Assignment 1 (Unsupervised Learning) - Final report

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Task 1: Visualization Task

Steps

1. Data Processing

Extracted the content of each twitter account, and remove the URL, punctuations.

2. Calculation

Counted the number of the words and the frequency of each word. Pick the Top 10 words and check if they related to health.

3. Elaborated Processing

Using the nltk package, Removes the stop words.

Tokenized the text

4. Visualization

Using matplotlib, plotted graph of the probability of occurrence for the 10 most common words.

Results

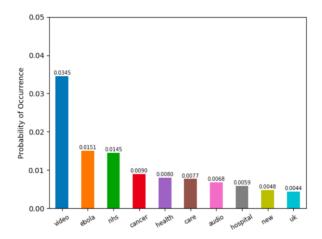
The 10 most common words and their word count (Take 'bbchealth' for example)

The most	10	common	words	are:
video		813		
ebola		355		
nhs		342		
cancer		213		
health		188		
care		182		
audio		160		
hospital		140		
new		114		
uk		104		

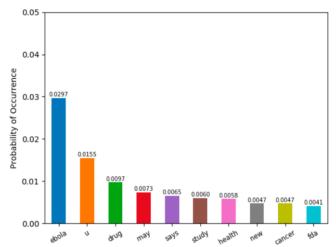
The total number of the words and after removing stop words (Take 'bbchealth' for example)

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The total number of words is: 23564
After removing stop words: 18731
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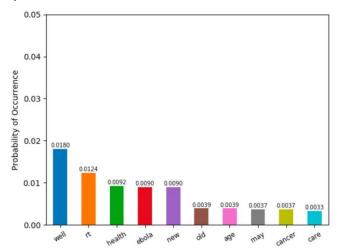
The probability of occurrence for the 10 most common words Bbchealth:



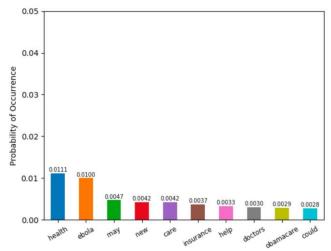
Reuters_health:



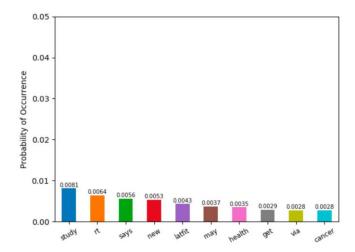
Nytimeshealth:



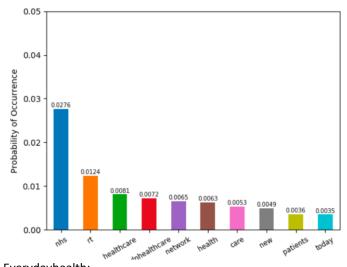
Nprhealth:



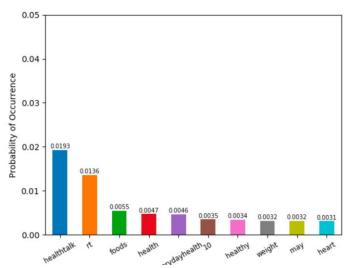
Latimeshealth:



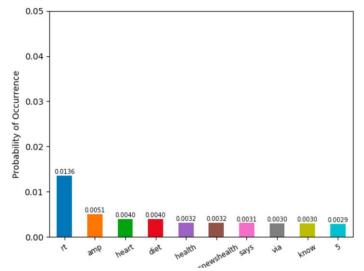
gdnhealthcare:



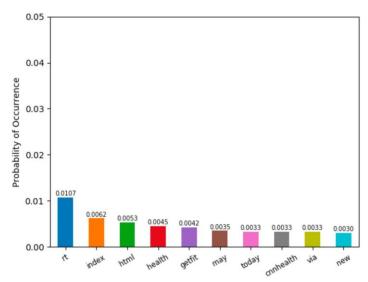
Everydayhealth:



