November 20, 2022

```
[]: 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')
                   ______
           FileNotFoundError
                                                   Traceback (most recent call,
     →last)
           <ipython-input-5-b43e29bce5ab> in <module>
       ----> 1 df = pd.read_csv('spam.csv',delimiter=',',encoding='latin-1')
             2 df.head()
           /usr/local/lib/python3.7/dist-packages/pandas/util/_decorators.py in_
     →wrapper(*args, **kwargs)
           309
                                  stacklevel=stacklevel,
           310
        --> 311
                          return func(*args, **kwargs)
           312
           313
                      return wrapper
```

```
/usr/local/lib/python3.7/dist-packages/pandas/io/parsers/readers.py in_
→read_csv(filepath_or_buffer, sep, delimiter, header, names, index_col,_u
→usecols, squeeze, prefix, mangle dupe cols, dtype, engine, converters,

→true_values, false_values, skipinitialspace, skiprows, skipfooter, nrows,

□

→na values, keep default na, na filter, verbose, skip blank lines, parse dates,
→infer_datetime_format, keep_date_col, date_parser, dayfirst, cache_dates, __
→iterator, chunksize, compression, thousands, decimal, lineterminator,
→quotechar, quoting, doublequote, escapechar, comment, encoding, ___
→encoding errors, dialect, error bad lines, warn bad lines, on bad lines,
→delim_whitespace, low_memory, memory_map, float_precision, storage_options)
               kwds.update(kwds_defaults)
       585
   --> 586
              return read(filepath or buffer, kwds)
       587
       588
       /usr/local/lib/python3.7/dist-packages/pandas/io/parsers/readers.py in u
→_read(filepath_or_buffer, kwds)
       480
       481
               # Create the parser.
   --> 482
               parser = TextFileReader(filepath or buffer, **kwds)
       483
       484
               if chunksize or iterator:
       /usr/local/lib/python3.7/dist-packages/pandas/io/parsers/readers.py in u
→ init (self, f, engine, **kwds)
       809
                       self.options["has_index_names"] = kwds["has_index_names"]
       810
   --> 811
                   self._engine = self._make_engine(self.engine)
       812
       813
               def close(self):
       /usr/local/lib/python3.7/dist-packages/pandas/io/parsers/readers.py in u
→_make_engine(self, engine)
      1038
      1039
                   # error: Too many arguments for "ParserBase"
                   return mapping[engine](self.f, **self.options) # type:
   -> 1040
→ignore[call-arg]
      1041
      1042
               def _failover_to_python(self):
```

