```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import LabelEncoder
from keras.models import Model
from keras.layers import LSTM, Activation, Dense, Dropout, Input, Embedding
from keras.optimizers import RMSprop
from keras.preprocessing.text import Tokenizer
from keras.preprocessing import sequence
from keras.utils import to categorical
from keras.callbacks import EarlyStopping
%matplotlib inline
from tensorflow.keras.preprocessing.sequence import pad_sequences
df = pd.read_csv('spam.csv',delimiter=',',encoding='latin-1')
df.drop(['Unnamed: 2', 'Unnamed: 3', 'Unnamed: 4'],axis=1,inplace=True)
df.info()
    <class 'pandas.core.frame.DataFrame'>
     RangeIndex: 5572 entries, 0 to 5571
    Data columns (total 2 columns):
         Column Non-Null Count Dtype
                 -----
     0
         v1
                 5572 non-null
                                 object
     1
         v2
                 5572 non-null
                                 object
    dtypes: object(2)
    memory usage: 87.2+ KB
sns.countplot(df.v1)
plt.xlabel('Label')
plt.title('Number of ham and spam messages')
```

```
/usr/local/lib/python3.7/dist-packages/seaborn/_decorators.py:43: FutureWarning: Pass th
       FutureWarning
    Text(0.5, 1.0, 'Number of ham and spam messages')
                  Number of ham and spam messages
       5000
       4000
X = df.v2
Y = df.v1
le = LabelEncoder()
Y = le.fit_transform(Y)
Y = Y.reshape(-1,1)
       1000 |
X_train,X_test,Y_train,Y_test = train_test_split(X,Y,test_size=0.15)
                     ham
                                         spam
\max \text{ words} = 1000
max len = 150
tok = Tokenizer(num_words=max_words)
tok.fit_on_texts(X_train)
sequences = tok.texts_to_sequences(X_train)
def RNN():
    inputs = Input(name='inputs',shape=[max len])
   layer = Embedding(max words,50,input length=max len)(inputs)
   layer = LSTM(64)(layer)
   layer = Dense(256,name='FC1')(layer)
   layer = Activation('relu')(layer)
   layer = Dropout(0.5)(layer)
   layer = Dense(1,name='out layer')(layer)
   layer = Activation('sigmoid')(layer)
   model = Model(inputs=inputs,outputs=layer)
    return model
model = RNN()
model.summary()
model.compile(loss='binary_crossentropy',optimizer=RMSprop(),metrics=['accuracy'])
    Model: "model"
      Layer (type)
                                 Output Shape
                                                          Param #
     ______
      inputs (InputLayer)
                                 [(None, 150)]
      embedding (Embedding)
                                 (None, 150, 50)
                                                          50000
      1stm (LSTM)
                                 (None, 64)
                                                          29440
```

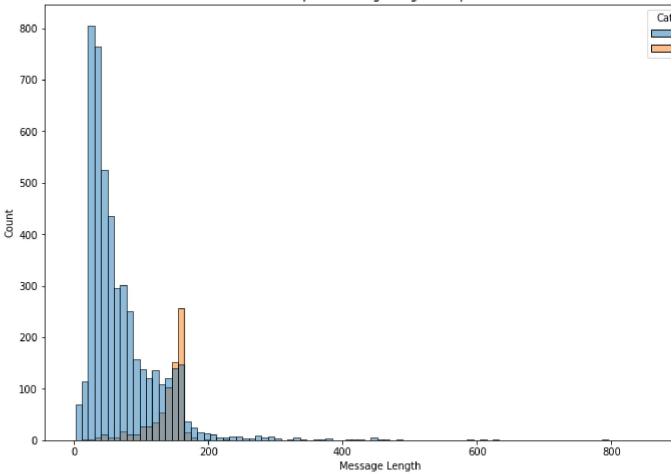
(None, 256)

16640

FC1 (Dense)

```
activation (Activation)
                               (None, 256)
     dropout (Dropout)
                               (None, 256)
                                                       0
     out_layer (Dense)
                               (None, 1)
                                                       257
     activation_1 (Activation)
                               (None, 1)
    ______
    Total params: 96,337
    Trainable params: 96,337
    Non-trainable params: 0
df.columns
    Index(['v1', 'v2', 'Count'], dtype='object')
data=df.rename(
   "v1": "Category",
   "v2":"Message"
},
   axis=1
)
df.info()
    <class 'pandas.core.frame.DataFrame'>
    Int64Index: 5572 entries, 1211 to 3623
    Data columns (total 3 columns):
     # Column Non-Null Count Dtype
              5572 non-null object
       v1
     0
     1 v2
               5572 non-null object
         Count 5572 non-null
                               int64
    dtypes: int64(1), object(2)
    memory usage: 174.1+ KB
data["Message Length"]=data["Message"].apply(len)
fig=plt.figure(figsize=(12,8))
sns.histplot(
   x=data["Message Length"],
   hue=data["Category"]
plt.title("ham & spam messege length comparision")
plt.show()
```

