## Autoimmune Tweets using the mostly preprocessed file from R and testing on Lemmatized 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 sklearn
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 [2]:
reviews = pd.read_csv('LemmaPythonRead.csv', encoding = 'unicode_escape')
#the encoding needed for python3 handling nonASCII chars
```

```
In [3]:
    reviews.head()
```

**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<br>\r\nth | Celiac_Disease |  |  |  |
| - 14 | <b>+</b>   -     | gfdenver\r\ngfdenver\r\n-\r\nnov \r\nhm<br>intere | Celiac_Disease |  |  |  |

```
reviews.tail()
```

Out[4]:

|     | LemmatizedTweets                               | StemmedTweets AutoImmuneDisorder                     |                  |  |  |  |  |  |
|-----|--|--|------------------|--|--|--|--|--|
| 502 |  | pharmabot\r\nthepharmabot\r\n-\r\nnov<br>\r\ncode    | Leukemia_Disease |  |  |  |  |  |
| 503 | wcm lymphoma wcmclymphoma · dec select initial | wcm<br>lymphoma\r\nwcmclymphoma\r\n-\r\ndec<br>\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<br>\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 [5]:
reviews.shape
Out[5]:
```

```
In [6]:
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\r\nh\... Hashimoto_Disease
lorilynn\r\nlorilynn\r\n'\nnov \r\nmultiple ... MS_Disease
unknown guild\r\ntheunknownguild\r\nr\r\nnov ... MS_Disease
drug topics\r\ndrugtopics\r\nr\r\ndec \r\nthe ... Leukemia_Disease
fms news bot\r\nfmsbot\r\nr\nnov \r\nnnov \r\nunknown... Fibromyalgia
```

In [7]:

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

Out[7]:

|                    | LemmatizedTweets |        |   |      | StemmedTweets |        |  |      |
|--------------------|------------------|--------|---|------|---------------|--------|--|------|
|                    | count            | unique | top   | freq | count         | unique | etop   | freq |
| AutolmmuneDisorder | -                |        |   |      |               |        |  |      |
| Celiac_Disease     | 50               | 50     | blake parson<br>blakepparsons ·<br>dec need help<br>for | 1    | 50            | 50     | np agarwal\r\nnpgrwl\r\n·\r\nnov<br>\r\ntreatment    | 1    |
| Chron_Disease      | 19               | 19     | thomas and<br>ethel bakery<br>thomasandethel<br>· nov r | 1    | 19            | 19     | tyler daniel\r\ntylerdaniel\r\n·\r\naug<br>\r\nb     | 1    |
| Fibromyalgia       | 99               | 95     | fibro bloggers<br>fibrobloggers -<br>nov unknown<br>tre | 2    | 99            | 95     | chronic disease<br>coalition\r\nchronicrights\r\n    | 2    |
| Hashimoto_Disease  | 30               | 29     | colorado<br>natural med<br>drgravesco -<br>dec naturopa | 2    | 30            | 29     | colorado natural<br>med\r\ndrgravesco\r\n-\r\ndec    | 2    |
| Kidney_Disease     | 43               | 43     | stock shark<br>stocksharks -<br>dec today<br>announce p | 1    | 43            | 43     | eclinic<br>diagnostics\r\neclinicnigeria\r\nyour     | 1    |
| Leukemia_Disease   | 119              | 116    | medivizor<br>medivizor · nov<br>cope with cml<br>check  | 3    | 119           | 116    | medivizor\r\nmedivizor\r\n-\r\nnov<br>\r\ncoping     | 3    |
| MS_Disease         | 119              | 119    | uonresearch<br>uonresearch ·<br>jan fund signal<br>new  | 1    | 119           | 119    | multiple<br>sclerosis\r\nunknownbio\r\n·\r\nnov<br>\ | 1    |
| RA_Disease         | 28               | 28     | gse health blog<br>gsehealth · sep<br>what be the tr    | 1    | 28            | 28     | frontiers<br>medicine\r\nfrontmedicine\r\n-\r\noc    | 1    |

In [8]:
reviews.groupby('AutoImmuneDisorder').describe()
Out[8]:

