## Autoimmune Tweets using the mostly preprocessed file from R and testing on Stemmed Tweets with 8 categories of autoimmune diseases¶

Those being: 0:Leukemia, 1: Fibromyalgia, 2:Kidney Disease, 3: Celiac Disease, 4: MS, 5: Hashimoto, 6: RA, 7: Chron's Disease

Tweets were taken from respective diseases in early December 2019 from 13 to 119 tweets for each disease, as many as were found that weren't mostly marketing, using "treatment in the search

```
%matplotlib inline
import pandas as pd
import matplotlib.pyplot as plt
from textblob import TextBlob
import numpy as np
from sklearn.feature_extraction.text import CountVectorizer
from sklearn.naive_bayes import MultinomialNB
from sklearn.metrics import classification_report, f1_score, accuracy_score, confusion __matrix

np.random.seed(47)
```

```
In [4]:

reviews = pd.read_csv('LemmaPythonRead.csv', encoding = 'unicode_escape')

#the encoding needed for python3 handling nonASCII chars
```

```
In [5]:
reviews.head()
```

Out[5]:

StemmedTweets AutoImmuneDisorder LemmatizedTweets unknown research unknown Celiac\_Disease unknownresearch the center fo... research\r\nunknownresearch\r\nthe cen.. lynn barter abc mc lbarter · dec lynn barter abc mc\r\nlbarter\r\n.\r\ndec Celiac Disease reply to thre... theona layne\r\ntheonawrites\r\n.\r\ndec theona layne theonawrites · dec Celiac\_Disease unknown diseas.. \r\nu...

	LemmatizedTweets	StemmedTweets AutoImmuneDisorder			
	<b>₹I</b>	bob simonoff\r\nsimonoffbob\r\n-\r\ndec \r\nth	Celiac_Disease		
- 14	<b>+</b>   -	gfdenver\r\ngfdenver\r\n-\r\nnov \r\nhm intere	Celiac_Disease		

```
In [6]: reviews.tail()
```

Out[6]:

	LemmatizedTweets	StemmedTweets	AutolmmuneDisorder
502		pharmabot\r\nthepharmabot\r\n-\r\nnov \r\ncode	Leukemia_Disease
503	wcm lymphoma wcmclymphoma · dec select initial	wcm lymphoma\r\nwcmclymphoma\r\n-\r\ndec \r\ns	Leukemia_Disease
504	with cml check	medivizor\r\nmedivizor\r\n·\r\ndec \r\ncoping	Leukemia_Disease
	to rickyspurs	abi\r\nbrokenleadheart\r\n·\r\ndec \r\nreplyin	Leukemia_Disease
506	brooke xbrooke · dec reply to itsjojosiwa dear	brooke\r\n\r\n\r\nxbrooke\r\n·\r\ndec \r\nrepl	Leukemia_Disease

```
In [7]:
reviews.shape
Out[7]:
```

```
In [8]:
reviews = reviews.reindex(np.random.permutation(reviews.index))
print(reviews.head())
print(reviews.tail())
                                      LemmatizedTweets
407 medivizor medivizor · nov cope with cml check ...
196 medical news bulletin mednewsbulletin \cdot jun a \dots
359 drtharanga kumari wickramasooriya drtharanga ....
     nola unknown unknowndiary · sep reply to nolan...
39
245 christine blome blomechristine · jan our new t...
                                         StemmedTweets AutoImmuneDisorder
407 medivizor\r\nmedivizor\r\n\r\nnov \r\ncoping ...
                                                        Leukemia Disease
196 medical news bulletin\r\nmednewsbulletin\r\n \...
                                                             Fibromyalgia
359 drtharanga kumari wickramasooriya\r\ndrtharang...
                                                           Kidney Disease
39
     nola unknown\r\nunknowndiary\r\n \r\nsep \r\n...
                                                           Celiac Disease
245 christine blome\r\nblomechristine\r\n \cdot\r\njan ...
                                                               MS Disease
                                      LemmatizedTweets \
     r unknownunknown runknownunknown · h chronic o...
264 lorilynn lorilynn · nov multiple unknown be a ...
327 unknown guild□ theunknownguild · nov fridayfin...
390 drug topic drugtopics \cdot dec the fda have appro...
```

```
StemmedTweets AutoImmuneDisorder
r unknownunknown\r\nrunknownunknown\r\nr\nh\... Hashimoto_Disease
lorilynn\r\nlorilynn\r\nnov\r\nmultiple ... MS_Disease
unknown guild\r\nr\ntheunknownguild\r\nr\r\nnov ... MS_Disease
drug topics\r\ndrugtopics\r\nr\ndec \r\nthe ... Leukemia_Disease
fms news bot\r\nfmsbot\r\nr\nnov\r\nunknown... Fibromyalgia
```

