DonorsChoose

DonorsChoose.org receives hundreds of thousands of project proposals each year for classroom projects in need of funding. Right now, a large number of volunteers is needed to manually screen each submission before it's approved to be posted on the DonorsChoose.org website.

Next year, DonorsChoose.org expects to receive close to 500,000 project proposals. As a result, there are three main problems they need to solve:

- How to scale current manual processes and resources to screen 500,000 projects so that they can be posted as quickly and as efficiently as possible
- · How to increase the consistency of project vetting across different volunteers to improve the experience for teachers
- How to focus volunteer time on the applications that need the most assistance

The goal of the competition is to predict whether or not a DonorsChoose.org project proposal submitted by a teacher will be approved, using the text of project descriptions as well as additional metadata about the project, teacher, and school. DonorsChoose.org can then use this information to identify projects most likely to need further review before approval.

About the DonorsChoose Data Set

The train.csv data set provided by DonorsChoose contains the following features:

Feature	Description
project_id	A unique identifier for the proposed project. Example: p036502
	Title of the project. Examples:
<pre>project_title</pre>	• Art Will Make You Happy!
	• First Grade Fun
	Grade level of students for which the project is targeted. One of the following enumerated values:
project grade category	• Grades PreK-2
project_grade_category	• Grades 3-5
	• Grades 6-8
	• Grades 9-12
	One or more (comma-separated) subject categories for the project from the following enumerated list of values:
	• Applied Learning
	• Care & Hunger
	• Health & Sports
	• History & Civics
	• Literacy & Language
project subject categories	• Math & Science
. 3 = 3 = 3	Music & The ArtsSpecial Needs
	• Warmth
	Examples:
	• Music & The Arts
	• Literacy & Language, Math & Science
school_state	State where school is located (Two-letter U.S. postal code). Example: WY
	One or more (comma-separated) subject subcategories for the project. Examples :
project subject subcategories	ene en mere (comma coparatou) eusjoch eusgenegenee ier mie projech =numproe r
F3333	
	• Literature & Writing, Social Sciences
	• Literature & Writing, Social Sciences
	• Literature & Writing, Social Sciences An explanation of the resources needed for the project. Example:
<pre>project_resource_summary</pre>	• Literature & Writing, Social Sciences
<pre>project_resource_summary project_essay_1</pre>	 Literacy Literature & Writing, Social Sciences An explanation of the resources needed for the project. Example: My students need hands on literacy materials to manage sensory
	• Literacy • Literature & Writing, Social Sciences An explanation of the resources needed for the project. Example: • My students need hands on literacy materials to manage sensory needs!

· ·	
Description Fourth application essay	Feature project_essay_4_
Datetime when project application was submitted. Example: 2016-04-28 12:43:56.245	<pre>project_submitted_datetime</pre>
A unique identifier for the teacher of the proposed project. Example: bdf8baa8fedef6bfeec7ae4ff1c15c56	teacher_id
Teacher's title. One of the following enumerated values:	
• nan Dr.	
• Mr.	teacher_prefix
• Mrs.	
• Ms.	
• Teacher.	
Number of project applications previously submitted by the same teacher. Example: 2	teacher_number_of_previously_posted_projects

^{*} See the section **Notes on the Essay Data** for more details about these features.

Additionally, the resources.csv data set provides more data about the resources required for each project. Each line in this file represents a resource required by a project:

Feature	Description
id	A project_id value from the train.csv file. Example: p036502
description	Desciption of the resource. Example: Tenor Saxophone Reeds, Box of 25
quantity	Quantity of the resource required. Example: 3
price	Price of the resource required. Example: 9.95

Note: Many projects require multiple resources. The <code>id</code> value corresponds to a <code>project_id</code> in train.csv, so you use it as a key to retrieve all resources needed for a project:

The data set contains the following label (the value you will attempt to predict):

Label	Description
	A binary flag indicating whether DonorsChoose approved the project. A value of 0 indicates the project was not approved,
project_is_approved	and a value of 1 indicates the project was approved.

Notes on the Essay Data

Prior to May 17, 2016, the prompts for the essays were as follows:

- __project_essay_1:__ "Introduce us to your classroom"
- __project_essay_2:__ "Tell us more about your students"
- __project_essay_3:__ "Describe how your students will use the materials you're requesting"
- __project_essay_3:__ "Close by sharing why your project will make a difference"

Starting on May 17, 2016, the number of essays was reduced from 4 to 2, and the prompts for the first 2 essays were changed to the following:

- __project_essay_1:__ "Describe your students: What makes your students special? Specific details about their background, your neighborhood, and your school are all helpful."
- __project_essay_2:__ "About your project: How will these materials make a difference in your students' learning and improve their school lives?"

For all projects with project_submitted_datetime of 2016-05-17 and later, the values of project_essay_3 and project_essay_4 will be NaN.

11th Assinment Truncated SVD

In [1]:

```
%matplotlib inline
import warnings
warnings.filterwarnings("ignore")
```

```
import sqiites
import pandas as pd
import numpy as np
import nltk
import string
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.feature extraction.text import TfidfTransformer
from sklearn.feature_extraction.text import TfidfVectorizer
from sklearn.feature extraction.text import CountVectorizer
from sklearn.metrics import confusion matrix
from sklearn import metrics
from sklearn.metrics import roc curve, auc
from nltk.stem.porter import PorterStemmer
import re
# Tutorial about Python regular expressions: https://pymotw.com/2/re/
import string
from nltk.corpus import stopwords
from nltk.stem import PorterStemmer
from nltk.stem.wordnet import WordNetLemmatizer
from gensim.models import Word2Vec
from gensim.models import KeyedVectors
import pickle
from tqdm import tqdm
import os
from plotly import plotly
import plotly.offline as offline
import plotly.graph_objs as go
offline.init_notebook_mode()
from collections import Counter
print('done')
!pip install -U -q PyDrive
done
                                     993kB 2.8MB/s
 Building wheel for PyDrive (setup.py) ... done
In [3]:
from pydrive.auth import GoogleAuth
from pydrive.drive import GoogleDrive
from google.colab import auth
from oauth2client.client import GoogleCredentials
# Authenticate and create the PyDrive client.
auth.authenticate user()
gauth = GoogleAuth()
gauth.credentials = GoogleCredentials.get_application_default()
drive = GoogleDrive(gauth)
# links to google drive
link='https://drive.google.com/open?id=18VAiuw3vfETGcuJOdicvkgQT0pSxF7Wy'
link3='https://drive.google.com/open?id=1Z6bjXmyCaoEzXYo_tRDwLTsfeA2F3K3j'
flufff, id2 = link3.split('=')
print (id2) # Verify that you have everything after '='
downloaded = drive.CreateFile({'id':id2})
downloaded.GetContentFile('glove vectors')
```