0



```
ham_desc=data[data["Category"]=="ham"]["Message Length"].describe()
spam_desc=data[data["Category"]=="spam"]["Message Length"].describe()
```

```
print("Ham Messege Length Description:\n",ham_desc)
print("******************************
print("Spam Message Length Description:\n",spam_desc)
```

Ham Messege Length Description:

	-
count	4825.000000
mean	71.023627
std	58.016023
min	2.000000
25%	33.000000
50%	52.000000
75%	92.000000
max	910.000000

Spam Message Length Description:

```
count 747.000000
mean 138.866131
std 29.183082
min 13.000000
25% 132.500000
50% 149.000000
```

75% 157.000000 max 224.00000

Name: Message Length, dtype: float64

## data.describe(include="all")

	Category	Message	Count	Message Length
count	5572	5572	5572.0	5572.000000
unique	2	5169	NaN	NaN
top	ham	Sorry, I'll call later	NaN	NaN
freq	4825	30	NaN	NaN
mean	NaN	NaN	0.0	80.118808
std	NaN	NaN	0.0	59.690841
min	NaN	NaN	0.0	2.000000
25%	NaN	NaN	0.0	36.000000
50%	NaN	NaN	0.0	61.000000
75%	NaN	NaN	0.0	121.000000
max	NaN	NaN	0.0	910.000000

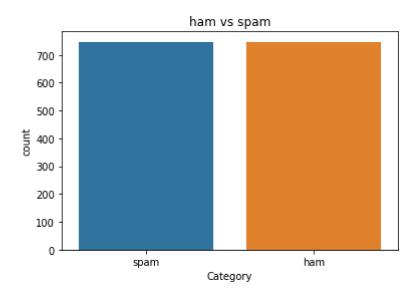
```
data["Category"].value_counts()

    ham          4825
    spam         747
    Name: Category, dtype: int64

sns.countplot(
    data=data,
    x="Category"
)
plt.title("ham vs spam")
plt.show()
```

```
5000
        4000
        3000
ham_count=data["Category"].value_counts()[0]
spam count=data["Category"].value counts()[1]
total_count=data.shape[0]
print("Ham contains:{:.2f}% of total data.".format(ham_count/total_count*100))
print("Spam contains:{:.2f}% of total data.".format(spam_count/total_count*100))
     Ham contains:86.59% of total data.
     Spam contains:13.41% of total data.
#compute the length of majority & minority class
minority_len=len(data[data["Category"]=="spam"])
majority_len=len(data[data["Category"]=="ham"])
#store the indices of majority and minority class
minority indices=data[data["Category"]=="spam"].index
majority_indices=data[data["Category"]=="ham"].index
#generate new majority indices from the total majority indices
#with size equal to minority class length so we obtain equivalent number of indices length
random majority indices=np.random.choice(
   majority indices,
   size=minority len,
   replace=False
)
#concatenate the two indices to obtain indices of new dataframe
undersampled_indices=np.concatenate([minority_indices,random_majority_indices])
#create df using new indices
df=data.loc[undersampled indices]
#shuffle the sample
df=df.sample(frac=1)
#reset the index as its all mixed
df=df.reset_index()
#drop the older index
df=df.drop(
   columns=["index"],
)
```

ham vs spam



df.head()

	Category	Message	Count	Message Length
0	spam	Eerie Nokia tones 4u, rply TONE TITLE to 8007	0	162
1	ham	That sucks. I'll go over so u can do my hair	0	70
2	ham	says that he's quitting at least5times a day	0	200
3	ham	Hey. For me there is no leave on friday. Wait	0	83
4	spam	Please call our customer service representativ	0	149

