NNDL ASSIGNMENT 9

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Task 9.1:

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Jupyter ICP_Assignment_9.1 Last Checkpoint: 2 minutes ago
       File Edit View Run Kernel Settings Help
       JupyterLab 📑 🐞 Python 3 (ipykernel) ○
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            [27]: import pandas as pd
                     from keras.preprocessing.text import Tokenizer
                    from keras.preprocessing.sequence import pad_sequences
from keras.models import Sequential
from keras.layers import Dense, Embedding, LSTM, SpatialDropoutID
                     from matplotlib import pyplot
                     from sklearn.model selection import train test split
                     from keras.utils.np_utils import to_categorical
                    import re
                    from sklearn.preprocessing import LabelEncoder
from keras.models import Sequential, load_model
                    import numpy
            [28]: #load data
                   data = pd.read_csv('data/Sentiment.csv')
data = data[['text', 'sentiment']]
                   data['text'] = data['text'].apply(lambda x: x.lower()) #convert text to Lowercase data['text'] = data['text'].apply((lambda x: re.sub('[^a-zA-z0-9\s]', '', x))) #remove special characters for idx, row in data.iterrows():
                    row[0] = row[0].replace('rt', ' ')
        . onto 1 . onto 1 . . chroce( . c ) /
[29]: #tokemization and padding
        max fatures = 2000
        tokenizer = Tokenizer(num_words=max_fatures, split=' ')
        tokenizer.fit_on_texts(data['text'].values)
        X = tokenizer.texts_to_sequences(data['text'].values)
        X = pad_sequences(X)
[30]: #encoding labels
        labelencoder = LabelEncoder()
integer_encoded = labelencoder.fit_transform(data['sentiment'])
       y = to_categorical(integer_encoded)
        X_{\text{train}}, X_{\text{test}}, Y_{\text{train}}, Y_{\text{test}} = train_test_split(X_{\text{y}}, test_size = 0.33, random_state = 42)
```

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[32]: #model architecture
          embed_dim = 128
lstm_out = 196
          def createmodel():
   model = Sequential()
               model.add(Embedding(max_fatures, embed_dim,input_length = X.shape[1])) #embedding layer
               model.add(LSTM(lstm_out, dropout=0.2, recurrent_dropout=0.2)) #LSTM Layer
model.add(Dense(3,activation='softmax')) #output Layer with softmax activation
model.compile(loss = 'categorical_crossentropy', optimizer='adam',metrics = ['accuracy'])
               return model
  [33]: #model training
          batch_size = 32
          model = createmodel()
          model.fit(X_train, Y_train, epochs = 1, batch_size=batch_size, verbose = 2)
          score,acc = model.evaluate(X_test,Y_test,verbose=2,batch_size=batch_size)
          print(score)
          print(acc)
          print(model.metrics_names)
          291/291 - 32s - loss: 0.8264 - accuracy: 0.6423
144/144 - 2s - loss: 0.7607 - accuracy: 0.6693
0.7606955766677856
           0.669287919998169
          ['loss', 'accuracy']
         ['loss', 'accuracy']
[34]: #saving and Loading model
        model.save('model.h5')#save the trained model
        mod = load_model('model.h5')
        print(mod.summary())
        Model: "sequential_4"
         Layer (type)
                                             Output Shape
                                                                               Param #
         embedding_4 (Embedding) (None, 28, 128)
                                                                               256000
        lstm_4 (LSTM)
                                            (None, 196)
                                                                               254800
         dense_4 (Dense)
                                             (None, 3)
                                                                               591
         Total params: 511,391
        Trainable params: 511,391
Non-trainable params: 0
         None
```

```
[36]: #prediction