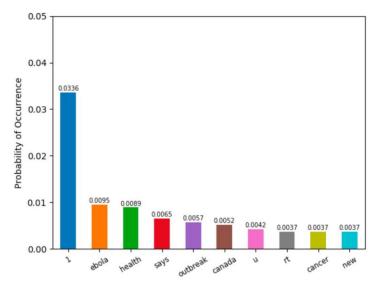
Usnewshealth:



Cnnhealth:



Cbchealth:

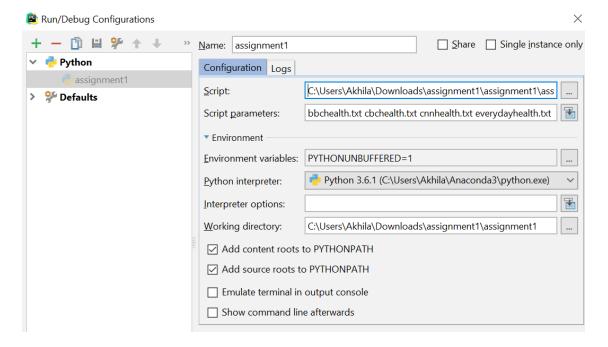


Task 2: Clustering Task

Steps

Files import
 Inputted all the files as script parameters to the program.

```
for file_name in sys.argv[1:]:
    list=[]
```



2. Data preprocessing

Extracted the content of each twitter account, and remove the URL, punctuations. Added original labels to assign value to tweets belonging to one Twitter account.

3. Vectorization of texts - TfidfVectorizer

Made a vector respresentation of all transformed tweets in the data set, with tfidf technique.

Limited the features to 5000.

Converted to array for dimensionality reduction

4. Principal Component Analysis & K means

As mentioned in the assignment document, fixed 2 dimensions for clustering. Used tfidf matrix as input to K means.

The number of clusters were fixed to 16 as mentioned in the assignment document.

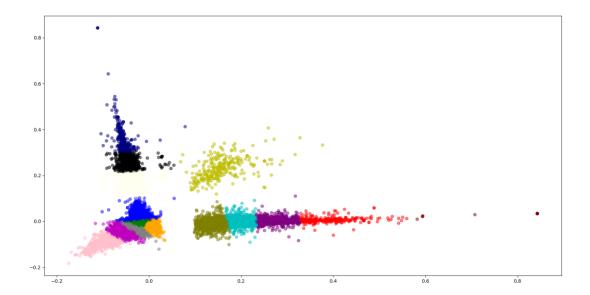
5. Comparison of Orginal Label vs Cluster Label

Compared the labeling of the tweets of cluster and original text.

6. Visualization

Defined color map.

Using matplotlib, plotted the clusters in 2 dimensions.



Analysis:

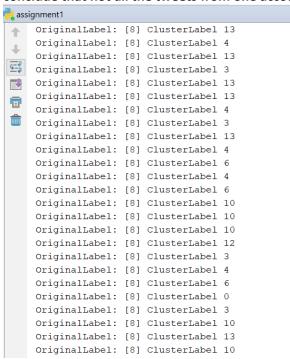
1.) Most common words were ebola, cancer, hospital, health.

All these words were related to health

Note: The word "Video" makes it to the list of top 10 words as its been used by every tweet to inform the followers of a link to Video.

The results would have been a bit different if "Video" would have been treated as a stop word.

2.) We analyzed that each twitter account doesn't form its own cluster.
As we compare the results of the labeling of cluster to the original data we can conclude that not all the tweets from one account formed a part of one cluster.



The clusters were formed due to familiarity in the text in tweets and since all of them had health related tweets, Its possible that most of content of the Tweets were similar.

• Problems encountered:

- 1. There are still some digitals and single letters in the Top10 common words list, as well as some special words used in social media, we need the further data processing and cleansing.
- 2. Some file cannot be opened in Pycharm due to the type of decoding, in the following days we are going to cope with it.