1Z6bjXmyCaoEzXYo_tRDwLTsfeA2F3K3j

1.1 Reading Data

```
In [4]:
```

```
fluff, id = link.split('=')
print (id) # Verify that you have everything after '='
```

```
# for project data
downloaded = drive.CreateFile({'id':id})
downloaded.GetContentFile('train data.csv')
project data = pd.read csv('train data.csv',nrows=40000)
print(project data.shape)
link1='https://drive.google.com/open?id=11uHEj9KOgWD9SU-CPgKyb6VrWqVos4uV'
# for resource data
fluff1, idi = link1.split('=')
print (idi) # Verify that you have everything after '='
downloaded = drive.CreateFile({'id':idi})
downloaded.GetContentFile('resources .csv')
resource data = pd.read csv('resources .csv')
print(resource_data .head(3))
18VAiuw3vfETGcuJOdicvkgQT0pSxF7Wy
(40000, 17)
11uHEj9KOgWD9SU-CPgKyb6VrWqVos4uV
       id
                                                description quantity price
0 p233245 LC652 - Lakeshore Double-Space Mobile Drying Rack 1 149.00
1 p069063 Bouncy Bands for Desks (Blue support pipes)
                                                                    3 14.95
2 p069063 Cory Stories: A Kid's Book About Living With Adhd
In [5]:
print("Number of data points in train data", project_data.shape)
print('-'*50)
print("The attributes of data :", project data.columns.values)
print(resource data.shape)
print(resource_data.columns.values)
Number of data points in train data (40000, 17)
______
The attributes of data: ['Unnamed: 0' 'id' 'teacher id' 'teacher prefix' 'school state'
 'project submitted datetime' 'project grade category'
 'project_subject_categories' 'project_subject_subcategories'
 'project_title' 'project_essay_1' 'project_essay_2' 'project_essay_3'
 'project essay 4' 'project resource summary'
 'teacher_number_of_previously_posted_projects' 'project_is_approved']
(1541272, 4)
['id' 'description' 'quantity' 'price']
In [6]:
#sort the datapoints by date <-
# how to replace elements in list python: https://stackoverflow.com/a/2582163/4084039
cols = ['Date' if x=='project submitted datetime' else x for x in list(project data.columns)]
#sort dataframe based on time pandas python: https://stackoverflow.com/a/49702492/4084039
project_data['Date'] = pd.to_datetime(project_data['project_submitted_datetime'])
project data.drop('project submitted datetime', axis=1, inplace=True) # we drop the col
project data.sort values(by=['Date'], inplace=True) # sort the values y date
# how to reorder columns pandas python: https://stackoverflow.com/a/13148611/4084039
project data = project data[cols]
```

```
project_data.head(2)
Out[6]:
       Unnamed:
                     id
                                           teacher_id teacher_prefix school_state
                                                                                Date project_grade_category project_s
                                                                               2016-
  473
         100660 p234804
                         cbc0e38f522143b86d372f8b43d4cff3
                                                                         GΑ
                                                                               04-27
                                                                                            Grades PreK-2
                                                             Mrs.
                                                                             00:53:00
                                                                               2016-
                                                                                                           Math &
                                                             Mrs.
                                                                                               Grades 3-5
 29891
         146723 p099708 c0a28c79fe8ad5810da49de47b3fb491
                                                                               04-27
                                                                             01:10:09
1.3 Text preprocessing
In [0]:
# merge two column text dataframe:
project_data["essay_title_combined"] = project_data["project_essay_1"].map(str) +\
                          project_data["project_essay_2"].map(str) + \
                           project_data["project_essay_3"].map(str) + \
                           project_data["project_essay_4"].map(str) + \
                           project_data["project_title"].map(str)
In [0]:
# merge two column text dataframe:
project_data["essay"] = project_data["project_essay_1"].map(str) +\
                          project_data["project_essay_2"].map(str) + \
                           project_data["project_essay_3"].map(str) + \
                           project_data["project_essay_4"].map(str)
In [9]:
project_data.head(2)
Out[9]:
       Unnamed:
                     id
                                           teacher_id teacher_prefix school_state
                                                                                Date project_grade_category project_s
                                                                               2016-
  473
         100660 p234804 cbc0e38f522143b86d372f8b43d4cff3
                                                             Mrs.
                                                                               04-27
                                                                                            Grades PreK-2
                                                                             00:53:00
                                                                               2016-
                                                                                                           Math &
 29891
         146723 p099708 c0a28c79fe8ad5810da49de47b3fb491
                                                                               04-27
                                                                                               Grades 3-5
                                                             Mrs.
                                                                         CA
                                                                             01:10:09
4
                                                                                                             F
In [0]:
# https://stackoverflow.com/a/47091490/4084039
import re
def decontracted(phrase):
    # specific
     phrase = re.sub(r"won't", "will not", phrase)
    phrase = re.sub(r"can\'t", "can not", phrase)
    # general
```

```
phrase = re.sub(r"n\'t", " not", phrase)
phrase = re.sub(r"\'re", " are", phrase)
phrase = re.sub(r"\'s", " is", phrase)
phrase = re.sub(r"\'d", " would", phrase)
phrase = re.sub(r"\'ll", " will", phrase)
phrase = re.sub(r"\'t", " not", phrase)
phrase = re.sub(r"\'t", " have", phrase)
phrase = re.sub(r"\'ve", " have", phrase)
phrase = re.sub(r"\'m", " am", phrase)
return phrase
```

In [0]:

```
# https://gist.github.com/sebleier/554280
# we are removing the words from the stop words list: 'no', 'nor', 'not'
stopwords= ['i', 'me', 'my', 'myself', 'we', 'our', 'ours', 'ourselves', 'you', "you're", "you've",
            "you'll", "you'd", 'yours', 'yourself', 'yourselves', 'he', 'him', 'his',
'himself', \
            'she', "she's", 'her', 'hers', 'herself', 'it', "it's", 'its', 'itself', 'they', 'them',
'their',\
            'theirs', 'themselves', 'what', 'which', 'who', 'whom', 'this', 'that', "that'll",
'these', 'those', \
            'am', 'is', 'are', 'was', 'were', 'be', 'been', 'being', 'have', 'has', 'had', 'having',
'do', 'does', \
            'did', 'doing', 'a', 'an', 'the', 'and', 'but', 'if', 'or', 'because', 'as', 'until', '
while', 'of', \
            'at', 'by', 'for', 'with', 'about', 'against', 'between', 'into', 'through', 'during',
'before', 'after',\
            'above', 'below', 'to', 'from', 'up', 'down', 'in', 'out', 'on', 'off', 'over', 'under'
, 'again', 'further',\
            'then', 'once', 'here', 'there', 'when', 'where', 'why', 'how', 'all', 'any', 'both', '&
ach', 'few', 'more', \
            'most', 'other', 'some', 'such', 'only', 'own', 'same', 'so', 'than', 'too', 'very', \
            's', 't', 'can', 'will', 'just', 'don', "don't", 'should', "should've", 'now', 'd', 'll'
, 'm', 'o', 're', \
            've', 'y', 'ain', 'aren', "aren't", 'couldn', "couldn't", 'didn', "didn't", 'doesn', "de
esn't", 'hadn',\
            "hadn't", 'hasn', "hasn't", 'haven', "haven't", 'isn', "isn't", 'ma', 'mightn',
"mightn't", 'mustn',\
           "mustn't", 'needn', "needn't", 'shan', "shan't", 'shouldn', "shouldn't", 'wasn',
"wasn't", 'weren', "weren't", \
            'won', "won't", 'wouldn', "wouldn't"]
                                                                                                 •
```

