,prediction2)
pres = precision_score(ytest,prediction2,average = 'macro')
f1 = f1_score(ytest,prediction2, average = 'macro')
rec = recall_score(ytest,prediction2, average = 'macro')
dummy3= pd.DataFrame({
'precision':[pres],
'recall' :[rec],
'accuracy':[ac],
'F1_score':[f1]
})
dummy3
     precision recall accuracy F1_score
     0.244898
             0.5 0.489796 0.328767
#LSTM
dummy4 = Sequential()
em = Embedding(vocablen,100,input_length=max_len,trainable = True)
dummy4.add(em)
dummy4.add(LSTM(100,return_sequences=True))
dummy4.add(Dropout(0.7))
dummy4.add(Dense(2,activation='relu'))
dummy4.add(Flatten())
dummy4.add(Dense(1,activation='sigmoid'))
dummy4.compile(loss ="binary_crossentropy", optimizer = 'sgd', metrics=["accuracy"])
dummy4.summary()
  Model: "sequential_5"
   Layer (type)
                    Output Shape
                                    Param #
   _____
   embedding_4 (Embedding)
                    (None, 37, 100)
                                    549800
   lstm_1 (LSTM)
                    (None, 37, 100)
                                    80400
   dropout_1 (Dropout)
                    (None, 37, 100)
                                    0
```

```
dense_6 (Dense)
                   (None, 37, 2)
                                 202
   flatten_3 (Flatten)
                   (None, 74)
   dense_7 (Dense)
                   (None, 1)
                                 75
  ______
  Total params: 630,477
  Trainable params: 630,477
  Non-trainable params: 0
dummy4.fit(xtrain,ytrain,batch size=20,epochs = 10)
  Epoch 1/10
  Epoch 2/10
  37/37 [=====
          Epoch 3/10
  37/37 [============ - - 2s 41ms/step - loss: 0.6931 - accuracy: 0.5048
  Epoch 4/10
  Epoch 5/10
  Epoch 6/10
  Epoch 7/10
  Epoch 8/10
  Epoch 9/10
  Epoch 10/10
  37/37 [=========== - - 2s 41ms/step - loss: 0.6929 - accuracy: 0.5170
  <keras.callbacks.History at 0x7ff9bc0b3c70>
prediction1= dummy4.predict(xtest)
prediction2 = np.argmax(prediction1,axis=1)
ac = accuracy_score(ytest,prediction2)
pres = precision_score(ytest,prediction2,average = 'macro')
f1 = f1_score(ytest,prediction2, average = 'macro')
rec = recall_score(ytest,prediction2, average = 'macro')
dummy4 = pd.DataFrame({
'precision':[pres],
'recall' :[rec],
'accuracy':[ac],
'F1_score':[f1]
})
dummy4
    precision recall accuracy F1_score
     0.244898
            0.5 0.489796 0.328767
```

✓ 0s completed at 20:38