|                   | LemmatizedTweets |        |   |      | StemmedTweets |        |  |      |
|-------------------|------------------|--------|---|------|---------------|--------|--|------|
|                   | count            | unique | top   | freq | count         | unique | top  | freq |
| AutolmmuneDisorde | r                | -      |   |      |               |        |  |      |
| Celiac_Disease    | 50               | 50     | blake parson<br>blakepparsons ·<br>dec need help<br>for | 1    | 50            | 50     | np agarwal\r\nnpgrwl\r\n·\r\nnov<br>\r\ntreatment    | 1    |
| Chron_Disease     | 19               | 19     | thomas and ethel bakery thomasandethel nov r            | 1    | 19            | riu    | tyler daniel\r\ntylerdaniel\r\n-\r\naug<br>\r\nb     | 1    |
| Fibromyalgia      | 99               | 95     | fibro bloggers<br>fibrobloggers -<br>nov unknown<br>tre | 2    | 99            | 95     | chronic disease<br>coalition\r\nchronicrights\r\n    | 2    |
| Hashimoto_Disease | 30               | 29     | colorado<br>natural med<br>drgravesco ·<br>dec naturopa | 2    | 30            |        | colorado natural<br>med\r\ndrgravesco\r\n·\r\ndec    | 2    |
| Kidney_Disease    | 43               | 43     | stock shark<br>stocksharks ·<br>dec today<br>announce p | 1    | 43            | 1/1/2  | eclinic<br>diagnostics\r\neclinicnigeria\r\nyour     | 1    |
| Leukemia_Disease  | 119              | 116    | medivizor<br>medivizor · nov<br>cope with cml<br>check  | 3    | 119           | 1116   | medivizor\r\nmedivizor\r\n·\r\nnov<br>\r\ncoping     | 3    |
| MS_Disease        | 119              | 119    | uonresearch<br>uonresearch ·<br>jan fund signal<br>new  | 1    | 119           |        | multiple<br>sclerosis\r\nunknownbio\r\n-\r\nnov<br>\ | 1    |
| RA_Disease        | 28               | 28     | gse health blog<br>gsehealth · sep<br>what be the tr    | 1    | 28            | 28     | frontiers<br>medicine\r\nfrontmedicine\r\n-\r\noc    | 1    |

```
In [9]:
reviews['length'] = reviews['LemmatizedTweets'].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\n\cdots\r\nov\r\ncoping ... Leukemia_Disease
196 medical news bulletin\r\nmednewsbulletin\r\n · \...
                                                            Fibromyalgia
359 drtharanga kumari wickramasooriya\r\ndrtharang...
                                                         Kidney Disease
                                                         Celiac Disease
39
    nola unknown\r\nunknowndiary\r\n \r\nsep \r\n...
245 christine blome\r\nblomechristine\r\n \cdot\r\njan ...
                                                              MS Disease
    length
407
       126
196
       245
```

```
359 312
39 319
245 196
```

```
reviews.length.plot(bins=20, kind='hist')

Out[10]:

<matplotlib.axes._subplots.AxesSubplot at 0x18f59ff4128>
```

```
In [11]:
reviews.length.describe()
                                                                               Out[11]:
         507.000000
count
mean
         243.998028
          92.843285
std
          77.000000
min
         175.500000
25%
50%
         244.000000
         287.000000
75%
         604.000000
max
Name: length, dtype: float64
```

```
In [12]:

print(list(reviews.LemmatizedTweets[reviews.length > 500].index)) #near the max for le
ngth of LemmatizedTweets

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

[75, 432, 105, 104, 58, 145, 26, 82, 109, 111, 99, 167, 149]
['Hashimoto_Disease', 'Leukemia_Disease', 'Fibromyalgia', 'Fibromyalgia']
```

```
In [13]:
%%time
reviews.hist(column='length', by='AutoImmuneDisorder', bins=10)
Wall time: 516 ms
                                                                              Out[13]:
array([[<matplotlib.axes._subplots.AxesSubplot object at 0x0000018F5A1780B8>,
        <matplotlib.axes._subplots.AxesSubplot object at 0x0000018F5A1C4E48>,
        [<matplotlib.axes._subplots.AxesSubplot object at 0x0000018F5A22C9E8>, <matplotlib.axes._subplots.AxesSubplot object at 0x0000018F5A25DF28>,
        <matplotlib.axes. subplots.AxesSubplot object at 0x0000018F5A29A518>],
       [<matplotlib.axes._subplots.AxesSubplot object at 0x0000018F5A2CCAC8>,
        <matplotlib.axes._subplots.AxesSubplot object at 0x0000018F5A3090F0>,
        <matplotlib.axes. subplots.AxesSubplot object at 0x0000018F5A309128>]],
      dtype=object)
                                                                              In [14]:
def split into tokens (review):
```

```
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 [15]: reviews.LemmatizedTweets.head().apply(split_into_tokens)
```

```
Out[15]:
407
       [medivizor, medivizor, ., nov, cope, with, cml...
196
       [medical, news, bulletin, mednewsbulletin, ., ...
359
       [drtharanga, kumari, wickramasooriya, drtharan...
       [nola, unknown, unknowndiary, \cdot, sep, reply, t...
39
       [christine, blome, blomechristine, ., jan, our...
Name: LemmatizedTweets, dtype: object
                                                                                In [16]:
TextBlob("hello world, how is it going?").tags
                                                                                Out[16]:
[('hello', 'JJ'),
 ('world', 'NN'),
 ('how', 'WRB'),
('is', 'VBZ'),
('it', 'PRP'),
 ('going', 'VBG')]
                                                                                In [17]:
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[17]:
True
                                                                                In [18]:
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 [19]:
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.LemmatizedTweets.head().apply(split_into_lemmas)
                                                                                 Out[19]:
407
       [medivizor, medivizor, \cdot, nov, cope, cml, chec...
       [medical, news, bulletin, mednewsbulletin, ., ...
196
359
       [drtharanga, kumari, wickramasooriya, drtharan...
39
       [nola, unknown, unknowndiary, ., sep, reply, n...
245
       [christine, blome, blomechristine, ·, jan, new...
Name: LemmatizedTweets, dtype: object
                                                                                  In [20]:
bow_transformer = CountVectorizer(analyzer=split_into_lemmas).fit(reviews['LemmatizedT
weets'])
print(len(bow transformer.vocabulary ))
4208
Wall time: 734 ms
                                                                                  In [21]:
review4 = reviews['LemmatizedTweets'][42]
print(review4)
purna purnamusic \cdot jun gluten shouldn \square t be so painful no sleep night two advice try
antihistamine ginger tea ibuprofen and activate charcoal over the last hour unknown tr
eatment
                                                                                  In [22]:
bow4 = bow transformer.transform([review4])
print(bow4)
  (0, 49)
  (0, 81)
                  1
  (0, 214)
  (0, 628)
                  1
  (0, 1458)
  (0, 1476)
  (0, 1665)
  (0, 1871)
                  1
  (0, 2113)
                  1
  (0, 2191)
                  1
  (0, 2713)
(0, 2848)
                  1
  (0, 3129)
  (0, 3130)
  (0, 3473)
                  1
  (0, 3702)
                  1
  (0, 3845)
                  1
  (0, 3873)
  (0, 3890)
  (0, 3944)
  (0, 4199)
                  1
  (0, 4206)
```

