

project Data Mining

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Objectifs :

- Maitriser l'API de twitter pour l'extraction des tweets
- Maitriser la partie NLP (natural language processing) avec NLTK en Python
- Appliquer les principes de nettoyage des données
- Classer les tweets : regrouper ensemble les tweets qui sont similaires. C'est une étape qui peut-être considérée comme une étape

Specifications

Imaginons que vous avez un compte Twitter, et que vous lez suivre les tweets sur ce reseau social. Vu le nombre colossal de Tweets, et faute de temps, vous n'avez pas la possibilite de les lire tous. Pour cela, vous avez besoin d'une application qui va jouer le role d'assistant et qui va vous effectuer un resume de toutes ces informations. Une des approches qu'on peut utiliser est de le classer sous forme de groupes de sorte a ce qu'on presente a l'utilisateur un seul Tweet de chaque groupe. Pour cela, on doit proceder en trois grandes etapes :

Travail faire

On a Télécharger les tweets a partir de Twitter en utilisant l'API de twitter. Pour cela, vous devriez un compte « Twitter Developer ». Pour cela, vous devriez télécharger au moins 10 mille tweets. Pour la documentation de l'API de twitter, vous pouvez consulter les liens suivants :

```
In [19]: import pandas as pd
import tweepy
consumer_key="LHZVzcEN30hfmN2cPBqkoB3wq"
consumer_secret="DGZ7gQFD1qXoPfmAUwH0sY2eMTA0qhGKVb3rbExcX8Vhav3x3a"
access_token="1325046107437752325-a2zNm36NnzJqTFBFkIagjzpkdCadjs"
access_token_secret="7ohQJ7WTF2DuHsr9NNwPkOPXq5zUkaycrzo2nPhPUoGLL"
auth = tweepy.OAuthHandler(consumer_key, consumer_secret)
auth.set_access_token(access_token, access_token_secret)
api = tweepy.API(auth)
```

```
In [20]: twitter_data_analysis = pd.DataFrame(columns = ['text'])
tweets = tweepy.Cursor(api. user_timeline , id="twitter").items( 15000)

# Iterate and print tweets
i=0
for tweet in tweets:
    twitter_data_analysis.loc[i,"text"] = tweet.text
    i+=1
```

```
In [21]: import csv
twitter_data_analysis.to_csv('twitter_data_analysis.csv',index = False)
twitter_data_analysis.head(10)
```

Out[21]:

	text
0	RT @shesooosaddity: if you had a twitter befor...
1	@CloudNaii 40404
2	@issahairplug drink water replaced good morning
3	@Ne_ThatGuy we're taking oomf to the Fleets
4	@_JusJust_ remember "I dedicate my 500th Tweet...
5	@ambr_ncole they're tourists
6	@PhallonXOXO proof you're doing it right 😊
7	some of you hating...\n\nbut we see you Fleeti...
8	That thing you didn't Tweet but wanted to but ...
9	@quakerraina this is art

```
In [22]: twitter_data_analysis.info()

<class 'pandas.core.frame.DataFrame'>
Int64Index: 3218 entries, 0 to 3217
Data columns (total 1 columns):
#   Column  Non-Null Count  Dtype
---  -
0    text    3218 non-null    object
dtypes: object(1)
memory usage: 210.3+ KB
```