we are going to consider

```
- school_state : categorical data
- clean_categories : categorical data
- clean_subcategories : categorical data
- project_grade_category : categorical data
- teacher_prefix : categorical data
- project_title : text data
- text : text data
- project_resource_summary: text data (optinal)
- quantity : numerical (optinal)
- teacher_number_of_previously_posted_projects : numerical
- price : numerical
```

Preprocessing of project_subject_categories

```
categories = list(project_data['project_subject_categories'].values)
# remove special characters from list of strings python:
https://stackoverflow.com/a/47301924/4084039
# https://www.geeksforgeeks.org/removing-stop-words-nltk-python/
```

```
# https://stackoverflow.com/questions/23669024/how-to-strip-a-specific-word-from-a-string
# https://stackoverflow.com/questions/8270092/remove-all-whitespace-in-a-string-in-python
cat list = []
for i in categories:
    temp = ""
    # consider we have text like this "Math & Science, Warmth, Care & Hunger"
    for j in i.split(','): # it will split it in three parts ["Math & Science", "Warmth", "Care & E
unger"
        if 'The' in j.split(): # this will split each of the catogory based on space "Math & Science"
e"=> "Math","&", "Science"
            j=j.replace('The','') # if we have the words "The" we are going to replace it with ''(i
.e removing 'The')
        j = j.replace(' ','') # we are placeing all the ' '(space) with ''(empty) ex:"Math &
Science"=>"Math&Science"
       temp+=j.strip()+" " #" abc ".strip() will return "abc", remove the trailing spaces
        \texttt{temp} = \texttt{temp.replace('\&','\_')} \ \textit{\# we are replacing the \& value into}
    cat list.append(temp.strip())
project data['clean categories'] = cat list
project data.drop(['project subject categories'], axis=1, inplace=True)
from collections import Counter
my_counter = Counter()
for word in project_data['clean_categories'].values:
   my counter.update(word.split())
cat_dict = dict(my_counter)
sorted cat dict = dict(sorted(cat dict.items(), key=lambda kv: kv[1]))
4
```

Preprocessing of project_subject_subcategories

```
In [0]:
```

```
sub catogories = list(project data['project subject subcategories'].values)
# remove special characters from list of strings python:
https://stackoverflow.com/a/47301924/4084039
# https://www.geeksforgeeks.org/removing-stop-words-nltk-python/
# https://stackoverflow.com/questions/23669024/how-to-strip-a-specific-word-from-a-string
# https://stackoverflow.com/questions/8270092/remove-all-whitespace-in-a-string-in-python
sub cat list = []
for i in sub catogories:
    temp = ""
    # consider we have text like this "Math & Science, Warmth, Care & Hunger"
    for j in i.split(','): # it will split it in three parts ["Math & Science", "Warmth", "Care & E
unger"]
       if 'The' in j.split(): # this will split each of the catogory based on space "Math & Science
e"=> "Math","&", "Science"
           j=j.replace('The','') # if we have the words "The" we are going to replace it with ''(i
.e removing 'The')
       j = j.replace(' ','') # we are placeing all the ' '(space) with ''(empty) ex:"Math &
Science"=>"Math&Science"
       temp +=j.strip()+" "#" abc ".strip() will return "abc", remove the trailing spaces
       temp = temp.replace('&',' ')
    sub cat list.append(temp.strip())
project data['clean subcategories'] = sub cat list
project data.drop(['project subject subcategories'], axis=1, inplace=True)
# count of all the words in corpus python: https://stackoverflow.com/a/22898595/4084039
my counter = Counter()
for word in project data['clean subcategories'].values:
   my counter.update(word.split())
sub cat dict = dict(my_counter)
sorted sub cat dict = dict(sorted(sub cat dict.items(), key=lambda kv: kv[1]))
4
```

```
print(project data['project grade category'][:20]) # we have to remove the graddes from every row
       Grades PreK-2
29891
         Grades 3-5
23374 Grades PreK-2
7176
       Grades PreK-2
        Grades 3-5
35006
5145
           Grades 3-5
36468 Grades PreK-2
36358 Grades PreK-2
39438 Grades PreK-2
       Grades PreK-2
2521
         Grades 6-8
           Grades 3-5
34399
         Grades 6-8
5364
         Grades 3-5
29183
33043
          Grades 3-5
37160
          Grades 6-8
         Grades 9-12
27157
         Grades 3-5
38830
10985 Grades PreK-2
15560 Grades PreK-2
Name: project_grade_category, dtype: object
In [0]:
d= list(project_data['project_grade_category'].values)
# remove special characters from list of strings python:
https://stackoverflow.com/a/47301924/4084039
# https://www.geeksforgeeks.org/removing-stop-words-nltk-python/
# https://stackoverflow.com/questions/23669024/how-to-strip-a-specific-word-from-a-string
# https://stackoverflow.com/questions/8270092/remove-all-whitespace-in-a-string-in-python
grade cat list = []
for i in d:
    # consider we have text like this:
    for j in i.split(' '): # # split by spae
       j=j.replace('Grades','') # clean grades from the row
    grade cat list.append(j.strip())
project_data['clean_grade'] = grade_cat_list
project_data.drop(['project_grade_category'], axis=1, inplace=True)
# count of all the words in corpus python: https://stackoverflow.com/a/22898595/4084039
my counter = Counter()
for word in project data['clean grade'].values:
   my counter.update(word.split())
project_grade_category_dict= dict(my_counter)
```

2. Preparing our data for the models

kv: kv[1]))

In [14]:

