

# DATA ANALYSIS - Spam Classification DATASET

# **ROADMAP**



# **INITIAL DATA CLEANING**

### Converted to ISO-8859-1 as UTF 8 was creating issues

	v1	v2	Unnamed: 2	Unnamed: 3	Unnamed: 4	
0	ham	Go until jurong point, crazy Available only	NaN	NaN	NaN	
	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	
4	ham	Nah I don't think he goes to usf, he lives aro	NaN	NaN	NaN	

# **INITIAL DATA CLEANING**

```
In [15]: # Dataset has extra columns- Remove
          #Renaming v1 and v2
         df.drop (columns=['Unnamed: 2', 'Unnamed: 3', 'Unnamed: 4'], axis=1, inplace=True)
         df.rename(columns={'v1': "label", 'v2': 'message'}, inplace=True)
In [16]: df.head()
         print ('Shape >', df.shape)
         Shape > (5572, 2)
In [17]: print ('ham and spam counts', '\n', df.label.value_counts())
         ham and spam counts
                  4825
                  747
         spam
         Name: label, dtype: int64
In [18]: #Ratio
         print ('spam ratio = ', round (len(df[df['label'] == 'spam'])/ len(df.label), 2)*100, '%')
         print ('ham ratio = ', round(len (df[df['label'] == 'ham']) / len(df.label), 2)*100, '%')
          spam ratio = 13.0 %
         ham ratio = 87.0 %
```

# **INITIAL DATA CLEANING**

Used label encoding to bifurcate into ham and spam messages. Also converted the messages to lower case.

```
In [20]: #Label coding 0 and 1
df['label' ].replace({'ham':0,'spam':1},inplace=True)

In [21]: # Convert all messages to Lower case
df['message'] = df['message'].str. lower()
df.head()

Out[21]: label message length

0 0 go until jurong point, crazy.. available only ... 111

1 0 ok lar... joking wif u oni... 29
2 1 free entry in 2 a wkly comp to win fa cup fina... 155
3 0 u dun say so early hor... u c already then say... 49
4 0 nah i don't think he goes to usf, he lives aro... 61
```

## **DATA CLEANING & WRANGLING**

Replaced the url in the messages with -'webaddress'

### Replaced numbers with 'numbr'

```
In [22]: # Replace email addresses with 'email'
         df['message'] = df['message'].str.replace(r'^.+@[^\.].*\.[a-z]{2,}$','emailaddress')
In [23]: #Replace URLS with 'webaddress' I
         df['message'] = df['message'].str.replace(r'^http\://[a-zA-Z0-9\-\.]+\.[a-zA-Z]{2,3}(/\S*)?$',
          'webaddress')
In [24]: # Replace money symbols with 'moneysymb(£ can by typed with ALT key + 156)
         df['message'] = df['message'].str.replace(r'f|\$', 'dollers')
In [25]: #RepLace 10 digit phone numbers (formats include paranthesis, spaces, no spaces, dashes) with 'phonenumber"
         df['message'] = df['message'].str.replace(r'^\(?P[\d]{3}\)?[\s-]?[\d]{3}\\,s-]?[\d]{4}$',
          'phonenumber')
In [26]: #Replace numbers with 'numbr
         df['message'] = df['message'].str.replace(r'\d+(\.\d+)?', 'numbr')
```