Pretraitement des tweets

Dans cette etape, l'objectif est d'eliminer le texte inutile des tweets tels que les #, les noms des utilisateurs, les url,emoji ...

```

In [23]: import re
for index, row in twitter_data_analysis.iterrows():
    err = row['text']
    new0 = re.sub(r"http\S+", "", err)
    new1 = re.sub(r"#\S+", "", new0)
    new2 = re.sub(r"@S+", "", new1)
    new3 = re.sub(r"\n+", "", new2)
    new4 = re.sub(r"RT+", "", new3)
    new5 = re.sub(r"hhh+", "", new4)
    emoji_pattern = re.compile("[
        u"\U0001F600-\U0001F64F" # emoticons
        u"\U0001F300-\U0001F5FF" # symbols & pictographs
        u"\U0001F680-\U0001F6FF" # transport & map symbols
        u"\U0001F1E0-\U0001F1FF" # flags (iOS)
        u"\U00002500-\U00002BEF" # chinese char
        u"\U00002702-\U000027B0"
        u"\U00002702-\U000027B0"
        u"\U000024C2-\U0001F251"
        u"\U0001F926-\U0001F937"
        u"\U00010000-\U0010ffff"
        u"\u2640-\u2642"
        u"\u2600-\u2B55"
        u"\u200d"
        u"\u23cf"
        u"\u23e9"
        u"\u231a"
        u"\ufe0f" # dingbats
        u"\u3030"
    "]" +", flags=re.UNICODE)
    new6 = re.sub(emoji_pattern, "", new5)
    twitter_data_analysis.loc[index, 'text'] = new6

```

```
In [24]: twitter_data_analysis.head(40)
```

```
Out[24]:
```

	text
0	if you had a twitter before 2020 rt this
1	40404
2	drink water replaced good morning
3	we're taking oomf to the Fleets
4	remember "I dedicate my 500th Tweet to:____"
5	they're tourists
6	proof you're doing it right
7	some of you hating...but we see you Fleeting
8	That thing you didn't Tweet but wanted to but ...
9	this is art
10	aren't we all six feet
11	this Tweet just graduated with honors
12	saw it, love it, can't wait for the wedding p...
13	
14	breathe
15	apology accepted
16	H2
17	THIRSTY
18	looking hydrated
19	the moon will share
20	bark among the stars
21	rubber ducky knew all along
22	If the moon can hydrate so can you
23	Reading an article before Retweeting it? That'...
24	Hey everyone, we made a temporary change to th...
25	Me seeing my Twitter friends I've never met ...
26	
27	dedication
28	not a single person on this app
29	but was it a good Tweet?
30	checks out
31	you forgot one:
32	mutual acknowledgment of good Tweets is frien...
33	

	text
34	how it started it never ended we get it
35	just make sure your DMs are open
36	that was a classic
37	no don't stop
38	strangers to bffs on Twitter real quick
39	cool cool cool cool cool

```
In [25]: twitter_data_analysis.to_csv('cleaning_twitter_data_analysis.csv', index = False)
```

```
In [26]: import nltk
nltk.download('stopwords')
```

```
[nltk_data] Downloading package stopwords to
[nltk_data] C:\Users\nidhal\AppData\Roaming\nltk_data...
[nltk_data] Package stopwords is already up-to-date!
```

```
Out[26]: True
```

Traitement des tweets: NLP (Natural LanguageProcessing)

On doit proceder a l'analyse du tweet en respectant les differentes etapes du NLP (Natural LanguageProcessing). La bibliotheque a utiliser est NLTK en Python.

```
In [27]: from nltk.stem.porter import PorterStemmer
from nltk.corpus import stopwords
ps = PorterStemmer()
stemmed_dataset=[]
for i in range(0, twitter_data_analysis.shape[0]):
    stemmed_array=twitter_data_analysis['text'][i].split()
    stemmed=[ps.stem(word) for word in stemmed_array if not word in set(stopwords)]
    stemmed=' '.join(stemmed)
    stemmed_dataset.append(stemmed)
print(stemmed_dataset[0:10])
```

```
['twitter 2020 rt', '40404', 'drink water replac good morn', "we'r take oomf fl", 'rememb "I dedic 500th tweet to:____"', "they'r tourist", 'proof right', 'hating...but see fleet', 'that thing didn't tweet want didn't got close like n', 'ah. We place for...', 'art']
```

```
In [28]: from sklearn.feature_extraction.text import CountVectorizer
cv = CountVectorizer()
X=cv.fit_transform(stemed_dataset)
print(X)
```

```
(0, 2571)      1
(0, 30)        1
(0, 2041)      1
(1, 44)        1
(2, 746)       1
(2, 2676)     1
(2, 1979)     1
(2, 1048)     1
(2, 1582)     1
(3, 2686)     1
(3, 2376)     1
(3, 1703)     1
(3, 938)      1
(4, 1972)     1
(4, 652)      1
(4, 49)       1
(4, 2558)     1
(4, 2476)     1
(4, 62)       1
(5, 2432)     1
(5, 2505)     1
(6, 1885)     1
(6, 2011)     1
(7, 938)      1
(7, 1126)     1
:             :
(3212, 1426)   1
(3212, 222)    1
(3212, 1691)   1
(3212, 516)    1
(3213, 349)    1
(3213, 2438)   1
(3213, 1327)   1
(3213, 181)    1
(3214, 2686)   1
(3214, 959)    1
(3214, 1426)   1
(3214, 2573)   1
(3214, 121)    1
(3214, 2724)   1
(3214, 515)    1
(3215, 1055)   1
(3215, 1162)   1
(3215, 884)    1
(3216, 429)    1
(3217, 1048)   2
(3217, 1582)   1
(3217, 2558)   1
(3217, 1642)   1
(3217, 142)    1
(3217, 1089)   1
```