2.1 Splitting data into Train and cross validation(or test): Stratified Sampling

sorted project grade category dict = dict(sorted(project grade category dict.items(), key=lambda

```
In [0]:

#Splitting Data into train and Test sklearn https://scikit-
learn.org/stable/modules/generated/sklearn.model_selection.train_test_split.html
from sklearn.model_selection import train_test_split
```

```
X_train, X_test, y_train, y_test = train_test_split(project_data,
                                                  project data['project is approved'],
                                                    stratify= project data['project is approved'],
                                                    test_size = 0.33
4
In [17]:
print(y train.value counts())
print(y_test.value_counts())
# huge imbalance
print(X train.head(1))
1 22663
Name: project is approved, dtype: int64
   11163
Name: project_is_approved, dtype: int64
     Unnamed: 0 id ... clean_subcategories clean_grade
         130430 p129765 ... Literature Writing
[1 rows x 19 columns]
In [0]:
#droping the y labels
#https://stackoverflow.com/questions/13411544/delete-column-from-pandas-dataframe-by-column-name
X_train.drop(["project_is_approved"], axis = 1, inplace = True)
#x test =
X_test.drop(["project_is_approved"], axis = 1, inplace = True)
```

Text preprocessing of train and test

```
In [19]:
```

```
#Proprocessing for essay
# Combining all the above stundents
from tqdm import tqdm
preprocessed_essays_train = []
# tqdm is for printing the status bar
for sentance in tqdm(X_train['essay'].values):
    sent = decontracted(sentance)
    sent = sent.replace('\\r', '')
    sent = sent.replace('\\r', '')
    sent = sent.replace('\\n', '')
    sent = re.sub('[^A-Za-z0-9]+', '', sent)
    # https://gist.github.com/sebleier/554280
    sent = ''.join(e for e in sent.split() if e.lower() not in stopwords)
    preprocessed_essays_train.append(sent.lower().strip())
```

```
In [20]:
```

```
#Proprocessing for essay
# Combining all the above stundents
from tqdm import tqdm
preprocessed_essays_test = []
# tqdm is for printing the status bar
for sentance in tqdm(X_test['essay'].values):
    sent = decontracted(sentance)
    sent = sent.replace('\\r', '')
    sent = sent.replace('\\"', '')
    sent = sent.replace('\\"', '')
    sent = re.sub('[^A-Za-z0-9]+', '', sent)
    # https://gist.github.com/sebleier/554280
    sent = ''.join(e for e in sent.split() if e.lower() not in stopwords)
    preprocessed essays test.append(sent.lower().strip())
```

```
100%| | 13200/13200 [00:07<00:00, 1852.69it/s]
In [21]:
#Proprocessing for essay
# Combining all the above stundents
from tqdm import tqdm
preprocessed titles train = []
# tqdm is for printing the status bar
for sentance in tqdm(X train['project title'].values):
 sent = decontracted(sentance)
 sent = sent.replace('\\r', ' ')
 sent = sent.replace('\\"', ' ')
sent = sent.replace('\\n', ' ')
 sent = re.sub('[^A-Za-z0-9]+', '', sent)
  # https://gist.github.com/sebleier/554280
 sent = ' '.join(e for e in sent.split() if e.lower() not in stopwords)
 preprocessed_titles_train.append(sent.lower().strip())
100%| 26800/26800 [00:00<00:00, 36155.19it/s]
In [22]:
#Proprocessing for essay
# Combining all the above stundents
from tqdm import tqdm
preprocessed titles test = []
# tqdm is for printing the status bar
for sentance in tqdm(X_test['project_title'].values):
 sent = decontracted(sentance)
 sent = sent.replace('\\r', '')
 sent = sent.replace('\\"', ' ')
 sent = sent.replace('\\n', '')
 sent = re.sub('[^A-Za-z0-9]+', ' ', sent)
  # https://gist.github.com/sebleier/554280
  sent = ' '.join(e for e in sent.split() if e.lower() not in stopwords)
 preprocessed_titles_test.append(sent.lower().strip())
100%| 13200/13200 [00:00<00:00, 37010.44it/s]
In [23]:
#Proprocessing for essay title combined
# Combining all the above stundents
from tqdm import tqdm
preprocessed essays titles train = []
# tqdm is for printing the status bar
for sentance in tqdm(X train['essay title combined'].values):
 sent = decontracted(sentance)
 sent = sent.replace('\\r', ' ')
 sent = sent.replace('\\"', ' ')
  sent = sent.replace('\\n', ' ')
 sent = re.sub('[^A-Za-z0-9]+', '', sent)
  # https://gist.github.com/sebleier/554280
 sent = ' '.join(e for e in sent.split() if e.lower() not in stopwords)
 preprocessed_essays_titles_train.append(sent.lower().strip())
        26800/26800 [00:14<00:00, 1790.21it/s]
In [241:
#Proprocessing for essay_titles_cobined
# Combining all the above stundents
from tqdm import tqdm
preprocessed essays titles test = []
# tqdm is for printing the status bar
for sentance in tqdm(X_test['essay_title_combined'].values):
  sent = decontracted(sentance)
  sent = sent.replace('\\r', ' ')
```

```
sent = sent.replace('\\"', ' ')
 sent = sent.replace('\\n', ' ')
 sent = re.sub('[^A-Za-z0-9]+', '', sent)
  # https://gist.github.com/sebleier/554280
 sent = ' '.join(e for e in sent.split() if e.lower() not in stopwords)
 preprocessed essays titles test.append(sent.lower().strip())
100%| | 13200/13200 [00:07<00:00, 1809.03it/s]
```

2.2 Make Data Model Ready: encoding numerical, categorical features

1.project subject categories convert categorical to vectors*

```
In [25]:
# convert train and test data of clean categories into vectors
# we use count vectorizer to convert the values into one
from sklearn.feature extraction.text import CountVectorizer
vectorizer1 = CountVectorizer(vocabulary=list(sorted cat dict.keys()), lowercase=False,
binary=True)
vectorizer1.fit(X train['clean categories'].values)
# firstly convert fit the train data into the vectoriaer then it learn hte vocablery
# we use the fitted CountVectorizer to convert the text to vector
X train cat = vectorizer1.transform(X train['clean categories'].values)
X test cat = vectorizer1.transform(X test['clean categories'].values)
print(vectorizer1.get feature names())
['Warmth', 'Care Hunger', 'History Civics', 'Music Arts', 'AppliedLearning', 'SpecialNeeds',
'Health_Sports', 'Math_Science', 'Literacy_Language']
In [26]:
print("After vectorizations")
print(X_train_cat.shape, y_train.shape)
print(X test cat.shape, y test.shape)
print("="*100)
After vectorizations
(26800, 9) (26800,)
(13200, 9) (13200,)
```