In [23]:

```
%%time
reviews bow = bow transformer.transform(reviews['LemmatizedTweets'])
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, 4208)
number of non-zeros: 11902
sparsity: 0.56%
Wall time: 781 ms
                                                                              In [24]:
# 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, 4208)
(107, 4208)
                                                                              In [25]:
%time review sentiment = MultinomialNB().fit(reviews bow train, reviews sentiment trai
Wall time: 78.1 ms
                                                                              In [26]:
print('predicted:', review sentiment.predict(bow4)[0])
print('expected:', reviews.AutoImmuneDisorder[42])
predicted: Celiac Disease
expected: Celiac Disease
                                                                              In [27]:
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' 'MS Disease'
'Celiac Disease' 'Fibromyalgia' 'Fibromyalgia' 'Fibromyalgia'
'MS_Disease' 'Leukemia_Disease' 'Fibromyalgia' 'MS_Disease'
'Leukemia_Disease' 'MS_Disease' 'Leukemia_Disease' 'Kidney_Disease'
'MS_Disease' 'Fibromyalgia' 'Fibromyalgia' 'MS_Disease' 'Leukemia_Disease' 'Fibromyalgia' 'Fibromyalgia'
'Leukemia Disease' 'Fibromyalgia' 'Celiac_Disease' 'MS_Disease'
'Fibromyalgia' 'MS Disease' 'Hashimoto Disease' 'Leukemia Disease'
'MS Disease' 'MS Disease' 'Celiac_Disease' 'MS_Disease' '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 [28]:

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

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

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

accuracy 0.6635514018691588

confusion matrix

[[2 0 1 0 1 0 3 0]

[2 0 0 0 0 0 2 1]

[0 0 23 0 0 0 2 0]

[2 0 3 6 0 0 0 0]

[1 0 0 0 0 1 1 1 0]

[0 0 0 0 0 0 21 1 0]

[1 0 2 0 0 1 18 0]

[0 0 5 0 0 0 6 0]]

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

```
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.
```

In [29]:

|                   | precision | recall | fl-score | support |
|-------------------|-----------|--------|----------|---------|
|                   |           |        |          |         |
| Celiac_Disease    | 0.25      | 0.29   | 0.27     | 7       |
| Chron Disease     | 0.00      | 0.00   | 0.00     | 5       |
| Fibromyalgia      | 0.68      | 0.92   | 0.78     | 25      |
| Hashimoto_Disease | 1.00      | 0.55   | 0.71     | 11      |
| Kidney Disease    | 0.50      | 0.25   | 0.33     | 4       |
| Leukemia Disease  | 0.91      | 0.95   | 0.93     | 22      |
| MS Disease        | 0.55      | 0.82   | 0.65     | 22      |
| RA Disease        | 0.00      | 0.00   | 0.00     | 11      |
| _                 |           |        |          |         |
| accuracy          |           |        | 0.66     | 107     |
| macro avg         | 0.49      | 0.47   | 0.46     | 107     |
|                   |           |        |          |         |

```
weighted avg 0.60 0.66 0.61 107
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 [40]:

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, '\t', 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]

```
In [39]:

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. 0.88 0.01 0.01 0.01 0.07 0.01]] M

ax: 0.88

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 0.09 0.13 0.13 0.05]]

Max: 0.39

energy. crazy. manic. depressed. angry. [[0.1 0.05 0.17 0.06 0.1 0.22 0.24 0.06]]

Max: 0.24
```

```
In []:
```