```
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
import tensorflow
%matplotlib inline
```

from tensorflow.keras.preprocessing.sequence import pad_sequences

```
df = pd.read_csv('spam.csv',delimiter=',',encoding='latin-1')
df.head()
```

	V1	V2	2	3	4
	0 ham	Go until jurong point, crazy Available only	NaN	NaN	NaN
,	1 ham	Ok lar Joking wif u oni	NaN	NaN	NaN
:	2 spam	Free entry in 2 a wkly comp to win FA Cup fina	NaN	NaN	NaN
		U dun sav so early hor U c already then			
<pre>df.drop(['Unnamed: 2', 'Unnamed: 3', 'Unnamed: 4'],axis=1,inplace=True) df.info()</pre>					

Unnamed:

Unnamed:

Unnamed:

dtypes: object(2)

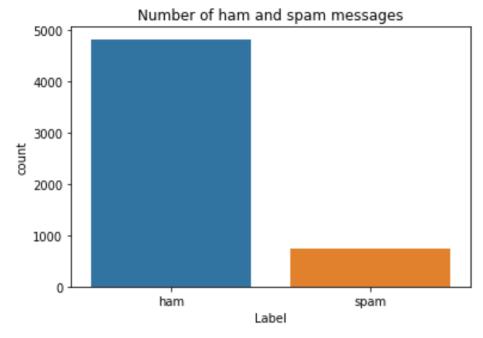
. . 4

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: FutureWarnin FutureWarning

Text(0.5, 1.0, 'Number of ham and spam messages')



```
X = df.v2
Y = df.v1
le = LabelEncoder()
Y = le.fit_transform(Y)
Y = Y.reshape(-1,1)

X_train,X_test,Y_train,Y_test = train_test_split(X,Y,test_size=0.15)

max_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_1"

Layer (type)	Output Shape	Param #
inputs (InputLayer)	[(None, 150)]	0
<pre>embedding_1 (Embedding)</pre>	(None, 150, 50)	50000
lstm_1 (LSTM)	(None, 64)	29440
FC1 (Dense)	(None, 256)	16640
activation_2 (Activation)	(None, 256)	0
dropout_1 (Dropout)	(None, 256)	0
out_layer (Dense)	(None, 1)	257
activation_3 (Activation)	(None, 1)	0
	=======================================	=========

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

df.columns

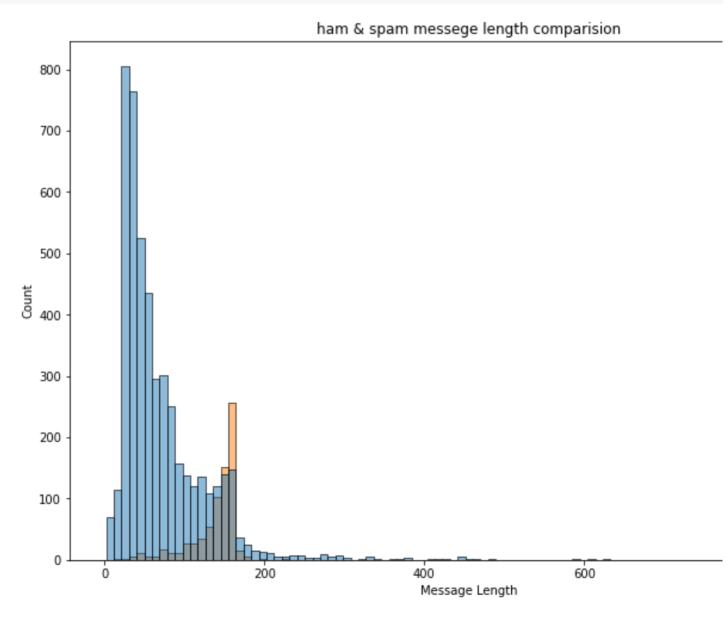
```
Index(['v1', 'v2'], dtype='object')
```

```
data=df.rename(
{
    "v1": "Category",
    "v2": "Message"
},
    axis=1
)
```

df.info()

```
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()
```



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
          71,023627
mean
std
          58.016023
min
           2.000000
25%
          33.000000
50%
          52.000000
75%
          92.000000
max
         910.000000
Name: Message Length, dtype: float64
***********
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
        224.000000
```

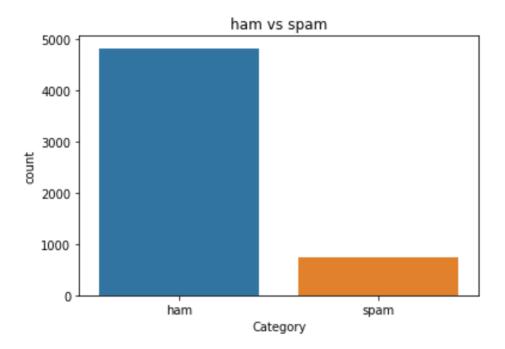
```
Name: Message Length, dtype: float64
```

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

```
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()
```



```
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.