2.project subject subcategories convert categorical to vectors*

```
In [27]:
```

```
# convert train and test data of clean categories into vectors
# we use count vectorizer to convert the values into one
from sklearn.feature_extraction.text import CountVectorizer
vectorizer2 = CountVectorizer(vocabulary=list(sorted_sub_cat_dict.keys()), lowercase=False, binary
vectorizer2.fit(X train['clean subcategories'].values)
# firstly convert fit the train data into the vectoriaer then it learn hte vocablery
```

```
# we use the fitted CountVectorizer to convert the text to vector
X train subcat = vectorizer2.transform(X train['clean subcategories'].values)
X test subcat = vectorizer2.transform(X test['clean subcategories'].values)
print(vectorizer2.get_feature_names())
['Economics', 'CommunityService', 'FinancialLiteracy', 'ParentInvolvement', 'Extracurricular',
'ForeignLanguages', 'Civics Government', 'NutritionEducation', 'Warmth', 'Care Hunger',
'SocialSciences', 'PerformingArts', 'CharacterEducation', 'TeamSports', 'Other',
'College CareerPrep', 'Music', 'History Geography', 'Health LifeScience', 'EarlyDevelopment', 'ESL
', 'Gym Fitness', 'EnvironmentalScience', 'VisualArts', 'Health Wellness', 'AppliedSciences',
'SpecialNeeds', 'Literature Writing', 'Mathematics', 'Literacy']
In [28]:
print("After vectorizations")
print(X train subcat.shape, y train.shape)
#print(X cv subcat.shape, y cv.shape)
print(X_test_subcat.shape, y_test.shape)
print("="*100)
After vectorizations
(26800, 30) (26800,)
(13200, 30) (13200,)
*3 school state convert categorical to vectors**
In [29]:
# now time to cont the each words
from collections import Counter
my counter = Counter()
for word in project data['school state'].values:
   my counter.update(word.split()) # count the words
school state dict = dict(my counter) # store in dicionary
sorted_school_state_dict = dict(sorted(school_state_dict.items(), key=lambda kv: kv[1]))# sor it
print(sorted school state dict)
{'VT': 22, 'WY': 39, 'ND': 54, 'MT': 85, 'RI': 107, 'NH': 107, 'SD': 115, 'AK': 116, 'NE': 121,
'DE': 130, 'WV': 181, 'HI': 183, 'ME': 184, 'NM': 187, 'DC': 204, 'KS': 228, 'ID': 238, 'IA': 241,
'AR': 344, 'CO': 422, 'MN': 443, 'MS': 461, 'OR': 461, 'KY': 492, 'MD': 526, 'NV': 539, 'AL': 620, 'CT': 630, 'UT': 631, 'TN': 632, 'WI': 663, 'VA': 739, 'NJ': 813, 'AZ': 816, 'OK': 836, 'MA': 858,
'LA': 872, 'WA': 891, 'MO': 924, 'IN': 936, 'OH': 960, 'PA': 1139, 'MI': 1185, 'SC': 1449, 'GA': 14
53, 'IL': 1598, 'NC': 1872, 'FL': 2238, 'TX': 2673, 'NY': 2730, 'CA': 5612}
4
In [30]:
# convert train,cv and test data of clean categories into vectors
# we use count vectorizer to convert the values into one
from sklearn.feature extraction.text import CountVectorizer
vectorizer3 = CountVectorizer(vocabulary=list(sorted school state dict.keys()), lowercase=False, b
vectorizer3.fit(project data['school state'].values)
# firstly convert fit the train data into the vectoriaer then it learn hte vocablery
# we use the fitted CountVectorizer to convert the text to vector
X_train_school_state = vectorizer3.transform(X_train['school_state'].values)
#X cv school state = vectorizer3.transform(X cv['school state'].values)
X test school state = vectorizer3.transform(X test['school state'].values)
print(vectorizer3.get feature names())
```

```
['VT', 'WY', 'ND', 'MT', 'RI', 'NH', 'SD', 'AK', 'NE', 'DE', 'WV', 'HI', 'ME', 'NM', 'DC', 'KS', 'I
D', 'IA', 'AR', 'CO', 'MN', 'MS', 'OR', 'KY', 'MD', 'NV', 'AL', 'CT', 'UT', 'TN', 'WI', 'VA', 'NJ',
'AZ', 'OK', 'MA', 'LA', 'WA', 'MO', 'IN', 'OH', 'PA', 'MI', 'SC', 'GA', 'IL', 'NC', 'FL', 'TX', 'NY
', 'CA'l
In [31]:
print("After vectorizations")
print(X train school state .shape, y train.shape)
#print(X cv school state .shape, y cv.shape)
print(X test school_state .shape, y_test.shape)
print("="*100)
After vectorizations
(26800, 51) (26800,)
(13200, 51) (13200,)
_____
In [321:
\# https://stackoverflow.com/questions/42224700/attributeerror-float-object-has-no-attribute-split
project data['clean grade']=project data['clean grade'].fillna("") # fill the nulll values with
# convert train, cv and test data of clean categories into vectors
# we use count vectorizer to convert the values into one
from sklearn.feature_extraction.text import CountVectorizer
vectorizer4 = CountVectorizer(vocabulary=list(sorted_project_grade_category_dict.keys()),
lowercase=False, binary=True)
vectorizer4.fit(project data['clean grade'].values)
# firstly convert fit the train data into the vectoriaer then it learn hte vocablery
# we use the fitted CountVectorizer to convert the text to vector
X_train_project_grade_category = vectorizer4.transform(X_train['clean_grade'].values)
#X_cv_project_grade_category = vectorizer4.transform(X_cv['clean_grade'].values)
X test project grade category = vectorizer4.transform(X test['clean grade'].values)
print(vectorizer4.get feature names())
['9-12', '6-8', '3-5', 'PreK-2']
In [33]:
print("After vectorizations")
print(X train project grade category .shape, y train.shape)
#print(X_cv_project_grade_category .shape, y_cv.shape)
print(X test project_grade_category .shape, y_test.shape)
print("="*100)
After vectorizations
(26800, 4) (26800,)
(13200, 4) (13200,)
In [0]:
#https://stackoverflow.com/questions/42224700/attributeerror-float-object-has-no-attribute-split
project data['teacher prefix']=project data['teacher prefix'].fillna(" ")# fill1 the null values
```

with space

```
my counter = Counter()
for word in project data['teacher prefix'].values:
   my counter.update(word.split())
# dict sort by value python: https://stackoverflow.com/a/613218/4084039
teacher_cat_dict = dict(my_counter)
sorted teacher prefix dict = dict(sorted(teacher cat dict.items(), key=lambda kv: kv[1]))
In [35]:
# convert train,cv and test data of clean categories into vectors
# we use count vectorizer to convert the values into one
from sklearn.feature_extraction.text import CountVectorizer
vectorizer5 = CountVectorizer(vocabulary=list(sorted_teacher_prefix_dict.keys()), lowercase=False,
binary=True)
vectorizer5.fit(project data['teacher prefix'].values.astype('U'))
# firstly convert fit the train data into the vectoriaer then it learn hte vocablery
# we use the fitted CountVectorizer to convert the text to vector
X train teacher prefix = vectorizer5.transform(X train['teacher prefix'].values.astype('U'))
#X cv teacher prefix= vectorizer5.transform(X cv['teacher prefix'].values.astype('U'))
\label{eq:continuous_continuous_state} X_{\texttt{teacher\_prefix}} = \text{vectorizer5.transform} (X_{\texttt{teacher\_prefix}'}]. values.astype('U'))
print(vectorizer5.get_feature_names())
# when i executeed this error comes
#np.nan is an invalid document, expected byte or unicode string.
# then iconvert to unicode just writ .astype('U') after the .values in fit and trainform
#https://stackoverflow.com/questions/39303912/tfidfvectorizer-in-scikit-learn-valueerror-np-nan-is
-an-invalid-document
['Dr.', 'Teacher', 'Mr.', 'Ms.', 'Mrs.']
In [36]:
print("After vectorizations")
print(X_train_teacher_prefix .shape, y_train.shape)
#print(X cv teacher prefix .shape, y cv.shape)
print(X_test_teacher_prefix .shape, y_test.shape)
print("="*100)
```

```
After vectorizations
```

(26800, 5) (26800,) (13200, 5) (13200,)

- - -

1.5.3 Vectorizing Numerical features¶

```
In [37]:
```

```
price_data = resource_data.groupby('id').agg({'price':'sum', 'quantity':'sum'}).reset_index()
project data = pd.merge(project data, price data, on='id', how='left')
print(price data.head(2))
# we also have to do this in tran, test and cv
# so also merge the resource data with the trian,cv and test
X train = pd.merge(X train, price data, on = "id", how = "left")
#print(x train.columns)
X_test = pd.merge(X_test, price_data, on = "id", how = "left")
#X_cv = pd.merge(X_cv, price_data, on = "id", how = "left")
```

```
0 p000001 459.56 7
1 p000002 515.89 21
```

Standadized price for the train, test and cv

```
In [0]:
```

```
# check this one: https://www.youtube.com/watch?v=0HOqOcln3Z4&t=530s
# standardization sklearn: https://scikit-
learn.org/stable/modules/generated/sklearn.preprocessing.StandardScaler.html \\
from sklearn.preprocessing import StandardScaler
from sklearn.preprocessing import MinMaxScaler
from sklearn import preprocessing
price scalar = StandardScaler()
price scalar.fit(X train['price'].values.reshape(-1,1)) # finding the mean and standard deviation
of this data
#print(f"Mean : {price_scalar.mean_[0]}, Standard deviation : {np.sqrt(price_scalar.var_[0])}")
# Now standardize the data with above maen and variance.
train price standar = price scalar.transform(X train['price'].values.reshape(-1, 1))
# Now standardize the data with above maen and variance.
test price standar = price scalar.transform(X test['price'].values.reshape(-1, 1))
# Now standardize the data with above maen and variance.
#cv price standar = price scalar.transform(X cv['price'].values.reshape(-1, 1))
```

Stadadized Previous_year_tecaher_projects train,test and cv

```
In [0]:
```

Standaized the Quantity column of the train,test and cv

```
In [0]:
```

```
price_scalar.fit(X_train['quantity'].values.reshape(-1,1)) # finding the mean and standard
deviation of this data
#print(f"Mean: {price_scalar.mean_[0]}, Standard deviation: {np.sqrt(price_scalar.var_[0])}")

# Now standardize the data with above maen and variance.
train_qnty_standar = price_scalar.transform(X_train['quantity'].values.reshape(-1, 1))
```

```
# Now standardize the data with above maen and variance.
#cv_qnty_standar = price_scalar.transform(X_cv['quantity'].values.reshape(-1, 1))

# Now standardize the data with above maen and variance.
test_qnty_standar = price_scalar.transform(X_test['quantity'].values.reshape(-1, 1))

In [41]:

print(X_train.shape)
print(X_test.shape)
```

New feature(No. of words in title)

```
In [0]:
```

(26800, 20) (13200, 20)

```
# For train data
title_length_train=[]
for i in range(0,X_train.shape[0]):
    title_length_train.append(len(X_train["project_title"][i].split()))

title_length_train=np.array(title_length_train)

#for test data titles
title_length_test=[]
for i in range(0,X_test.shape[0]):
    title_length_test.append(len(X_test["project_title"][i].split()))

title_length_test=np.array(title_length_test)
```

New feature(No. of words in combined essays)

```
In [0]:
```

```
#for test data esssay
essay_length_test=[]
for i in range(0,X_test.shape[0]):
    essay_length_test.append(len(X_test["essay"][i].split()))
essay_length_test=np.array(essay_length_test)

#for train data essay
essay_length_train=[]
for i in range(0,X_train.shape[0]):
    essay_length_train.append(len(X_train["essay"][i].split()))
essay_length_train=np.array(essay_length_train)
```

New feature(Sentiment scores of each combined essay's)

```
In [44]:
```

```
import nltk
from nltk.sentiment.vader import SentimentIntensityAnalyzer
nltk.download('vader_lexicon')

#https://www.programcreek.com/python/example/100005/nltk.sentiment.vader.SentimentIntensityAnalyzes
def analyze_sentiment(df):
```

```
sentiments = []
    sid = SentimentIntensityAnalyzer()
    for i in range(df.shape[0]):
       line = df['essay'][i]# take one essay
        sentiment = sid.polarity scores(line) # calculate the sentiment
        sentiments.append([sentiment['neg'], sentiment['pos'],
                           sentiment['neu'], sentiment['compound']]) # list of lists
    df[['neg', 'pos', 'neu', 'compound']] = pd.DataFrame(sentiments)
    df['Negative'] = df['compound'] < -0.1</pre>
    df['Positive'] = df['compound'] > 0.1
    return df
                                                                                                 Þ
/usr/local/lib/python3.6/dist-packages/nltk/twitter/ init .py:20: UserWarning:
The twython library has not been installed. Some functionality from the twitter package will not b
e available.
[nltk data] Downloading package vader lexicon to /root/nltk data...
In [0]:
X train=analyze sentiment(X train)
X test=analyze sentiment(X test)
```