In [9]:

reviews.groupby('AutoImmuneDisorder').describe()

Out[9]:

	LemmatizedTweets				StemmedTweets				
	count	tunique	;top	frec	coun	tunique	etop		fre
AutolmmuneDisorder									1_
Celiac_Disease	50	50	erin smith gfreefun - nov unknown disease trea	1	50	50	cari a\r\ncariknits\r\n·\r\nnov \r\nt ha	oday i	1
Chron_Disease	19	19	matthew johnson maerial · feb submission for d	1	19	19	purpose \r\nhappybelieber\r\n·\r\ı \r\nnig	njan	1
Fibromyalgia	99	95	woman in pain forgrace · nov for grace be prou	2	99	95	chronic disease coalition\r\nchronicrights\r\n		2
Hashimoto_Disease	30	29	colorado natural med drgravesco · dec naturopa		30	29	colorado natural med\r\ndrgravesco\r\n-\r\ndec		2
Kidney_Disease	43	43	coffeemeetscarnivore cofemtscarnivor · dec rep		43	43	marketsticker\r\nmarketsticker\r\r \r\	n-\r\nde	<sup>3</sup> 1
Leukemia_Disease	119	116	medivizor medivizor nov cope with cml check		119	116	medivizor\r\nmedivizor\r\n·\r\nnov	V	3
MS_Disease	119	119	erectile dysfunction news research erectiledys	1	119	119	amesh adalja\r\nameshaa\r\n·\r\r \r\n" they	ndec	1
RA_Disease	28	28	dr john cush unknownnow · oct the committee fo	1	28	28	gse health blog\r\ngsehealth\r\n-\'\r\nw	r\nsep	1

```
In [10]:

reviews['length'] = reviews['StemmedTweets'].map(lambda text: len(text))

print(reviews.head())

LemmatizedTweets \

407 medivizor medivizor · nov cope with cml check ...

196 medical news bulletin mednewsbulletin · jun a ...
359 drtharanga kumari wickramasooriya drtharanga ·...
39 nola unknown unknowndiary · sep reply to nolan...
245 christine blome blomechristine · jan our new t...

StemmedTweets AutoImmuneDisorder \
407 medivizor\r\nmedivizor\r\nr\nroping ... Leukemia_Disease
196 medical news bulletin\r\nmednewsbulletin\r\n·\... Fibromyalgia
```

```
359 drtharanga kumari wickramasooriya\r\ndrtharang... Kidney_Disease
39 nola unknown\r\nunknowndiary\r\n·\r\nsep \r\n... Celiac_Disease
245 christine blome\r\nblomechristine\r\n·\r\njan ... MS_Disease

length
407    140
196    256
359    328
39    337
245    203
```

```
In [12]:
reviews.length.describe()
                                                                              Out[12]:
         507.000000
count
         258.960552
mean
         97.788402
         83.000000
min
25%
         188.000000
50%
         258.000000
75%
         304.000000
         678.000000
max
Name: length, dtype: float64
```

```
print(list(reviews.StemmedTweets[reviews.length > 500].index)) #near the max for lengt
h of LemmatizedTweets
```