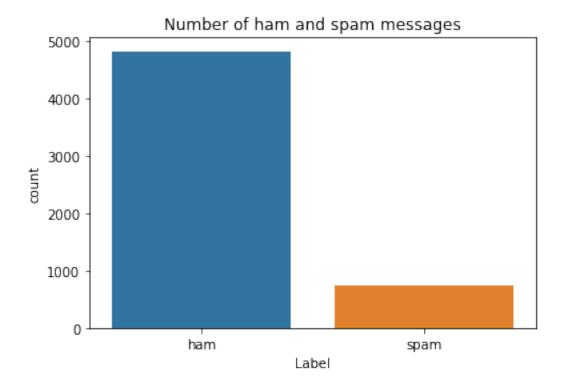
```
→c_parser_wrapper.py in __init__(self, src, **kwds)
             49
             50
                        # open handles
        ---> 51
                        self._open_handles(src, kwds)
             52
                        assert self.handles is not None
             53
            /usr/local/lib/python3.7/dist-packages/pandas/io/parsers/base_parser.py_
     →in _open_handles(self, src, kwds)
                            memory_map=kwds.get("memory_map", False),
            227
            228
                            storage_options=kwds.get("storage_options", None),
        --> 229
                            errors=kwds.get("encoding_errors", "strict"),
                        )
            230
            231
            /usr/local/lib/python3.7/dist-packages/pandas/io/common.py in_
     →get_handle(path_or_buf, mode, encoding, compression, memory_map, is_text, __
     →errors, storage_options)
            705
                                encoding=ioargs.encoding,
            706
                                errors=errors,
        --> 707
                                newline="",
            708
            709
                        else:
            FileNotFoundError: [Errno 2] No such file or directory: 'spam.csv'
[]: 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/pandas/io/parsers/

/usr/local/lib/python3.7/dist-packages/seaborn/_decorators.py:43: FutureWarning: Pass the following variable as a keyword arg: x. From version 0.12, the only valid positional argument will be `data`, and passing other arguments without an explicit keyword will result in an error or misinterpretation.

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)

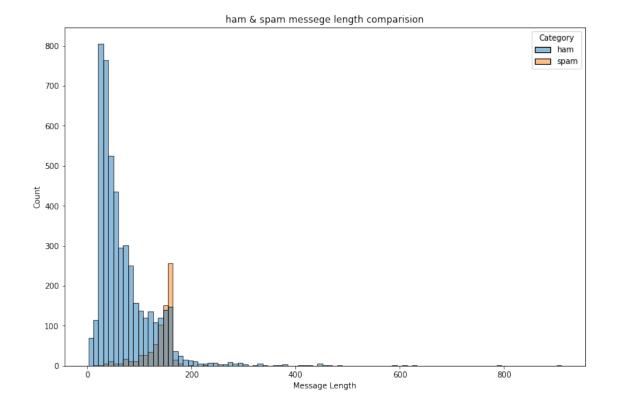
[]: X_train,X_test,Y_train,Y_test = train_test_split(X,Y,test_size=0.15)

[]: 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"
     Layer (type)
                                 Output Shape
                                                           Param #
     inputs (InputLayer)
                                 [(None, 150)]
     embedding (Embedding)
                                 (None, 150, 50)
                                                            50000
     1stm (LSTM)
                                 (None, 64)
                                                            29440
     FC1 (Dense)
                                 (None, 256)
                                                           16640
     activation (Activation)
                                 (None, 256)
     dropout (Dropout)
                                 (None, 256)
     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
        v2
                5572 non-null
                                object
     1
                                int64
        Count
                5572 non-null
    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()
```



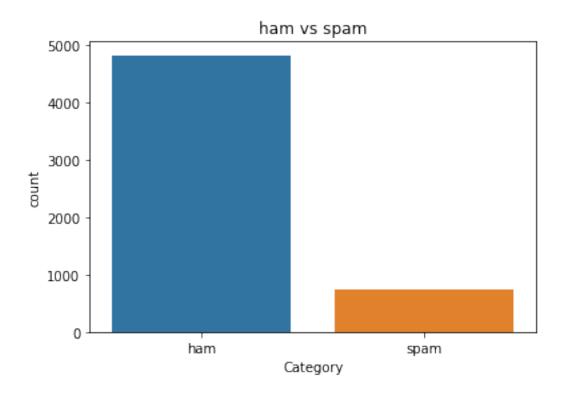
Ham Messege Length Description:

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
             224.000000
    Name: Message Length, dtype: float64
[]: data.describe(include="all")
[]:
            Category
                                      Message
                                                 Count
                                                        Message Length
                5572
                                          5572
                                                5572.0
                                                            5572.000000
     count
     unique
                                          5169
                                                   NaN
                                                                    NaN
     top
                 ham
                      Sorry, I'll call later
                                                   NaN
                                                                    NaN
                4825
                                                   NaN
    freq
                                            30
                                                                    NaN
    mean
                 NaN
                                           NaN
                                                   0.0
                                                             80.118808
     std
                 NaN
                                           NaN
                                                   0.0
                                                              59.690841
                 NaN
                                                   0.0
                                                               2.000000
    min
                                           NaN
     25%
                                                   0.0
                 NaN
                                           NaN
                                                              36.000000
     50%
                                                   0.0
                 NaN
                                           NaN
                                                              61.000000
     75%
                 NaN
                                           NaN
                                                   0.0
                                                             121.000000
                                                   0.0
                                                             910.000000
     max
                 NaN
                                           NaN
[]: data["Category"].value_counts()
[ ]: ham
             4825
              747
     spam
     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))
```

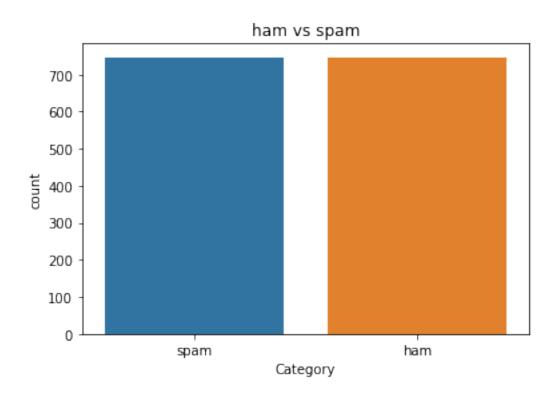
Ham contains:86.59% of total data. Spam contains:13.41% of total data.

