

▼ Assignment-4 (SMS SPAM Classification)

```
1. Student Name: Poovarasan
2. Register Number: 814819106005
3. Team ID: PNT2022TMID46103
```

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```
import numpy as np
import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt
```

```
data = pd.read_csv('/content/sample_data/spam.csv',delimiter=',',encoding='latin-1')
data.head()
```



	v1	v2	Unnamed: 2	Unnamed: 3	Unnamed: 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
3	ham	U dun say so early hor... U c already then say...	NaN	NaN	NaN

```
data.columns
```

```
Index(['v1', 'v2', 'Unnamed: 2', 'Unnamed: 3', 'Unnamed: 4'], dtype='object')
```

```
#drop the unamed columns
```

```
data=data.drop(columns=["Unnamed: 2","Unnamed: 3","Unnamed: 4"])
```

```
#rename the two relevant columns
```

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

```
data.head()
```

	Category	Message
0	ham	Go until jurong point, crazy.. Available only ...
1	ham	Ok lar... Joking wif u oni...
2	spam	Free entry in 2 a wkly comp to win FA Cup fina...
3	ham	U dun say so early hor... U c already then say...

```
#check for null values
```

```
data.isnull().sum()
```

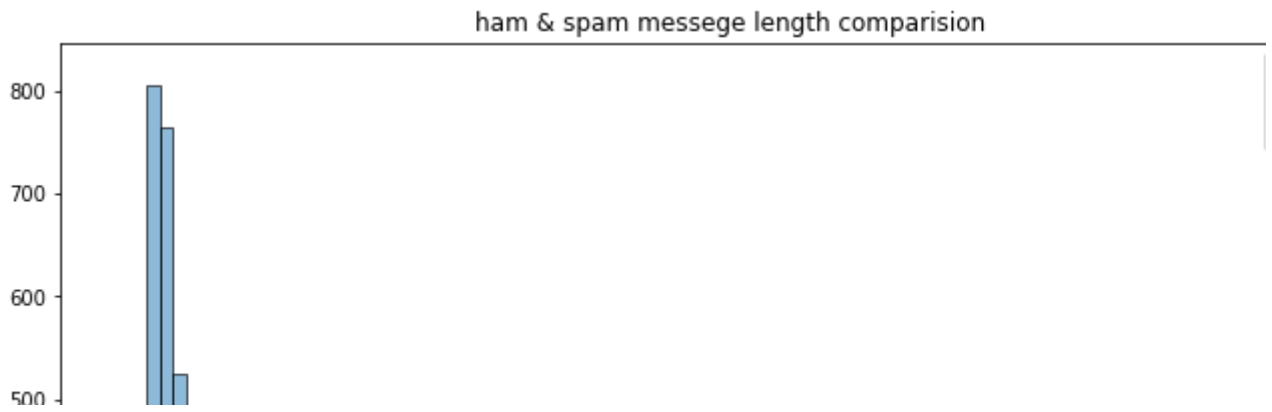
```
Category    0
Message     0
dtype: int64
```

```
data.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 5572 entries, 0 to 5571
Data columns (total 2 columns):
#   Column      Non-Null Count  Dtype
---  -
0   Category    5572 non-null   object
1   Message     5572 non-null   object
dtypes: object(2)
memory usage: 87.2+ 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()
```



#Display the description of length of ham and spam messages seperately on an individual se

```
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

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")
```

	Category	Message	Message Length
count	5572	5572	5572.000000
unique	2	5169	NaN
top	ham	Sorry, I'll call later	NaN
freq	4825	30	NaN

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

```

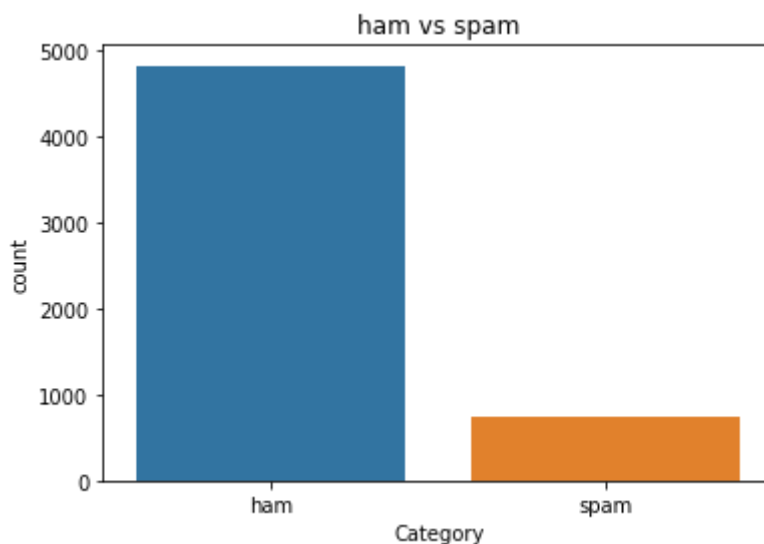
ham      4825
spam     747
Name: Category, dtype: int64
25%      NaN      NaN      36.000000

```

```

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.

```

```

#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"],
)

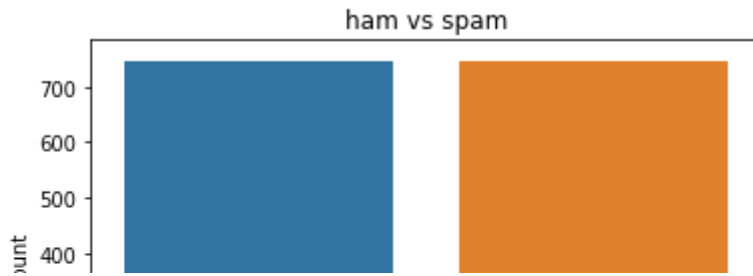
df.shape

(1494, 3)

df["Category"].value_counts()

ham      747
spam     747
Name: Category, dtype: int64

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



```
df.head()
```

	Category	Message	Message Length
0	ham	Aah! A cuddle would be lush! I'd need lots of ...	87
1	ham	I'm in solihull, do you want anything?	40
2	spam	Double Mins & 1000 txts on Orange tariffs. Lat...	151
3	ham	No we put party 7 days a week and study lightl...	126
4	spam	URGENT!! Your 4* Costa Del Sol Holiday or å£50...	161

```
#Created new column Label and encode ham as 0 and spam as 1
```

```
df["Label"]=df["Category"].map(
    {
        "ham":0,
        "spam":1
    }
)
```

```
df.head()
```

	Category	Message	Message Length	Label
0	ham	Aah! A cuddle would be lush! I'd need lots of ...	87	0
1	ham	I'm in solihull, do you want anything?	40	0
2	spam	Double Mins & 1000 txts on Orange tariffs. Lat...	151	1
3	ham	No we put party 7 days a week and study lightl...	126	0
4	spam	URGENT!! Your 4* Costa Del Sol Holiday or å£50...	161	1

```
import re
import nltk
from nltk.corpus import stopwords
from nltk.stem import PorterStemmer
```

```
stemmer=PorterStemmer()
```

```
nltk.download('stopwords')
```

```
[nltk_data] Downloading package stopwords to /root/nltk_data...
```

```

[nltk_data]   Unzipping corpora/stopwords.zip.
True

#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()

    #perform stemming using PorterStemmer for all non-english-stopwords
    message=[stemmer.stem(words)
              for words in message
              if words not in set(stopwords.words("english"))
             ]
    #join the word lists with the whitespace
    message=" ".join(message)

    #append the message in corpus list
    corpus.append(message)

from tensorflow.keras.preprocessing.text import one_hot
vocab_size=10000

oneHot_doc=[one_hot(words,n=vocab_size)
             for words in corpus
            ]

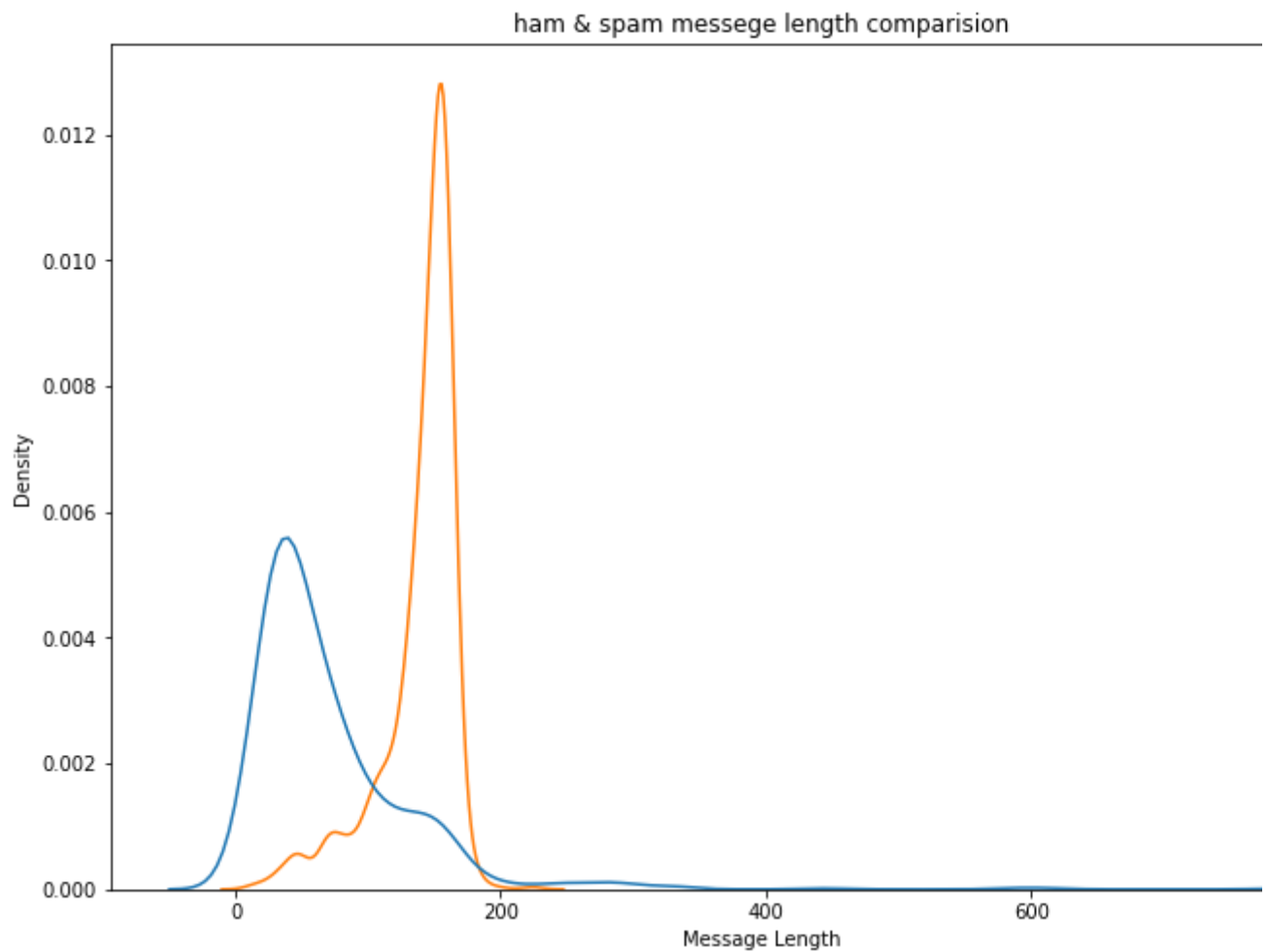
df["Message Length"].describe()

count      1494.000000
mean        105.203481
std         61.166448
min          3.000000
25%         48.000000
50%        118.000000
75%        153.000000
max         790.000000
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()
```



```
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	3	4	5	6	7	8	9	...	191	192	193	194	195	196	197	198
0	0	0	0	0	0	0	0	0	0	0	...	2090	1632	4289	7158	478	5808	6133	8348
1	0	0	0	0	0	0	0	0	0	0	...	0	0	0	0	0	0	8663	4425
2	0	0	0	0	0	0	0	0	0	0	...	1275	702	1694	4114	4162	3935	4162	8536
3	0	0	0	0	0	0	0	0	0	0	...	3705	9946	5462	7158	9883	4500	8030	8630

```
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(
    X,
    y,
    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
)
```

```
from tensorflow.keras.layers import LSTM
from tensorflow.keras.layers import Dense
from tensorflow.keras.layers import Embedding
from tensorflow.keras.models import Sequential
model=Sequential()
```