Classification des tweets

Etant donne un ensemble de tweets, l'objectif est de les resumer sous formes de groupes de sorte a ce que les Tweets qui sont dans le meme groupe soient similaires. Ainsi, l'utilisateur pourra par la suite lire juste un Tweet de chaque groupe (le Tweet qui est le centro"ide de groupes). on a Utiliser l'algorithme K-Means pour classer les Tweets en k classes ,valeurs de k allant de 1 a30 par exemple).

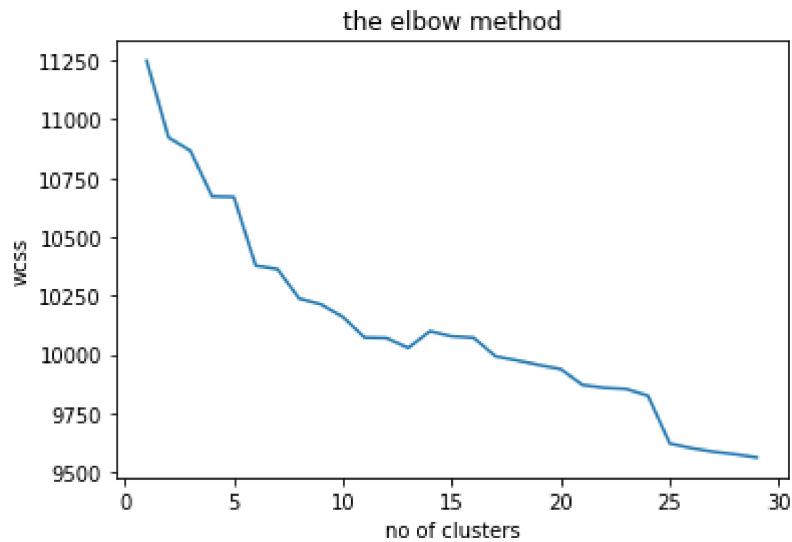
```
In [29]: from sklearn.cluster import KMeans  
wcss=[]
```

```
In [30]: for i in range(1,30):  
    Kmeans=KMeans(n_clusters=i,init='k-means++',max_iter=300,n_init=10,random_state=0)  
    Kmeans.fit(X)  
    wcss.append(Kmeans.inertia_)
```

```
Initialization complete  
Iteration 0, inertia 10683.000  
Iteration 1, inertia 10060.406  
Iteration 2, inertia 9975.293  
Iteration 3, inertia 9894.788  
Iteration 4, inertia 9657.315
```

In [31]:

```
import matplotlib.pyplot as plt
plt.plot(range(1,30),wcss)
plt.title('the elbow method')
plt.xlabel('no of clusters')
plt.ylabel('wcss')
plt.show()
```