In [13]:

```
[75, 432, 105, 104, 176, 58, 129, 174, 145, 193, 26, 82, 109, 111, 99, 167, 149]
['Hashimoto_Disease', 'Leukemia_Disease', 'Fibromyalgia', 'Fib
', 'Celiac Disease', 'Chron Disease', 'Fibromyalgia', 'Fibromyalgia',
 'Fibromyalgia', 'Fibromyalgia']
                                                                                                                                                                                                                                In [14]:
%%time
reviews.hist(column='length', by='AutoImmuneDisorder', bins=10)
Wall time: 484 ms
                                                                                                                                                                                                                               Out[14]:
array([[<matplotlib.axes. subplots.AxesSubplot object at 0x000001F61F5E5240>,
                        <matplotlib.axes. subplots.AxesSubplot object at 0x000001F61F625278>,
                       <matplotlib.axes._subplots.AxesSubplot object at 0x000001F61F651828>],
                     [<matplotlib.axes._subplots.AxesSubplot object at 0x000001F61F684DD8>,
                       <matplotlib.axes._subplots.AxesSubplot object at 0x000001F61F6C13C8>,
                       <matplotlib.axes._subplots.AxesSubplot object at 0x000001F61F6F1978>],
                     [<matplotlib.axes._subplots.AxesSubplot object at 0x000001F61F723F28>,
                       <matplotlib.axes. subplots.AxesSubplot object at 0x000001F61F75E550>,
                       <matplotlib.axes. subplots.AxesSubplot object at 0x000001F61F75E588>]],
                  dtype=object)
```

print(list(reviews.AutoImmuneDisorder[reviews.length > 500]))

```
In [15]:

def split_into_tokens(review):

    #review = unicode(review, 'iso-8859-1')# in python 3 the default of str() previous
ly python2 as unicode() is utf-8
```

```
return TextBlob (review) .words
```

```
In [16]:
reviews.StemmedTweets.head().apply(split_into_tokens)
                                                                              Out[16]:
       [medivizor, medivizor, ., nov, coping, with, c...
       [medical, news, bulletin, mednewsbulletin, ., ...
359
       [drtharanga, kumari, wickramasooriya, drtharan...
39
       [nola, unknown, unknowndiary, ., sep, replying...
245
       [christine, blome, blomechristine, ., jan, our...
Name: StemmedTweets, dtype: object
                                                                               In [17]:
TextBlob("hello world, how is it going?").tags
                                                                              Out[17]:
[('hello', 'JJ'),
 ('world', 'NN'),
 ('how', 'WRB'),
('is', 'VBZ'),
 ('it', 'PRP'),
 ('going', 'VBG')]
                                                                               In [18]:
import nltk
nltk.download('stopwords')
[nltk data] Downloading package stopwords to
[nltk data] C:\Users\m\AppData\Roaming\nltk data...
[nltk_data]
              Package stopwords is already up-to-date!
                                                                              Out[18]:
True
                                                                               In [19]:
from nltk.corpus import stopwords
stop = stopwords.words('english')
stop = stop + [u'a',u'b',u'c',u'd',u'e',u'f',u'g',u'h',u'i',u'j',u'k',u'l',u'm',u'n',u
'o',u'p',u'q',u'r',u's',u't',u'v',u'w',u'x',u'y',u'z']
                                                                               In [20]:
def split into lemmas(review):
    #review = unicode(review, 'iso-8859-1')
    review = review.lower()
    #review = unicode(review, 'utf8').lower()
    #review = str(review).lower()
```

```
words = TextBlob(review).words
    # for each word, take its "base form" = lemma
    return [word.lemma for word in words if word not in stop]
reviews.StemmedTweets.head().apply(split into lemmas)
                                                                               Out[20]:
407
       [medivizor, medivizor, ., nov, coping, cml, ch...
       [medical, news, bulletin, mednewsbulletin, ...
196
359
       [drtharanga, kumari, wickramasooriya, drtharan...
       [nola, unknown, unknowndiary, , sep, replying...
       [christine, blome, blomechristine, ·, jan, new...
Name: StemmedTweets, dtype: object
                                                                                In [21]:
%%time
bow transformer = CountVectorizer(analyzer=split into lemmas).fit(reviews['StemmedTwee
print(len(bow transformer.vocabulary ))
4585
Wall time: 859 ms
                                                                                In [22]:
review4 = reviews['StemmedTweets'][42]
print(review4)
purna
purnamusic
gluten shouldn\Boxt be so painful no sleep night two advice tried antihistamines ginger t
ea ibuprofen and activated charcoal over the last hours unknown treatment
                                                                                In [23]:
bow4 = bow transformer.transform([review4])
print(bow4)
  (0, 48)
                 1
  (0, 83)
                 1
  (0, 224)
                 1
  (0, 678)
                 1
  (0, 1602)
                 1
  (0, 1622)
                 1
  (0, 1827)
(0, 2034)
                 1
  (0, 2298)
                 1
  (0, 2385)
                 1
  (0, 2942)
  (0, 3086)
  (0, 3387)
  (0, 3388)
```

```
(0, 3732) 1

(0, 3770) 1

(0, 4025) 1

(0, 4183) 1

(0, 4196) 1

(0, 4231) 1

(0, 4290) 1

(0, 4583) 1
```

```
In [24]:

%%time

reviews_bow = bow_transformer.transform(reviews['StemmedTweets'])

print('sparse matrix shape:', reviews_bow.shape)

print('number of non-zeros:', reviews_bow.nnz)

print('sparsity: %.2f%%' % (100.0 * reviews_bow.nnz / (reviews_bow.shape[0] * reviews_bow.shape[1])))

sparse matrix shape: (507, 4585)

number of non-zeros: 11905

sparsity: 0.51%

Wall time: 734 ms
```

```
# Split/splice into training ~ 80% and testing ~ 20%

reviews_bow_train = reviews_bow[:400]

reviews_bow_test = reviews_bow[400:]

reviews_sentiment_train = reviews['AutoImmuneDisorder'][:400]

reviews_sentiment_test = reviews['AutoImmuneDisorder'][400:]

print(reviews_bow_train.shape)

print(reviews_bow_test.shape)

(400, 4585)
(107, 4585)
```

```
In [26]:
%time review_sentiment = MultinomialNB().fit(reviews_bow_train, reviews_sentiment_train)

Wall time: 15.6 ms
```

```
print('predicted:', review_sentiment.predict(bow4)[0])
print('expected:', reviews.AutoImmuneDisorder[42])

predicted: Celiac_Disease
expected: Celiac_Disease
```

```
predictions = review sentiment.predict(reviews bow test)
print(predictions)
['Fibromyalgia' 'Fibromyalgia' 'MS_Disease' 'Leukemia_Disease'
 'MS Disease' 'Fibromyalgia' 'Leukemia Disease' 'Kidney Disease'
 'Hashimoto Disease' 'Fibromyalgia' 'Fibromyalgia' 'Leukemia Disease'
 'Fibromyalgia' 'Fibromyalgia' 'MS Disease' 'MS Disease' 'MS Disease'
 'Fibromyalgia' 'Fibromyalgia' 'Leukemia Disease' 'Leukemia Disease'
 'Fibromyalgia' 'Fibromyalgia' 'Leukemia Disease' 'Fibromyalgia'
 'Celiac Disease' 'Leukemia Disease' 'Fibromyalgia' 'Leukemia Disease'
 'Leukemia Disease' 'Fibromyalgia' 'Leukemia Disease' 'Leukemia Disease'
 'MS Disease' 'MS_Disease' 'Fibromyalgia' 'Leukemia_Disease' 'MS_Disease'
 'MS_Disease' 'Fibromyalgia' 'Hashimoto_Disease' 'MS_Disease' 'MS_Disease'
 'MS_Disease' 'MS_Disease' 'Leukemia_Disease' 'MS_Disease' 'Fibromyalgia'
 'Fibromyalgia' 'Fibromyalgia' 'Fibromyalgia' 'Fibromyalgia' 'MS Disease'
 'Leukemia Disease' 'Fibromyalgia' 'MS Disease' 'Leukemia Disease'
 'MS Disease' 'Leukemia Disease' 'Fibromyalgia' 'MS Disease' 'MS Disease'
 'Fibromyalgia' 'MS Disease' 'Leukemia Disease' 'Leukemia Disease'
 'Fibromyalgia' 'MS Disease' 'Leukemia Disease' 'Fibromyalgia'
 'Celiac Disease' 'MS Disease' 'Fibromyalgia' 'MS Disease' 'Fibromyalgia'
 'Leukemia Disease' 'MS Disease' 'MS Disease' 'Celiac Disease'
 'Fibromyalgia' 'Fibromyalgia' 'MS Disease' 'MS Disease' 'Fibromyalgia'
 'Leukemia Disease' 'Leukemia Disease' 'MS Disease' 'RA Disease'
 'Hashimoto Disease' 'Celiac Disease' 'MS Disease' 'Hashimoto Disease'
 'Celiac Disease' 'Fibromyalgia' 'Fibromyalgia' 'Celiac Disease'
 'MS_Disease' 'Fibromyalgia' 'Hashimoto_Disease' 'Celiac Disease'
 'MS Disease' 'Fibromyalgia' 'Fibromyalgia' 'MS Disease' 'MS Disease'
 'Leukemia Disease' 'Fibromyalgia']
```

```
In [29]:

print('accuracy', accuracy_score(reviews_sentiment_test, predictions))

print('confusion matrix\n', confusion_matrix(reviews_sentiment_test, predictions))

print('(row=expected, col=predicted)')

accuracy 0.6261682242990654

confusion matrix

[[2 0 3 0 0 0 2 0]

[1 0 1 0 0 0 0 2 1]

[0 0 21 0 0 0 4 0]

[2 0 4 5 0 0 0 0]

[1 0 0 0 0 1 1 1 0]

[0 0 0 0 0 0 21 1 0]

[1 0 3 0 0 1 17 0]

[0 0 5 0 0 0 6 0]]

(row=expected, col=predicted)
```

```
In [30]:

print(classification_report(reviews_sentiment_test, predictions))

#The F1 score can be interpreted as a weighted average of the precision and recall,

#where an F1 score reaches its best value at 1 and worst score at 0.
```

	precision	recall	f1-score	support
Celiac_Disease	0.29	0.29	0.29	7
Chron_Disease	0.00	0.00	0.00	5
Fibromyalgia	0.57	0.84	0.68	25

```
1.00 0.45
1.00 0.25
0.91 0.95
0.52 0.77
0.00 0.00
                                               0.62
                                                           11
Hashimoto Disease
   Kidney Disease
                                               0.40
 Leukemia Disease
                                               0.93
                                                            22
      MS Disease
                                               0.62
                                                            22
                                                            11
       RA Disease
                                               0.00
         accuracy
                                               0.63
                                                           107
                         0.54
                                    0.44
                                               0.44
                                                           107
        macro avq
     weighted avg
                         0.59
                                    0.63
                                               0.58
                                                           107
```

c:\users\m\anaconda2\envs\python36\lib\site-packages\sklearn\metrics\classification.py :1437: UndefinedMetricWarning: Precision and F-score are ill-defined and being set to 0.0 in labels with no predicted samples.

'precision', 'predicted', average, warn for)

```
In [31]:

def predict_review(new_review):
    new_sample = bow_transformer.transform([new_review])
    p = np.around(review_sentiment.predict_proba(new_sample), decimals=2)
    print(new_review, p,'\tMax: ', np.max(p), '\n')
```

The respective probabilities correspond to those diseases alphebatized as [[1-Celiac Disease, 2-Chron's Disease, 3-Fibromyalgia, 4-Hashimoto, 5-Kidney Disease, 6-Leukemia, 7-Multiple Sclerosis, 8-Rheumatoid Arthritis]

```
predict_review('sick. pain. sleepless. anxious.')

predict_review('digestive. hungry.')

predict_review('bruising. sleepy. tired. headache.')

predict_review('energy. crazy. manic. depressed. angry.')

sick. pain. sleepless. anxious. [[0.01 0.  0.89 0.01 0.01 0.01 0.05 0.01]] Max: 0.89

digestive. hungry. [[0.11 0.03 0.19 0.05 0.1 0.24 0.24 0.04]] Max: 0.24

bruising. sleepy. tired. headache. [[0.09 0.08 0.39 0.05 0.08 0.13 0.14 0.04]] Max: 0.39

energy. crazy. manic. depressed. angry. [[0.1 0.05 0.17 0.06 0.1 0.23 0.25 0.05]] Max: 0.25
```

```
In [ ]:
```