```
[]: #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(
```

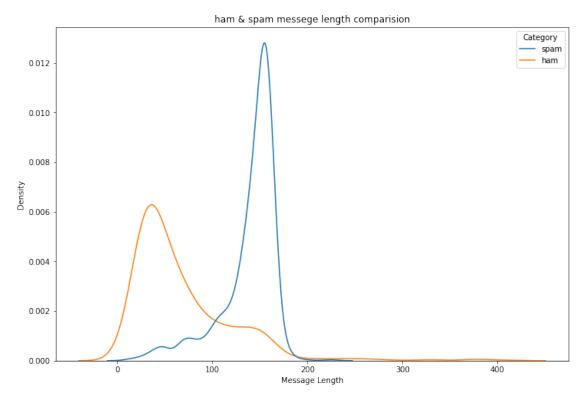
```
majority_indices,
         size=minority_len,
         replace=False
     #concatenate the two indices to obtain indices of new dataframe
     undersampled_indices=np.concatenate([minority_indices,random_majority_indices])
     #create df using new indices
     df=data.loc[undersampled_indices]
     #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, 4)
[]: df["Category"].value_counts()
[]: spam
             747
             747
    ham
    Name: Category, dtype: int64
[]: sns.countplot(
         data=df,
         x="Category"
     plt.title("ham vs spam")
     plt.show()
```



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

```
[]: df.head()
[]:
                                                            Message Count \
       Category
                 Eerie Nokia tones 4u, rply TONE TITLE to 8007 \dots
     0
           spam
                                                                        0
     1
                 That sucks. I'll go over so u can do my hair. ...
                                                                        0
     2
                  says that he's quitting at least5times a day ...
                                                                        0
            ham
     3
            ham Hey. For me there is no leave on friday. Wait ...
                                                                        0
                 Please call our customer service representativ...
        Message Length Label
     0
                   162
                            1
                    70
     1
                            0
     2
                            0
                   200
     3
                    83
                            0
     4
                   149
[]: 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()
[]:
              1
                  2
                      3
                           4
                               5
                                   6
                                        7
                                            8
                                                       191
                                                             192
                                                                  193
                                                                        194
                                                                             195
                                                                                  196
     O NaN NaN NaN NaN NaN NaN NaN NaN NaN
                                                       {\tt NaN}
                                                            {\tt NaN}
                                                                             {\tt NaN}
                                                                                  NaN
                                                                  {\tt NaN}
                                                                       {\tt NaN}
     1 NaN NaN NaN NaN NaN NaN NaN NaN NaN
                                                       \mathtt{NaN}
                                                            NaN
                                                                  NaN
                                                                        {\tt NaN}
                                                                             \mathtt{NaN}
                                                                                  NaN
     {\tt NaN}
                                                            {\tt NaN}
                                                                  {\tt NaN}
                                                                        {\tt NaN}
                                                                             {\tt NaN}
                                                                                  {\tt NaN}
     {\tt NaN}
                                                            NaN
                                                                  NaN
                                                                        NaN
                                                                             {\tt NaN}
                                                                                  NaN
     4 Nan Nan Nan Nan Nan Nan Nan Nan Nan
                                                       {\tt NaN}
                                                            {\tt NaN}
                                                                  {\tt NaN}
                                                                       {\tt NaN}
                                                                             {\tt NaN}
                                                                                  {\tt NaN}
        197
             198
                   199
                        Label
     0 NaN
             {\tt NaN}
                   NaN
     1 NaN
                             0
             \mathtt{NaN}
                   NaN
                             0
     2 NaN
             NaN
                   {\tt NaN}
     3 NaN
             NaN
                   NaN
     4 NaN
             {\tt NaN}
                  NaN
     [5 rows x 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(
         Х,
         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
activation_6 (Activation)	(None, 256)	0
dropout_3 (Dropout)	(None, 256)	0
<pre>out_layer (Dense)</pre>	(None, 1)	257
activation_7 (Activation)	(None, 1)	0

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