tx + [['A lot of good things are happening. We are respected again throughout the world, and thats a great 'thing_@realDonaldrump']]

may_data = pd.DataFrame(tx, index-range(0, 1, 1), columns-list('t'))

may_data['t'] = max_dfi['t'] = paply(lambda x: re.sub('['a-zA-z0-9\s]', '', x)))

features = 2000

tokenizer = Tokenizer(num_words-features, split='')

tokenizer.fit_on_text(nam_data['t'].values))

X = pad_sequence(x, max_data['t'].values))

Out = mod.predict(X)

print(out)

print(numpy.where(max(out[0])), ":", (max(out[0])))

print(mod.summary())

[[0.acign274 o.07214401 0.12661776]]

(array([0], dtype=int64),) : 0.80123734

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Task 9.2:

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[14]: import pandas as pd from keras.preprocessing.text import Tokenizer
                                                                                                                                                       {\bf from} \ \ {\bf keras.preprocessing.sequence} \ \ {\bf import} \ \ {\bf pad\_sequences}
       from keras.models import Sequential from keras.layers import Dense, Embedding, LSTM, SpatialDropout1D
       from matplotlib import pyplot
from sklearn.model_selection import train_test_split
       from keras.utils.np_utils import to_categorical
       import re
        from sklearn.preprocessing import LabelEncoder
        from keras.wrappers.scikit_learn import KerasClassifier
        from sklearn.model_selection import GridSearchCV
[15]: #load data
       data = pd.read_csv('data/Sentiment.csv')
       data = data[['text','sentiment']]
       data['text'] = data['text'].apply(lambda x: x.lower()) #convert text to lowercase
       data['text'] = data['text'].apply((lambda x: re.sub('[^a-zA-z0-9\s]', '', x))) #remove special characters
[16]: for idx, row in data.iterrows(): #remove rretweet tag
        row[0] = row[0].replace('rt', '')
[17]: #tokenization and padding
       max_fatures = 2000
       tokenizer = Tokenizer(num_words=max_fatures, split=' ')
tokenizer.fit_on_texts(data['text'].values)
       X = tokenizer.texts_to_sequences(data['text'].values)
       X = pad_sequences(X)
```

```
[18]: #model architecture
embed_dim = 128
lstm_out = 196
def_createmodel():
    model = Sequential()
    model = Sequential()
    model.add(Embedding(max_fatures, embed_dim,input_length = X.shape[1]))
    model.add(LSTM(lstm_out, dropout=0.2, recurrent_dropout=0.2))
    model.add(Dense(3,activation='softmax'))
    model.compile(loss = 'categorical_crossentropy', optimizer='adam',metrics = ['accuracy'])
    return model

[19]: #encoding_labels
labelencoder = LabelEncoder()
integer_encoded = labelEncoder.fit_transform(data['sentiment'])
y = to_categorical(integer_encoded)

[20]: #train-test_split
X_train, X_test, Y_train, Y_test = train_test_split(X,y, test_size = 0.33, random_state = 42)
```

```
X_train, X_test, Y_train, Y_test = train_test_split(X,y, test_size = 0.33, random_state = 42)

[21]: #model training
batch_size = 32
model = createmodel()
model.fit(X_train, Y_train, epochs = 1, batch_size-batch_size, verbose = 2)
score_acc = model.evaluate(X_test,Y_test,verbose=2,batch_size)
print(score)
print(acc)
print(model.metrics_names)

291/291 - 31s - loss: 0.8313 - accuracy: 0.6416
144/144 - 4s - loss: 0.7781 - accuracy: 0.6566
0.7780812978744507
0.65661859512291
['loss', 'accuracy']

[22]: #grid_search_for_hyperparameter_tuning
print(X_train.shape,Y_train.shape)
model = KerasClassifier(build_fn=createmodel,verbose=0)
epochs = [1, 2]
param_grid=dict(epochs=epochs)
grid = GridSearch(V(estisator=model)_param_grid=param_grid, n_jobs=1)
grid_creating_field(X_train, Y_train,batch_size=32)
print(S_test_shape,Y_train.shape)
print(S_test_shape,Y_train.shape)
indel = KerasClassifier(build_fn=createmodel,verbose=0)
epochs = [1, 2]
param_grid=dict(epochs=epochs)
grid = GridSearch(V(estisator=model)_param_grid, n_jobs=1)
grid_result_grid_frit(X_train, Y_train,batch_size=32)
print("Esst: Kf using %s" % (grid_result.best_params_))

(9293, 28) (9293, 3)
(4578, 28) (4578, 3)
Best: 0.674915 using ('epochs': 2)
```