```
from keras preprocessing.sequence import pad sequences
import numpy as np
import pandas as pd
from gensim.models.word2vec import Word2Vec
from sklearn.model selection import train test split
from keras.utils import to categorical
from keras.layers import Dense, Dropout, Conv1D, MaxPool1D, GlobalMaxPool1D, Embedding, Activ
from keras.preprocessing.text import Tokenizer
from keras.models import Sequential
import re
import nltk
from nltk.corpus import stopwords
from nltk.stem.snowball import PorterStemmer
from sklearn import preprocessing
import matplotlib.pyplot as plt
import seaborn as sns
df·=·pd.read_csv('data.csv',delimiter=',',encoding='latin-1')
df·=·df[['Category','Message']]
df -= · df[pd.notnull(df['Message'])]
df.rename(columns⋅=⋅{'Message':'Message'}, ⋅inplace⋅=⋅True)
df.head()
```

Message	Category	(
Go until jurong point, crazy Available only	ham	0
Ok lar Joking wif u oni	ham	1
Free entry in 2 a wkly comp to win FA Cup fina	spam	2
U dun say so early hor U c already then say	ham	3
Nah I don't think he goes to usf, he lives aro	ham	4

```
df.shape
```

(5572, 2)

```
cnt_pro = df['Category'].value_counts()
plt.figure(figsize=(12,4))
sns.barplot(cnt_pro.index, cnt_pro.values, alpha=0.8)
plt.ylabel('Number of Occurrences', fontsize=12)
plt.xlabel('Category', fontsize=12)
plt.xticks(rotation=90)
plt.show();
```

/usr/local/lib/python3.8/dist-packages/seaborn/_decorators.py:36: FutureWarning: Pass the warnings.warn(

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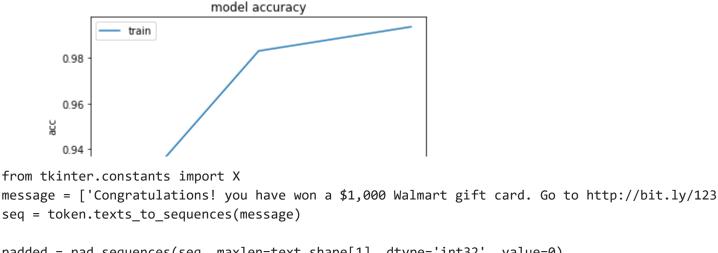
```
def preprocess_text(sen):
   # Remove punctuations and numbers
   sentence = re.sub('[^a-zA-Z]', ' ', sen)
   # Single character removal
   sentence = re.sub(r"\s+[a-zA-Z]\s+", ' ', sentence)
   # Removing multiple spaces
   sentence = re.sub(r'\s+', ' ', sentence)
   stops = stopwords.words('english')
   #print(stops)
   porter = PorterStemmer()
   for word in sentence.split():
        if word in stops:
            sentence = sentence.replace(word, '')
        sentence = sentence.replace(word, porter.stem(word))
   return sentence.lower()
mes = []
for i in df['Message']:
   mes.append(i.split())
print(mes[:2])
     [['Go', 'until', 'jurong', 'point,', 'crazy..', 'Available', 'only', 'in', 'bugis', 'n',
word2vec_model = Word2Vec(mes, size=500, window=3, min_count=1, workers=16)
print(word2vec_model)
     WARNING:gensim.models.base_any2vec:under 10 jobs per worker: consider setting a smaller
```

Word2Vec(vocab=15686, size=500, alpha=0.025)

```
token = Tokenizer(7229)
token.fit on texts(df['Message'])
text = token.texts to sequences(df['Message'])
text = pad_sequences(text, 75)
print(text[:2])
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le = preprocessing.LabelEncoder()
y = le.fit transform(df['Category'])
y = to categorical(y)
y[:2]
     array([[1., 0.],
            [1., 0.]], dtype=float32)
x_train, x_test, y_train, y_test = train_test_split(np.array(text), y, test_size=0.2, stratif
keras model = Sequential()
keras_model.add(word2vec_model.wv.get_keras_embedding(True))
keras model.add(Dropout(0.2))
keras_model.add(Conv1D(50, 3, activation='relu', padding='same', strides=1))
keras_model.add(Conv1D(50, 3, activation='relu', padding='same', strides=1))
keras model.add(MaxPool1D())
keras model.add(Dropout(0.2))
keras model.add(Conv1D(100, 3, activation='relu', padding='same', strides=1))
keras_model.add(Conv1D(100, 3, activation='relu', padding='same', strides=1))
keras model.add(MaxPool1D())
keras model.add(Dropout(0.2))
keras_model.add(Conv1D(200, 3, activation='relu', padding='same', strides=1))
keras model.add(Conv1D(200, 3, activation='relu', padding='same', strides=1))
keras model.add(GlobalMaxPool1D())
keras model.add(Dropout(0.2))
keras model.add(Dense(200))
keras model.add(Activation('relu'))
keras model.add(Dropout(0.2))
keras model.add(Dense(2))
keras model.add(Activation('softmax'))
```

```
Text_Classification_CNN_with_Embeddings.ipynb - Colaboratory
keras model.compile(loss='binary crossentropy', metrics=['acc'], optimizer='adam')
history = keras model.fit(x train, y train, batch size=16, epochs=3, validation data=(x test,
    Epoch 1/3
    279/279 [============ ] - 20s 70ms/step - loss: 0.2702 - acc: 0.9069 -
    Epoch 2/3
    279/279 [============ ] - 19s 69ms/step - loss: 0.0713 - acc: 0.9829 -
    Epoch 3/3
    279/279 [============ ] - 21s 74ms/step - loss: 0.0253 - acc: 0.9935 -
```

```
plt.plot(history.history['acc'])
plt.title('model accuracy')
plt.ylabel('acc')
plt.xlabel('epochs')
plt.legend(['train', 'test'], loc='upper left')
plt.show()
plt.savefig('model accuracy.png')
# summarize history for loss
plt.plot(history.history['loss'])
#plt.plot(history.history['val_loss'])
plt.title('model loss')
plt.ylabel('loss')
plt.xlabel('epochs')
plt.legend(['train', 'test'], loc='upper left')
plt.show()
plt.savefig('model_loss.png')
```



Analysis:

Throughtout my process of evaulating three different neural network models, it has come to my attention that with embeddings added, it's pretty hard to fit the model without it.

Sequential model as itself without CNN or RNN or even any other types of Neural Networks seems to give out very good accuracy but when testing the values to check if the following sentence is spam or not. It fails to give the right result.

All the other models that I tested both CNN and RNN gave very good accuracy with the right results as intended.