ICP9

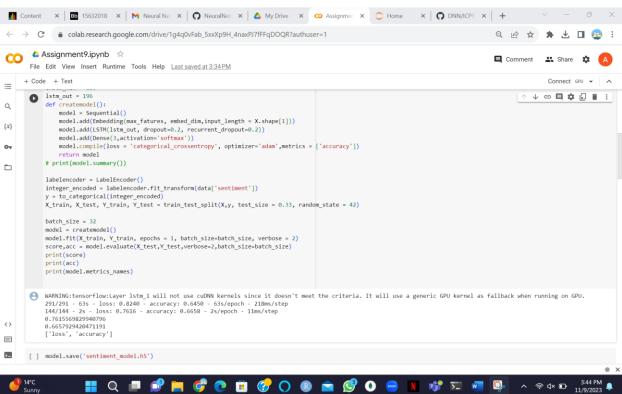
Angel Reddy Nakkala

700748217

GitHub Link: https://github.com/angelreddy09/DNN/tree/main/ICP9

```
import pandas as pd # data processing, CSV file I/O (e.g. pd.read csv)
from keras.preprocessing.text import Tokenizer
from tensorflow.keras.preprocessing.sequence import 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
data = pd.read csv('Sentiment.csv')
# Keeping only the neccessary columns
data = data[['text', 'sentiment']]
data['text'] = data['text'].apply(lambda x: x.lower())
data['text'] = data['text'].apply((lambda x: re.sub('[^a-zA-z0-9\s]', '',
x)))
for idx, row in data.iterrows():
    row[0] = row[0].replace('rt', '')
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)
embed dim = 128
lstm out = 196
def createmodel():
    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
# print(model.summary())
labelencoder = LabelEncoder()
integer encoded = labelencoder.fit transform(data['sentiment'])
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)
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)
```



```
model.save('sentiment_model.h5')
```

This code loads the saved model using the load_model function, and then preprocesses the new text data in the same way as the training data. The predict method is called on the loaded model to get the predicted class probabilities for the new text data. The class with the highest probability is chosen as the predicted sentiment. The predicted sentiment and probabilities are then printed to the console. To apply GridSearchCV on the provided source code, we can use the GridSearchCV class from sklearn to search for the best combination of hyperparameters for the LSTM model. The hyperparameters that can be tuned are the number of LSTM units, the dropout rate, and the learning rate of the optimizer

```
from keras.models import load_model
import numpy as np

loaded_model = load_model('sentiment_model.h5')

new_text = ["A lot of good things are happening. We are respected again throughout the world, and that's a great thing.@realDonaldTrump"]
new_text = tokenizer.texts_to_sequences(new_text)
new_text = pad_sequences(new_text, maxlen=X.shape[1], dtype='int32', value=0)
sentiment_prob = loaded_model.predict(new_text, batch_size=1, verbose=2)[0]

sentiment_classes = ['Negative', 'Neutral', 'Positive']
sentiment_pred = sentiment_classes[np.argmax(sentiment_prob)]

print("Predicted sentiment: ", sentiment_pred)
print("Predicted probabilities: ", sentiment_prob)
```

```
from keras.models import load_model import numpy as np

loaded_model = load_model('sentiment_model.h5')

new_text = ["A lot of good things are happening. We are respected again throughout the world, and that's a great thing.@realDonaldTrump"]

new_text = tokenizer.texts_to_sequences(new_text)

new_text = pad_sequences(new_text, maxlen-X.shape[1], dtype='int32', value=0)

sentiment_prob = loaded_model.predict(new_text, batch_size=1, verbose=2)[0]

sentiment_classes = ['Negative', 'Neutral', 'Positive']

sentiment_prob = sentiment_classes(np.argmax(sentiment_prob))

print("Predicted sentiment: ", sentiment_prob)

MARNING:tensorflow:Layer lstm_1 will not use cuDNN kernels since it doesn't meet the criteria. It will use a generic GPU kernel as fallback when running on GPU.

1/1 - 0s - 428ms/epoch - 428ms/step

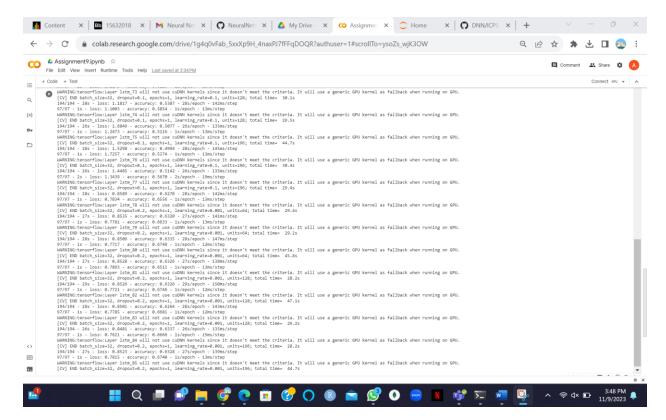
Predicted sentiment: Positive

Predicted sentiment: Positive

Predicted probabilities: [0.39611092 0.19630554 0.4075835]
```

■ Q ■ ● ■ ● ■ ● ■ ● ■ ■ ■ ■ ■ ● ■ ■ ■ ■ ●

```
from keras.wrappers.scikit learn import KerasClassifier
from sklearn.model selection import GridSearchCV
from keras.optimizers import Adam
def create model(units=196, dropout=0.2, learning rate=0.001):
    model = Sequential()
    model.add(Embedding(max fatures, embed dim,input length = X.shape[1]))
    model.add(LSTM(units, dropout=dropout, recurrent dropout=dropout))
    model.add(Dense(3, activation='softmax'))
    optimizer = Adam(lr=learning rate)
    model.compile(loss='categorical crossentropy', optimizer=optimizer,
metrics=['accuracy'])
    return model
model = KerasClassifier(build fn=create model, verbose=2)
units = [64, 128, 196]
dropout = [0.1, 0.2, 0.3]
learning rate = [0.001, 0.01, 0.1]
epochs = [1]
batch size = [32]
param grid = dict(units=units, dropout=dropout,
learning rate=learning rate, epochs=epochs, batch size=batch size)
grid = GridSearchCV(estimator=model, param grid=param grid, cv=3,
verbose=2)
grid result = grid.fit(X train, Y train)
print("Best: %f using %s" % (grid result.best score ,
grid result.best params ))
```



This code defines the create_model function that returns a Keras model with the specified hyperparameters. The KerasClassifier class is used to create a wrapper for the create_model function, which can be used as an estimator for GridSearchCV. The hyperparameters to be tuned are defined in the param_grid dictionary. GridSearchCV is then called with the KerasClassifier object, the param_grid dictionary