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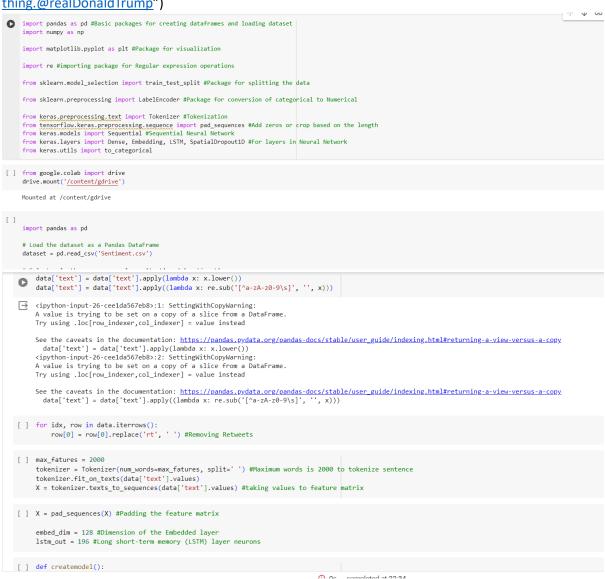
Neural Networks & Deep Learning - Assignment - 9

Github link: https://github.com/ShaikRumana301/Neural-Network-DL-ICP-9.git

Video Link: https://drive.google.com/file/d/1-d_RUX0323BMBBMtZ0TOewoc8l5f8F--/view?usp=sharing

In class programming:

1. Save the model and use the saved model to predict on new text data (ex, "A lot of good things are happening. We are respected again throughout the world, and that's a great thing.@realDonaldTrump")



```
[ ] def createmodel():
                   createmodel():
model = Sequential() #Sequential Neural Network
model.add(Embedding(max_fatures, embed_dim,input_length = X.shape[1])) #input dimension 2000 Neurons, output dimension 128 Neurons
model.add(LSTM(lstm_out, dropout=0.2, recurrent_dropout=0.2)) #Drop out 20%, 196 output Neurons, recurrent dropout 20%
model.add(Dense(3,activation='softmax')) #3 output neurons[positive, Neutral, Negative], softmax as activation
model.compile(loss = 'categorical_crossentropy', optimizer='adam',metrics = ['accuracy']) #Compiling the model
                     return model
          # print(model.summary())
  [ ] labelencoder = LabelEncoder() #Applying label Encoding on the label matrix integer_encoded = labelencoder.fit_transform(data['sentiment']) #fitting the model
             y = to_categorical(integer_encoded)
X_train, X_test, Y_train, Y_test = train_test_split(X,y, test_size = 0.33, random_state = 42) #67% training
  [ ] batch_size = 32 #Batch size 32
             model = createmodel() #function call to Sequential Neural Network
model.fit(X_train, Y_train, epochs = 1, batch_size=batch_size, verbose = 2) #verbose the higher, the more messages
score,acc = model.evaluate(X_test,Y_test,verbose=2,batch_size=batch_size) #evaluating the model
             print(score)
           291/291 - 51s - loss: 0.8261 - accuracy: 0.6442 - 51s/epoch - 175ms/step
144/144 - 3s - loss: 0.7469 - accuracy: 0.6689 - 3s/epoch - 22ms/step
0.7469469904899597
             0.6688510179519653
  [ ] print(model.metrics_names) #metrics of the model
            ['loss', 'accuracy']
      1. Save the model and use the saved model to predict on new text data (ex, "A lot of good things are happening. We are respected again
           throughout the world, and that's a great thing.@realDonaldTrump")
 [ ] model.save('sentimentAnalysis.h5') #Saving the model
         /usr/local/lib/python3.10/dist-packages/keras/src/engine/training.py:3103: UserWarning: You are saving your model as an HDF5 file via `model.save()`. This file format is considered legacy. Washing_api.save_model(
 [ ] from keras.models import load_model #Importing the package for importing the saved model model= load_model('sentimentAnalysis.h5') #loading the saved model
  [ ] print(integer_encoded)
print(data['sentiment'])
          [1 2 1 ... 2 0 2]
                       Neutral
Positive
Neutral
Positive
Positive
          13866 Negative
13867 Positive
13868 Positive
13869 Negative
13870 Positive
           Name: sentiment, Length: 13871, dtype: object
[] # Predicting on the text data
sentence = ['A lot of good things are happening. We are respected again throughout the world, and that is a great thing.@realDonaldTrump']
sentence = tokenizer.texts_to_sequences(sentence) # Tokenizing the sentence
sentence = pad_sequences(sentence, maxlen=28, dtype='int27', value=0) # Predicting the sentence
sentiment_probs = model.predict(sentence, batch_size=1, verbose=2)[0] # Predicting the sentence text
sentiment = np.argmax(sentiment_probs)
         print(sentiment_probs)
if sentiment == 0:
         print("Neutral")
elif sentiment < 0:
        sentiment < 0:
    print("Negative")
elif sentiment > 0:
    print("Positive")
else:
                 print("Cannot be determined")
         1/1 - 0s - 259ms/epoch - 259ms/step
[0.5638268    0.13124636    0.30492687]
Neutral
```