In [32]:

```
true_k=30
Kmeans=KMeans(n_clusters=true_k,init='k-means++',n_init=1)
Kmeans.fit(X)
```

Out[32]: KMeans(algorithm='auto', copy_x=True, init='k-means++', max_iter=300, n_clusters=30, n_init=1, n_jobs=None, precompute_distances='auto', random_state=None, tol=0.0001, verbose=0)

La cellule suivante contient les mots cles de chaque cluster

In [33]:

```
print("Top terms per cluster:")
order_centroids = Kmeans.cluster_centers_.argsort()[:, ::-1]
terms = cv.get_feature_names()
for i in range(true_k):
    print("Cluster %d:" % i)
    for ind in order_centroids[i, :10]:
        print(' %s' % terms[ind])
    print()
print("\n")
```

```
read
that
saw
growth
pic
can
love
embrac
```

Cluster 11:

```
tweet
one
like
that
we
go
got
look
didn
eye
```

On a choisir un Tweet par classe comme representant. Les tweets choisis seront les resumes de toutes les informations contenues dans les tweets.

```
In [34]: result_final_twitter_data_analysis = pd.DataFrame(columns = [ 'text'])
i=0
j=0
while i<30:
    while True:
        Y=cv.transform([stemmed_dataset[j]])
        prediction=kmeans.predict(Y)
        if i == prediction:
            print("tweet of cluster "+str(prediction)+twitter_data_analysis.loc[i,"text"])
            result_final_twitter_data_analysis.loc[i,"text"] =twitter_data_analysis.loc[i,"text"]
            j=0
            break
        j+=1
    i+=1
result_final_twitter_data_analysis.to_csv('result_final_twitter_data_analysis.csv')
```

```
tweet of cluster [0] if you had a twitter before 2020 rt this
tweet of cluster [1] 40404
tweet of cluster [2] drink water replaced good morning
tweet of cluster [3] we're taking oomf to the Fleets
tweet of cluster [4] remember "I dedicate my 500th Tweet to:____"
tweet of cluster [5] they're tourists
tweet of cluster [6] proof you're doing it right
tweet of cluster [7]some of you hating...but we see you Fleeting
tweet of cluster [8]That thing you didn't Tweet but wanted to but didn't but go
t so close but then were like nah. We have a place for...
tweet of cluster [9] this is art
tweet of cluster [10] aren't we all six feet
tweet of cluster [11] this Tweet just graduated with honors
tweet of cluster [12] saw it, love it, can't wait for the wedding pics
tweet of cluster [13]
tweet of cluster [14]breathe
tweet of cluster [15] apology accepted
tweet of cluster [16] H2
tweet of cluster [17] THIRSTY
tweet of cluster [18] looking hydrated
tweet of cluster [19] the moon will share
tweet of cluster [20] bark among the stars
tweet of cluster [21] rubber ducky knew all along
tweet of cluster [22]If the moon can hydrate so can you
tweet of cluster [23]Reading an article before Retweeting it? That's growth.Bef
ore you Retweet an article, we'll remind you to read it...
tweet of cluster [24]Hey everyone, we made a temporary change to the Retweet fu
nction. When you hit the Retweet button, you can either...
tweet of cluster [25] Me seeing my Twitter friends I've never met in person su
ceed.
tweet of cluster [26]
tweet of cluster [27] dedication
tweet of cluster [28] not a single person on this app
tweet of cluster [29] but was it a good Tweet?
```

conclusion :

on a charge les tweets d'apres l'api de twitter, on les a mis dans le fichier csv
twitter_data_analysis. puis on a fait le data cleaning et on a mis le resultat dans le fichier

cleaning_twitter_data_analysis.csv.Et enfin on a mis un tweet de chaque cluster dans le fichier **result_final_twitter_data_analysis.**

[lien github](https://github.com/hazbri/projectDataMining/)
[\(https://github.com/hazbri/projectDataMining/\)](https://github.com/hazbri/projectDataMining/)

In []: