Information about data:

- -> The data set is taken from the UCI machine learning repository
- -> All the messages in the data set are divided into SPAM or HAM
- -> There were 5574 messages in the data set

OBJECTIVE:

- -> Analyze the messgaes from both spam and ham
- -> Dividing the messages into spam or ham by using Natural Language Processing

Importing required libraries and dataset

In [60]:

```
import nltk
from nltk.corpus import stopwords
import pandas as pd
import matplotlib.pyplot as mp
import seaborn as s
```

In [61]:

```
data = pd.read_csv("SMSSpamCollection", sep='\t', names= ['label', 'message'
])
print(data.head(5))
```

```
label 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...

4 ham Nah I don't think he goes to usf, he lives aro...
```

In [62]:

```
data.describe()
```

Out[62]:

	label	message
count	5572	5572
unique	2	5169
top	ham	Sorry, I'll call later
freq	4825	30

In [63]:

```
data.groupby('label').describe()
```

Out[63]:

	messa	ge					
	count	unique	top	freq			
label							
ham	4825	4516	Sorry, I'll call later	30			
spam	747	653	Please call our customer service representativ	4			

Adding length column to the dataframe which is the length of each Message

In [64]:

```
data['length'] = data['message'].apply(len)
```

In [65]:

```
# Sample of daat frame data.head(5)
```

Out[65]:

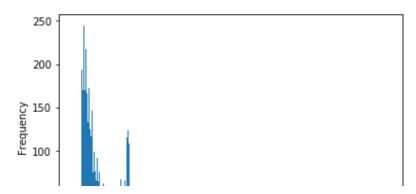
	label	message	length
0	ham	Go until jurong point, crazy Available only	111
1	ham	Ok lar Joking wif u oni	29
2	spam	Free entry in 2 a wkly comp to win FA Cup fina	155
3	ham	U dun say so early hor U c already then say	49
4	ham	Nah I don't think he goes to usf, he lives aro	61

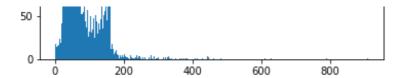
In [66]:

```
# Histogram of lengths of messages
data['length'].plot.hist(bins=400)
```

Out[66]:

```
<matplotlib.axes._subplots.AxesSubplot at 0x1fbf1c49278>
```





In [67]:

```
#Information about the length column data['length'].describe()
```

Out[67]:

count	5572.000000
mean	80.489950
std	59.942907
min	2.000000
25%	36.000000
50%	62.000000
75%	122.000000
max	910.000000

Name: length, dtype: float64

In [68]:

```
# Highest length message
data[data['length']==910]['message'].iloc[0]
```

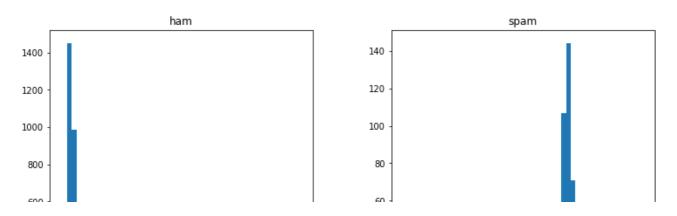
Out[68]:

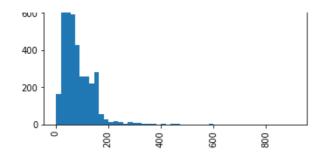
"For me the love should start with attraction.i should feel that I need her every time around me.she should be the first thing which comes in my though ts.I would start the day and end it with her.she should be there every time I dream.love will be then when my every breath has her name.my life should happen around her.my life will be named to her.I would cry for her.will giv e all my happiness and take all her sorrows.I will be ready to fight with a nyone for her.I will be in love when I will be doing the craziest things for her.love will be when I don't have to proove anyone that my girl is the m ost beautiful lady on the whole planet.I will always be singing praises for her.love will be when I start up making chicken curry and end up making sa mbar.life will be the most beautiful then.will get every morning and thank god for the day because she is with me.I would like to say a lot..will tell later.."

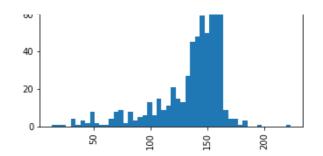
In [69]:

```
print("Length histograms of ham and spam")
data.hist(column='length', by='label',bins=50, figsize=(12,6))
mp.show()
```

Length histograms of ham and spam



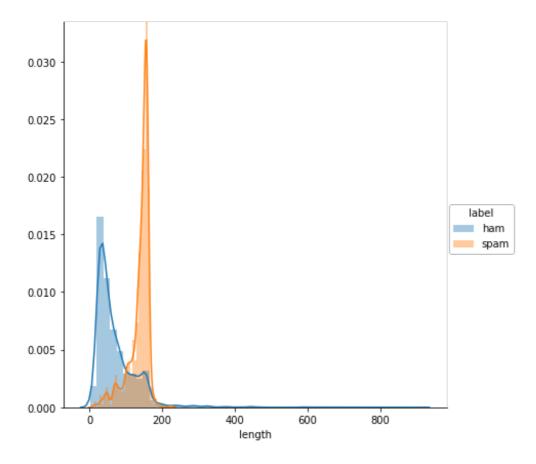




In [70]:

```
print("Using Seaborn the length distribution plots")
s.FacetGrid(data=data, hue='label', size=6).map(s.distplot,
    'length').add_legend()
mp.show()
```