```
feature_num=100
model.add(
    Embedding(
        input_dim=vocab_size,
        output_dim=feature_num,
        input_length=sentence_len
    )
)
model.add(
    LSTM(
        units=128
    )
)
```

```
model.add(
    Dense(
        units=1,
```

```

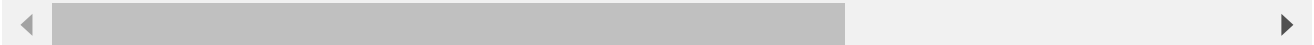
        activation="sigmoid"
    )
)

from tensorflow.keras.optimizers import Adam
model.compile(
    optimizer=Adam(
        learning_rate=0.001
    ),
    loss="binary_crossentropy",
    metrics=["accuracy"]
)

model.fit(
    X_train,
    y_train,
    validation_data=(
        X_val,
        y_val
    ),
    epochs=10
)

Epoch 1/10
34/34 [=====] - 8s 33ms/step - loss: 0.5258 - accuracy: 0.76
Epoch 2/10
34/34 [=====] - 1s 16ms/step - loss: 0.1718 - accuracy: 0.94
Epoch 3/10
34/34 [=====] - 1s 16ms/step - loss: 0.0533 - accuracy: 0.98
Epoch 4/10
34/34 [=====] - 1s 15ms/step - loss: 0.0254 - accuracy: 0.99
Epoch 5/10
34/34 [=====] - 1s 16ms/step - loss: 0.0184 - accuracy: 0.99
Epoch 6/10
34/34 [=====] - 1s 16ms/step - loss: 0.0134 - accuracy: 0.99
Epoch 7/10
34/34 [=====] - 1s 16ms/step - loss: 0.0150 - accuracy: 0.99
Epoch 8/10
34/34 [=====] - 1s 16ms/step - loss: 0.0112 - accuracy: 0.99
Epoch 9/10
34/34 [=====] - 1s 16ms/step - loss: 0.0062 - accuracy: 0.99
Epoch 10/10
34/34 [=====] - 1s 16ms/step - loss: 0.0050 - accuracy: 0.99
<keras.callbacks.History at 0x7fa3263a7850>

```