Assinment 11: Truncated SVD

- step 1 Select the top 2k words from essay text and project_title (concatinate essay text with project title and then find the top 2k words) based on their <u>idf</u> values
- step 2 Compute the co-occurance matrix with these 2k words, with window size=5 (ref)
- step 3 Use <u>TruncatedSVD</u> on calculated co-occurance matrix and reduce its dimensions, choose the number of components (n_components) using <u>elbow method</u>
 - The shape of the matrix after TruncatedSVD will be 2000*n, i.e. each row represents a vector form of the corresponding word.
 - Vectorize the essay text and project titles using these word vectors. (while vectorizing, do ignore all the words which are not in top 2k words)
- step 4 Concatenate these truncatedSVD matrix, with the matrix with features
 - school_state : categorical data
 - clean_categories : categorical data
 - clean_subcategories : categorical data
 - project_grade_category :categorical data
 - teacher_prefix : categorical data
 - quantity : numerical data
 - teacher_number_of_previously_posted_projects : numerical data
 - price : numerical data
 - sentiment score's of each of the essay : numerical data
 - number of words in the title : numerical data
 - number of words in the combine essays : numerical data
 - word vectors calculated in step 3: numerical data

- step 5: Apply GBDT on matrix that was formed in step 4 of this assignment, DO REFER THIS BLOG: XGBOOST DMATRIX
- step 6:Hyper parameter tuning (Consider any two hyper parameters)
 - Find the best hyper parameter which will give the maximum AUC value
 - Find the best hyper paramter using k-fold cross validation or simple cross validation data
 - Use gridsearch cv or randomsearch cv or you can also write your own for loops to do this task of hyperparameter tuning

Step 1: Selecting top 2000 words from essay and project title based on their idf values.

```
F--7--- ---- ----- -----
In [66]:
from sklearn.feature_extraction.text import TfidfTransformer
from sklearn.feature_extraction.text import CountVectorizer
#Select the top 2k words from essay text and project_title (concatinate essay text with project ti
tle and then find the top 2k words) based on their idf_ values
#instantiate CountVectorizer()
cv=CountVectorizer()
word count vector=cv.fit transform(preprocessed essays titles train)
word count vector.shape
# this means 37022 unique words we have
Out[66]:
(26800, 37022)
In [67]:
 tfidf transformer=TfidfTransformer(smooth idf=True,use idf=True)
tfidf_transformer.fit(word_count_vector)
Out[67]:
TfidfTransformer(norm='12', smooth_idf=True, sublinear_tf=False, use_idf=True)
idf_ values and his coresponding feature name:
In [114]:
# we want descending order , because basically that means if the idf high that means its occurence
#in the corpus is very very less, that means -> most important and unique words that can define ou
r statement
# print idf values
df_idf = pd.DataFrame(tfidf_transformer.idf_
index=cv.get feature names(),columns=["tf idf weights"])
# sort ascending
sorted descending idf=df idf.sort values(by=['tf idf weights'],ascending=False)
print(sorted descending idf.head(5))
idf features=sorted_descending_idf.index[:2000]
                  tf idf weights
mcauliffe
                       10.503047
nannanafternoon
                       10.503047
nannanadvocating
                       10.503047
                       10 503047
nannanadventure
nannanadvancement
                       10.503047
```

step 2:Computing Co-occurance matrix

```
In [0]:

# # For reference:
# # https://gist.github.com/nkt1546789/e9fc84579b9c8356f1e5
# # https://stackoverflow.com/questions/41661801/python-calculate-the-co-occurrence-matrix

# # co-occurrence matrix code for train_data

# # ->window size
# window=5
# # -> 2000*2000 matrix of zeros
# cocc train = np_zeros/(len/idf_features) len/idf_features) np_float64)
```

```
# COOC_CLAIN - NP. Zelos ((len(lul_leacules), len(lul_leacules)), NP. Lloacol)
      -> Make a dictionary in which keys are your features, values are the indexes.
# list of words dict = {idf features[i]:i for i in range(len(idf features))}
      ->Loop to the total 2k words
# for i in (list_of_words_dict.keys()):
        ->Loop to the corpus
   for j in preprocessed_essays_titles_train:
     #split each word
#
       j = j.split()
#-> condition if our which are in 2k words not in this sentence then just skip this
sentence
       if str(i) not in j:
         continue
         # Loop to 2k words indexes, if our word is under window of some word then add it to the
co-occurence matrix;
#
        else:
         for x in list of words dict.values():
#
           if abs(list of words dict[i]-x)<window:</pre>
              cooc train[list of words dict[i],x]+=1
# # end loop:)
```

In [01:

```
# # start func:)
# # -> 2000*2000 matrix of zeros
cooc train=np.zeros([2000,2000])# 2k*2k features
def co occurence matrix(win,vocab,corpus,coo matrix):
      ->window size
 window=win
 # # -> total 2k words
 a=vocab
 # # -> loop to the each word in the top2k features
 for q,word in enumerate(vocab):
   # # ->Loop to the corpus
   for i in corpus:
    ## -> if word present in corpus then go further othersize skip it go to the next essay
     if word in i:
       # # -> split it into teh words in a list
       arr=[g for g in i.split(' ')]
        # for each word
       for j,d in enumerate(arr):
         arrr=[]
         # just make a loop to the and append all the elements which are under current word's win
dow
         for i in range(max(0,j-window),min(j+window,len(arr)-1)): # window size
           arrr.append(arr[x])
           #loop to the window start to end
           for f,w in enumerate(arrr):
             if wd in vocab:
               if wd!=word:
                 index=vocab.index(wd)
                 coo_matrix[q,index]+=1
   return coo_matrix
# # end func:)
```

(2000, 2000)

Step 3: Applying TruncatedSVD and Calculating Vectors for essay and project title

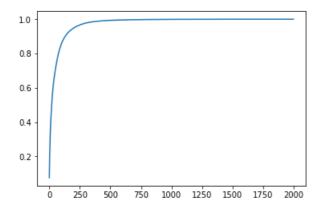
```
In [0]:
```

```
from sklearn.decomposition import TruncatedSVD

svd = TruncatedSVD(n_components=1999, n_iter=7, random_state=42)
svd.fit(cooc_train)
ratio=svd.explained_variance_ratio_
cum_sum=np.cumsum(ratio)
plt.plot(cum_sum)
```

Out[0]:

[<matplotlib.lines.Line2D at 0x7ff43e24aac8>]



Observations: From the 250 dimensions onwards, graph is not increasing so much, so we choose 250

```
In [0]:
```

```
svd = TruncatedSVD(n_components=250, n_iter=7, random_state=42)
result_train=svd.fit_transform(cooc_train)
```