Using Seaborn the length distribution plots



Observation:

- \rightarrow Most of the ham messages are near by to the length of 100
- $\ensuremath{\text{--}}\xspace$ Spam messages are more in the range between the length of 150 and 200

DATA PRE-PROCESSING

In [107]:

```
import string
print("Sample processing of a message")
print("-----")
```

```
mess = 'Message @ pre - processing to show the <sample>'
print("Message :", mess)
print("----")
print("After removing the punctuations with the help of string punctuation
all the words in the message are converted to list")
print("----")
nopunc = [c for c in mess if c not in string.punctuation]
print(nopunc)
print('----')
print ("Joining the characters into words to form a sentence and for further
computation")
print("----")
nopunc = ''.join(nopunc)
print(nopunc)
print("----")
print("Splitting to remove stop words")
nopunc = nopunc.split()
print(nopunc)
clean mess = [word for word in nopunc if word.lower() not in stopwords.word
s('english')]
print("----")
print("After removal of stop words")
print(clean mess)
Sample processing of a message
_____
Message : Message @ pre - processing to show the <sample>
_____
After removing the punctuations with the help of string punctuation all the
words in the message are converted to list
['M', 'e', 's', 's', 'a', 'g', 'e', ' ', ' ', 'p', 'r', 'e', ' ', ' 'p',
'r', 'o', 'c', 'e', 's', 'i', 'n', 'g', ' ', 't', 'o', ' ', 's', 'h',
'o', 'w', ' ', 't', 'h', 'e', ' ', 's', 'a', 'm', 'p', 'l', 'e']
Joining the characters into words to form a sentence and for further comput
ation
_____
Message pre processing to show the sample
_____
Splitting to remove stop words
['Message', 'pre', 'processing', 'to', 'show', 'the', 'sample']
After removal of stop words
['Message', 'pre', 'processing', 'show', 'sample']
In [109]:
def text process(mess):
   nopunc = [char for char in mess if char not in string.punctuation]
   nopunc = ''.join(nopunc)
   return [word for word in nopunc.split() if word.lower() not in
stopwords.words('english')]
In [110]:
print("Messages before applying the pre process")
```

```
print("Messages before applying the pre process")
data['message'].head(5)
```

Messages before applying the pre process

~

```
Out | 110 | :
0
     Go until jurong point, crazy.. Available only ...
1
                         Ok lar... Joking wif u oni...
2
     Free entry in 2 a wkly comp to win FA Cup fina...
3
     U dun say so early hor... U c already then say...
     Nah I don't think he goes to usf, he lives aro...
Name: message, dtype: object
In [111]:
print("Messages after applying the pre process")
data['message'].head(5).apply(text process)
Messages after applying the pre process
Out[111]:
     [Go, jurong, point, crazy, Available, bugis, n...
1
                         [Ok, lar, Joking, wif, u, oni]
2
     [Free, entry, 2, wkly, comp, win, FA, Cup, fin...
         [U, dun, say, early, hor, U, c, already, say]
3
     [Nah, dont, think, goes, usf, lives, around, t...
Name: message, dtype: object
Using Bag of Words approach to convert text to vectors
In [114]:
from sklearn.feature extraction.text import CountVectorizer
cv bow = CountVectorizer(analyzer=text process).fit(data['message'])
In [118]:
print("The number of words in the data set are:",len(cv_bow.vocabulary_))
The number of words in the data set are: 11425
In [128]:
print("Sample message before vector conversion")
m = data['message'][45]
print("Message:", data['message'][45])
print("Sample message after vector conversion")
bb = cv bow.transform([m])
print(bb)
Sample message before vector conversion
Message: No calls..messages..missed calls
Sample message after vector conversion
  (0, 5297) 1
  (0, 5298) 1
In [127]:
print("In the above message the ")
print(cv bow.get feature names()[5297])
print(cv bow.get feature names()[5298])
calls
callsmessagesmissed
```

Observation:

- \rightarrow By converting the sentence to vector with Bag of Words approach, has resulted with features 5297, 5298
- -> The features from the model has returned exactly the sentence whi ch is converted to vectors

In [154]:

```
bow_messages = cv_bow.transform(data['message'])
print("The shape of the resultant sparse matrix is:", bow_messages.shape)
```

The shape of the resultant sparse matrix is: (5572, 11425)

In [155]:

```
x_train = bow_messages[0:4000]
x_test = bow_messages[4000:]
y_train = data['label'][0:4000]
y_test = data['label'][4000:]
```

In [156]:

```
print(x_train.shape)
print(y_train.shape)
print(x_test.shape)
print(y_test.shape)
```

(4000, 11425) (4000,) (1572, 11425) (1572,)

In [157]:

```
from sklearn.naive_bayes import BernoulliNB
from sklearn.metrics import accuracy_score
spam_model = BernoulliNB().fit(x_train,y_train)
```

In [158]:

```
pred = spam_model.predict(x_test)
acc = accuracy_score(pred, y_test)
print(acc*100)
```

96.56488549618321

In [159]:

```
from sklearn.metrics import classification_report
print(classification_report(y_test,pred))
```

support	il-score	recall	precision	-
1359	0.98	0.99	0.97	ham
213	0.86	0.79	0.95	spam
1572	0.96	0.97	0.97	avg / total