2. Apply GridSearchCV on the source code provided in the class

```
2. Apply GridSearchCV on the source code provided in the class
[2] pip install scikeras
  Collecting scikeras
Downloading scikeras-0.12.0-py3-none-any.whl (27 kB)
Requirement already satisfied: packaging>=0.21 in /usr/local/lib/python3.10/dist-packages (from scikeras) (24.0)
Requirement already satisfied: scikit-learn>=1.0.0 in /usr/local/lib/python3.10/dist-packages (from scikit-learn>=1.0.0->scikeras) (1.2.2)
Requirement already satisfied: numpy>=1.17.3 in /usr/local/lib/python3.10/dist-packages (from scikit-learn>=1.0.0->scikeras) (1.25.2)
Requirement already satisfied: scipy>=1.3.2 in /usr/local/lib/python3.10/dist-packages (from scikit-learn>=1.0.0->scikeras) (1.11.4)
Requirement already satisfied: joblib>=1.1.1 in /usr/local/lib/python3.10/dist-packages (from scikit-learn>=1.0.0->scikeras) (1.3.2)
Requirement already satisfied: threadpooletl>=2.0.0 in /usr/local/lib/python3.10/dist-packages (from scikit-learn>=1.0.0->scikeras) (3.4.0)
Totalling collected packages: scikeras
        Installing collected packages: scikeras
Successfully installed scikeras-0.12.0
  [ ] from sklearn.model_selection import train_test_split
         from scikeras.wrappers import KerasClassifie
        from sklearn.model_selection import GridSearchCV
        model = KerasClassifier(build_fn=createmodel(),verbose=2) #initiating model to test performance by applying multiple hyper parameters
        batch_size= [10, 20, 40] #hyper parameter batch_size
epochs = [1, 2] #hyper parameter no. of epochs
        param_grid= {batch_size':batch_size, 'epochs':epochs} #creating dictionary for batch size, no. of epochs
grid = GridSearchCV(estimator=model, param_grid=param_grid) #Applying dictionary with hyper parameters
grid_result= grid.fit(X_train,Y_train) #Fitting the model
       print("Best: %f using %s" % (grid_result.best_score_, grid_result.best_params_)) #best score, best hyper parameters
  186/186 - 24s - 10ss: 0.6869 - accuracy: 0.7076 - 24s/epocn - 129ms/step
  47/47 - 1s - loss: 0.7555 - accuracy: 0.6799 - 737ms/epoch - 16ms/step
  WARNING:tensorflow:Layer lstm_29 will not use cuDNN kernels since it doesn't meet the criteria. It will
  generic GPU kernel as fallback when running on GPU.
  Epoch 1/2
  186/186 - 36s - loss: 0.8497 - accuracy: 0.6370 - 36s/epoch - 192ms/step
  Epoch 2/2
  186/186 - 26s - loss: 0.6874 - accuracy: 0.7052 - 26s/epoch - 139ms/step
  47/47 - 1s - loss: 0.7363 - accuracy: 0.6889 - 748ms/epoch - 16ms/step
  WARNING:tensorflow:Layer lstm_30 will not use cuDNN kernels since it doesn't meet the criteria. It will
  generic GPU kernel as fallback when running on GPU.
  Epoch 1/2
  186/186 - 37s - loss: 0.8370 - accuracy: 0.6371 - 37s/epoch - 198ms/step
  Epoch 2/2
  186/186 - 26s - loss: 0.6795 - accuracy: 0.7098 - 26s/epoch - 140ms/step
  47/47 - 1s - loss: 0.7777 - accuracy: 0.6652 - 730ms/epoch - 16ms/step
  WARNING:tensorflow:Layer lstm_31 will not use cuDNN kernels since it doesn't meet the criteria. It will
  generic GPU kernel as fallback when running on GPU.
  Epoch 1/2
  465/465 - 74s - loss: 0.8138 - accuracy: 0.6524 - 74s/epoch - 159ms/step
  Epoch 2/2
  465/465 - 62s - loss: 0.6739 - accuracy: 0.7108 - 62s/epoch - 134ms/step
  Best: 0.681371 using {'batch_size': 20, 'epochs': 2}
```