```
minority_len=len(data[data["Category"]=="spam"])
majority_len=len(data[data["Category"]=="ham"])
minority_indices=data[data["Category"]=="spam"].index
majority_indices=data[data["Category"]=="ham"].index
random_majority_indices=np.random.choice(
    majority_indices,
```

```
size=minority_len,
    replace=False
)
undersampled_indices=np.concatenate([minority_indices,random_majority_indices])
df=data.loc[undersampled_indices]
df=df.sample(frac=1)

df=df.reset_index()
df=df.drop(
    columns=["index"],
)
df.shape
```

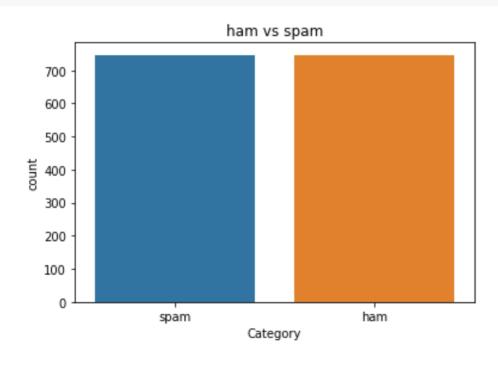
(1494, 3)

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

spam 747 ham 747

Name: Category, dtype: int64

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



	Category	Message	Message Length	7
0	spam	UR awarded a City Break and could WIN a å£200	148	
1	spam	No. 1 Nokia Tone 4 ur mob every week! Just txt	143	
2	ham	Don't worry, * is easy once have ingredients!	45	
3	ham	I just made some payments so dont have that mu	98	
4	spam	Moby Pub Quiz.Win a å£100 High Street prize if	156	
<pre>df["Labe { })</pre>	l"]=df["Ca "ham":0, "spam":1	tegory"].map(

df.head()

	Category	Message	Message Length	Labe
0	spam	UR awarded a City Break and could WIN a å£200	148	
1	spam	No. 1 Nokia Tone 4 ur mob every week! Just txt	143	
2	ham	Don't worry, * is easy once have ingredients!	45	
3	ham	I just made some payments so dont have that mu	98	
4	spam	Moby Pub Quiz.Win a å£100 High Street prize if	156	

```
import re
import nltk
from nltk.corpus import stopwords
from nltk.stem import PorterStemmer
stemmer=PorterStemmer()
```

```
corpus=[]
for message in df["Message"]:
    message=re.sub("[^a-zA-Z]"," ",message)
    message=message.lower()
    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
               105.866801
     mean
                56.509846
     std
     min
                 3.000000
     25%
                51.000000
     50%
               119.000000
     75%
               153.000000
```

```
max 450.000000
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()
```

```
0.012
from tensorflow.keras.preprocessing.sequence import pad sequences
sentence len=200
embedded_doc=pad_sequences(
    oneHot_doc,
    maxlen=sentence len,
    padding="pre"
)
      o 000€]
extract features=pd.DataFrame(
    data=embedded doc
)
target=df["Label"]
df_final=pd.concat([extract_features,target],axis=1)
df_final.head()
                 1
                                   4
                                                                        191
                                                                              192
            0
                       2
                             3
                                        5
                                              6
                                                    7
                                                          8
                                                               9
                                                                                    1
      0
         NaN
              NaN
                    NaN
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     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=37,
```

```
test_size=0.2
)
X_train,X_val,y_train,y_val=train_test_split(
    X_trainval,
    y_trainval,
    random_state=37,
    test_size=0.2
)
```

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

Model: "model_2"

Layer (type)	Output Shape	Param #
inputs (InputLayer)	[(None, 150)]	0
<pre>embedding_2 (Embedding)</pre>	(None, 150, 50)	50000
lstm_2 (LSTM)	(None, 64)	29440
FC1 (Dense)	(None, 256)	16640
activation_4 (Activation)	(None, 256)	0
dropout_2 (Dropout)	(None, 256)	0
out_layer (Dense)	(None, 1)	257
activation_5 (Activation)	(None, 1)	0
=======================================		========

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

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