```

y_pred=model.predict(X_test)
y_pred=(y_pred>0.5)

```

```

8/8 [=====] - 0s 8ms/step

```

```

from sklearn.metrics import accuracy_score,confusion_matrix

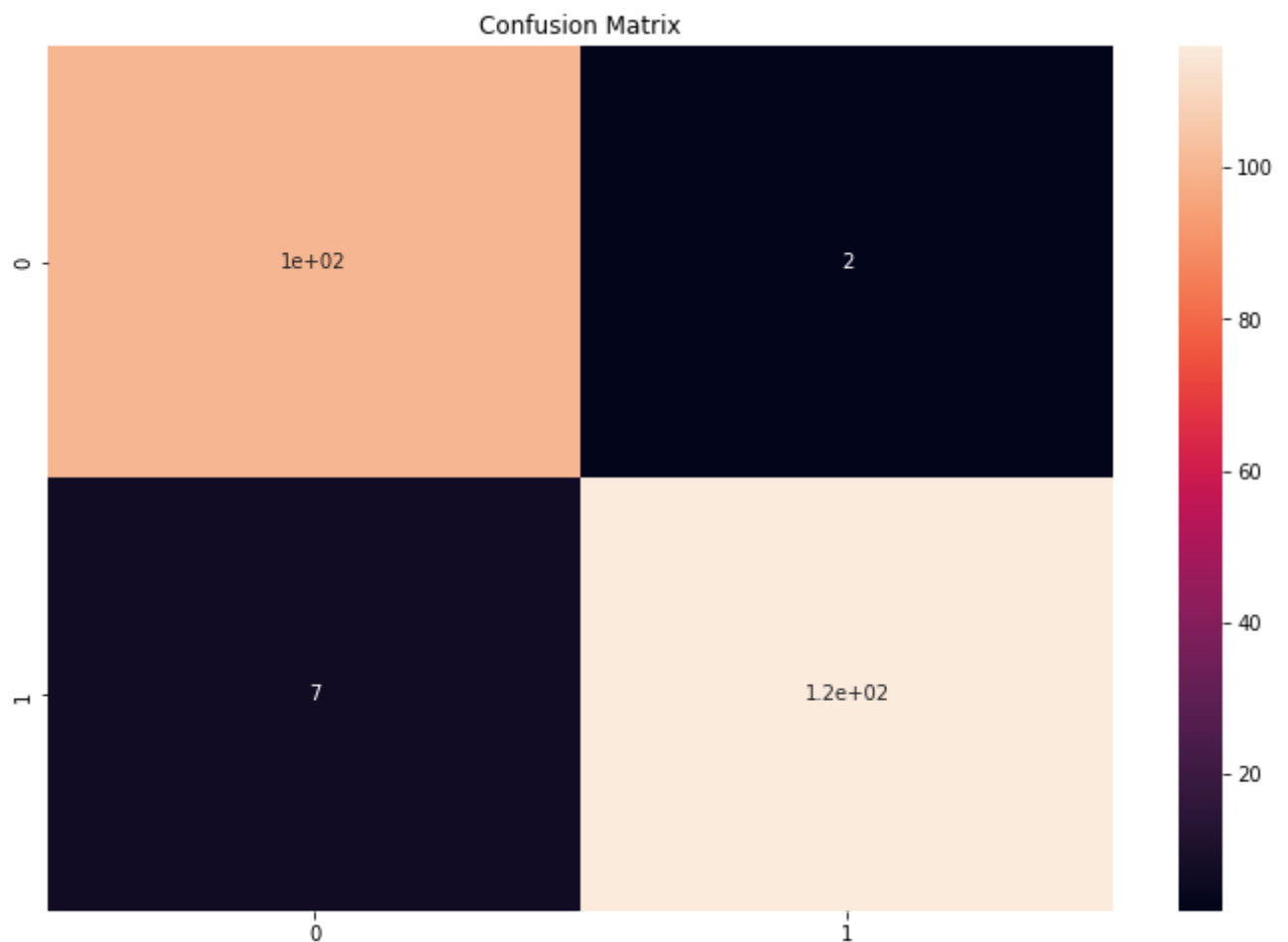
```

```
score=accuracy_score(y_test,y_pred)
print("Test Score:{:.2f}%".format(score*100))
```

Test Score:96.00%

```
cm=confusion_matrix(y_test,y_pred)
fig=plt.figure(figsize=(12,8))
sns.heatmap(
    cm,
    annot=True,
)
plt.title("Confusion Matrix")
cm
```

```
array([[100,  2],
       [ 7, 116]])
```



#The function take model and message as parameter

```
def classify_message(model,message):
```

```
    #We will treat message as a paragraphs containing multiple sentences(lines)
```

```
    #we will extract individual lines
```

```
    for sentences in message:
```

```
        sentences=nlTK.sent_tokenize(message)
```

```
    #Iterate over individual sentences
```

```
    for sentence in sentences:
```

```
        #replace all special characters
```

```
words=re.sub("[^a-zA-Z]", " ",sentence)

#perform word tokenization of all non-english-stopwords
if words not in set(stopwords.words('english')):
    word=nlk.word_tokenize(words)
    word=" ".join(word)
```

```
#perform one_hot on tokenized word
oneHot=[one_hot(word,n=vocab_size)]
```

```
#create an embedded documnet using pad_sequences
#this can be fed to our model
text=pad_sequences(oneHot,maxlen=sentence_len,padding="pre")
```

```
#predict the text using model
predict=model.predict(text)
```

```
#if predict value is greater than 0.5 its a spam
if predict>0.5:
    print("It is a spam")
#else the message is not a spam
else:
    print("It is not a spam")
```

```
message1="I am having a bad day and I would like to have a break today"
```

```
message2="This is to inform you had won a lottery and the subscription will end in a week"
```

```
nlk.download('punkt')
```

```
[nlk_data] Downloading package punkt to /root/nltk_data...
[nltk_data]   Unzipping tokenizers/punkt.zip.
True
```

```
classify_message(model,message1)
```

```
1/1 [=====] - 0s 21ms/step
It is not a spam
```

```
classify_message(model,message2)
```

```
1/1 [=====] - 0s 22ms/step
It is a spam
```

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