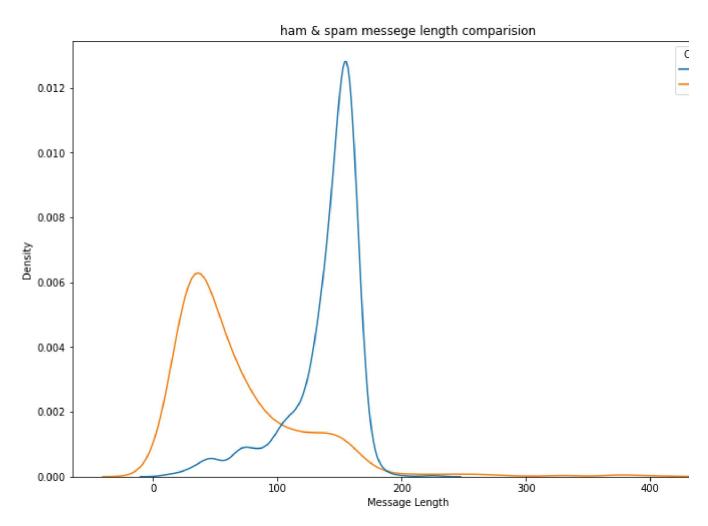
```
"ham":0,
"spam":1
}
)
```

df.head()

	Category	Message	Count	Message Length	Label
0	spam	Eerie Nokia tones 4u, rply TONE TITLE to 8007	0	162	1
1	ham	That sucks. I'll go over so u can do my hair	0	70	0
2	ham	says that he's quitting at least5times a day	0	200	0
3	ham	Hey. For me there is no leave on friday. Wait	0	83	0
4	spam	Please call our customer service representativ	0	149	1

```
import re
import nltk
from nltk.corpus import stopwords
from nltk.stem import PorterStemmer
stemmer=PorterStemmer()
#declare empty list to store tokenized message
corpus=[]
#iterate through the df["Message"]
for message in df["Message"]:
   #replace every special characters, numbers etc.. with whitespace of message
   #It will help retain only letter/alphabets
   message=re.sub("[^a-zA-Z]"," ",message)
   #convert every letters to its lowercase
   message=message.lower()
   #split the word into individual word list
   message=message.split()
from tensorflow.keras.preprocessing.text import one_hot
vocab_size=10000
oneHot_doc=[one_hot(words,n=vocab_size)
```

```
for words in corpus
df["Message Length"].describe()
     count
              1494.000000
     mean
               103.384873
     std
                55.635473
     min
                 2.000000
     25%
                48.000000
     50%
               115.000000
     75%
               152.750000
               408.000000
     max
     Name: Message Length, dtype: float64
fig=plt.figure(figsize=(12,8))
sns.kdeplot(
    x=df["Message Length"],
    hue=df["Category"]
plt.title("ham & spam messege length comparision")
plt.show()
```



from tensorflow.keras.preprocessing.sequence import pad\_sequences

```
sentence_len=200
embedded_doc=pad_sequences(
   oneHot_doc,
   maxlen=sentence_len,
   padding="pre"
)
extract_features=pd.DataFrame(
   data=embedded_doc
target=df["Label"]
df_final=pd.concat([extract_features,target],axis=1)
df_final.head()
           0
                 1
                       2
                                        5
                                                   7
                                                                       191
                                                                             192
                                                                                  193
                                                                                        194
                            3
                                  4
                                             6
                                                         8
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        NaN
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                               NaN
                                                            NaN
                                                                                       NaN
     5 rows × 201 columns
X=df_final.drop("Label",axis=1)
y=df_final["Label"]
from sklearn.model_selection import train_test_split
X_trainval,X_test,y_trainval,y_test=train_test_split(
   Χ,
   у,
   random state=42,
   test_size=0.15
)
X_train,X_val,y_train,y_val=train_test_split(
   X_trainval,
   y_trainval,
   random_state=42,
   test_size=0.15
)
```

```
model = RNN()
model.summary()
model.compile(loss='binary_crossentropy',optimizer=RMSprop(),metrics=['accuracy'])
```

Model: "model\_3"

Layer (type)	Output Shape	Param #
inputs (InputLayer)	[(None, 150)]	0
embedding_4 (Embedding)	(None, 150, 50)	50000
lstm_4 (LSTM)	(None, 64)	29440
FC1 (Dense)	(None, 256)	16640
<pre>activation_6 (Activation)</pre>	(None, 256)	0
dropout_3 (Dropout)	(None, 256)	0
out_layer (Dense)	(None, 1)	257
activation_7 (Activation)	(None, 1)	0

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Total params: 96,337 Trainable params: 96,337 Non-trainable params: 0

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