## **DATA CLEANING & WRANGLING**

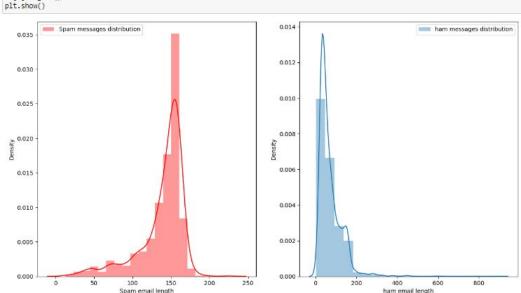
### Remove stop words with nltk library

```
In [22]: # Replace email addresses with 'email'
         df['message'] = df['message'].str.replace(r'^.+@[^\.].*\.[a-z]{2,}$','emailaddress')
In [23]: #Replace URLS with 'webaddress' I
         df['message'] = df['message'].str.replace(r'^http\://[a-zA-Z0-9\-\.]+\.[a-zA-Z]{2,3}(/\S*)?$',
         'webaddress')
In [24]: # Replace money symbols with 'moneysymb(£ can by typed with ALT key + 156)
         df['message'] = df['message'].str.replace(r'f|\$', 'dollers')
In [25]: #RepLace 10 digit phone numbers (formats include paranthesis, spaces, no spaces, dashes) with 'phonenumber"
         df['message'] = df['message'].str.replace(r'^\(?P[\d]{3}\)?[\s-]?[\d]{3}[\s-]?[\d]{4}$',
          'phonenumber')
In [26]: #Replace numbers with 'numbr
         df['message'] = df['message'].str.replace(r'\d+(\.\d+)?', 'numbr')
```

# **DATA CLEANING & WRANGLING**

### Original length vs clean length

```
In [37]: # Message distribution BEFORE cleaning
f, ax = plt. subplots(1,2, figsize= (15,8))
    sns.distplot(df[df['label']==]['length'], bins=20, ax=ax[0], label='Spam messages distribution',color='r')
    ax[0].set_xlabel('spam email length')
    ax[0].legend()
    sns.distplot(df[df['label']==0]['length'],bins=20, ax=ax[1], label='ham messages distribution')
    ax[1].set_xlabel('ham email length')
    ax[1].legend()
    plt.show()
```



# **DATA ANALYSIS**

### Some of the common spam words revealed with word cloud

[38]: #Getting sense of word cloud in spam

```
from wordcloud import WordCloud
spams=df['message'][df['label']==1]
spam_cloud = WordCloud(width=700,height=500, background_color='white', max_words=20).generate(' '.join(spams))
plt.figure(figsize=(10,8), facecolor='r')
plt.imshow(spam_cloud)
plt.axis('off')
plt.tight_layout(pad=0)
plt.show()
            llnumbr_ numbrp
         points t numbrppm
```

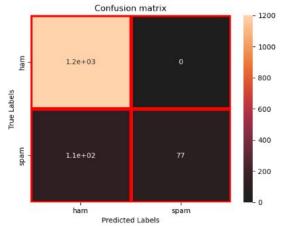
# **DATA ANALYSIS**

### Accuracy score and classification report

```
In [42]: # Train and predict
         X_train,x_test, Y_train,y_test=train_test_split(X,y,random_state=42)
         naive.fit(X train, Y train)
         y_pred=naive.predict(x_test)
         print('Final score = > ', accuracy_score(y_test,y_pred))
         Final score = > 0.91816223977028
In [43]: print (classification_report(y_test, y_pred))
                       precision
                                   recall f1-score support
                            0.91
                                     1.00
                                               0.95
                                                         1202
                                     0.40
                           1.00
                                               0.57
                                                          191
                                               0.92
                                                         1393
             accuracy
            macro avg
                            0.96
                                     0.70
                                               0.76
                                                         1393
         weighted avg
                            0.93
                                               0.90
                                     0.92
                                                         1393
```

# **DATA ANALYSIS-CORRELATION MATRIX**

```
In [51]: # plot confusion matrix heatmap
conf_mat = confusion_matrix(y_test,y_pred)
ax=plt.subplot()
sns.heatmap(conf_mat, annot=True, ax=ax, linewidths=5,linecolor='r',center=0)
ax.set_xlabel('Predicted Labels');ax.set_ylabel('True Labels')
ax.set_title('Confusion matrix')
ax.xaxis.set_ticklabels (['ham', 'spam'])
ax.yaxis.set_ticklabels(['ham', 'spam'])
plt.show()
```



In [52]: conf\_mat

Out[52]: array([[1202, 0]





# THANKS!