In [0]:

```
print(result_train.shape)
(2000, 250)
```

In [0]:

```
model = result_train
glove_words = idf_features
keys={}
for i,j in enumerate(glove_words):
    keys[j]=i
```

```
# average Word2Vec
# compute average word2vec for each review.
def func(wordlist):

train_avg_w2v_vectors = []; # the avg-w2v for each sentence/review is stored in this list
```

```
ior sentence in tqum(wordlist): # ior each review/sentence
    vector = np.zeros(250) # as word vectors are of zero length
                                                                    # we are taking the 300
dimensions very large
   cnt_words =0; # num of words with a valid vector in the sentence/review
    for word in sentence.split(): # for each word in a review/sentence
        if word in glove_words:
          vector += model[keys[word]]
          cnt words += 1
    if cnt_words != 0:
        vector /= cnt words
    train_avg_w2v_vectors.append(vector)
  print(len(train_avg_w2v_vectors))
  print(len(train_avg_w2v_vectors[0]))
  return train_avg_w2v_vectors
In [0]:
train_avg_w2v_vectors=func (preprocessed_essays_train)
test_avg_w2v_vectors=func(preprocessed_essays_test)
test_avg_w2v_vectors_title=func (preprocessed_titles_test)
train_avg_w2v_vectors_title=func(preprocessed_titles_train)
              | 26800/26800 [01:12<00:00, 370.07it/s]
               | 76/13200 [00:00<00:34, 377.01it/s]
  1% |
26800
250
              | 13200/13200 [00:35<00:00, 368.11it/s]
100%I
               | 1162/13200 [00:00<00:01, 11618.74it/s]
  9%|
13200
250
100%|
             | 13200/13200 [00:01<00:00, 10828.80it/s]
               | 1867/26800 [00:00<00:02, 9620.25it/s]
  7% [
13200
250
         26800/26800 [00:02<00:00, 11022.08it/s]
100%|
26800
250
```

Step 4: Concatenate these truncatedSVD matrix, with the matrix with features

```
, \verb|X_train_project_grade_category|, \verb|X_train_school_state|, \verb|\#all_categorials||
                                                                                                    train qnty standar, train price standar, train prev proj standar,
                                                                                                   essay_length_train.reshape(-1,1),title_length_train.reshape(-1,1),
                                                                                                   \verb|pos.reshape(-1,1)|, \verb|neg.reshape(-1,1)|, \verb|com.reshape(-1,1)|, \verb|train_avg_w2v_vectors|, train_avg_w2v_vectors|, train_
ain avg w2v vectors title
                                                                                                                                                                                                                                                                                                                                             )) # all numericals
print(X_set5_train.shape, y_train.shape)
 #X train['pos'],X train['neg'],X train['neu'],
(26800, 607) (26800,)
In [0]:
pos=list(X_test['pos'])
pos=np.array(pos)
neg=list(X_test['neg'])
neg=np.array(neg)
com=list(X_test['compound'])
com=np.array(com)
 # combine all
from scipy.sparse import hstack
 # with the same hstack function we are concatinating a sparse matrix and a dense matirx :)
X set5_test = hstack((
                                                                                                    X_test_teacher_prefix,X_test_cat,X_test_subcat ,X_test_project_grade_category
 ,X_test_school_state, #all categorials
                                                                                                    test_qnty_standar,test_price_standar,test_prev_proj_standar,
                                                                                                    essay_length_test.reshape(-1,1),title_length_test.reshape(-1,1),
                                                                                                   \verb|pos.reshape(-1,1)|, \verb|neg.reshape(-1,1)|, \verb|com.reshape(-1,1)|, \verb|test_avg_w2v_vectors|, test_avg_w2v_vectors|, test_avg_w2v_vectors|
t avg w2v vectors title
                                                                                                                                                                                                                                                                                                                                              )) # all numericals
print(X_set5_test.shape, y_test.shape)
 #X_train['pos'],X_train['neg'],X_train['neu'],
(13200, 607) (13200,)
```

Step 5: Apply XgBoost on you matrix

In [01:

```
#Define a class
  #DO REFER THIS BLOG: XGBOOST DMATRIX
from sklearn.metrics import roc_auc_score
import matplotlib.pyplot as plt
from sklearn.ensemble import RandomForestClassifier
import xgboost as xgb
from sklearn.model_selection import train_test_split
from sklearn.model selection import GridSearchCV
from sklearn.model selection import cross val score
from sklearn.tree import DecisionTreeClassifier
from sklearn.ensemble import GradientBoostingClassifier
class XGBoostClassifier():
   def init (self, num boost round=10, **params):
       self.clf = None
       self.num_boost_round = num_boost_round
       self.params = params
       self.params.update({'objective': 'multi:softprob'})
   def fit(self, X, y, num_boost_round=None):
```

num_boost_round = num_boost_round or self.num_boost_round

self.label2num = {label: i for i, label in enumerate(sorted(set(y)))}

```
dtrain = xgb.DMatrix(X, label=[self.label2num[label] for label in y])
        self.clf = xgb.train(params=self.params, dtrain=dtrain, num_boost_round=num_boost_round, ve
rbose_eval=1)
   def predict(self, X):
       num2label = {i: label for label, i in self.label2num.items()}
       Y = self.predict proba(X)
       y = np.argmax(Y, axis=1)
       return np.array([num2label[i] for i in y])
   def predict_proba(self, X):
        dtest = xgb.DMatrix(X)
       return self.clf.predict(dtest)
   def score(self, X, y):
       Y = self.predict_proba(X)[:,1]
       return roc_auc_score(y, Y)
   def get_params(self, deep=True):
       return self.params
   def set_params(self, **params):
       if 'num boost round' in params:
            self.num_boost_round = params.pop('num_boost_round')
       if 'objective' in params:
            del params['objective']
       self.params.update(params)
       return self
```

Hyperparameter tuning:

```
In [0]:
```

Plot the grid

```
print(clf7.cv_results_.keys())
max_scores1 = pd.DataFrame(clf7.cv_results_).groupby(['param_num_boost_round', 'param_max_depth'])
.max().unstack()[['mean_test_score', 'mean_train_score']]

fig, ax = plt.subplots(1,2, figsize=(20,6))
sns.heatmap(max_scores1.mean_train_score, annot = True, fmt='.4g', ax=ax[0])
sns.heatmap(max_scores1.mean_test_score, annot = True, fmt='.4g', ax=ax[1])

ax[0].set_title('Train_Set')
ax[1].set_title('CV_Set')
```

```
plt.show()
dict_keys(['mean_fit_time', 'std_fit_time', 'mean_score_time', 'std_score_time',
'param_max_depth', 'param_num_boost_round', 'params', 'split0_test_score', 'split1_test_score', 's plit2_test_score', 'mean_test_score', 'std_test_score', 'rank_test_score', 'split0_train_score', '
split1_train_score', 'split2_train_score', 'mean_train_score', 'std_train_score'])
                          Train Set
                                                                                                    CV Set
                                                                                                                                    - 0.696
       0.6785
                 0.6973
                           0.7661
                                               0.953
                                                                                 0.6597
                                                                                          0.6686
                                                                                                              0.6648
                                                                                                                        0.6505
                                                           - 0.96
                                                                                                                                    - 0.688
       0.6905
                                               0.9909
                                                                                                              0.6689
                                                                                                                        0.6577
 _num_boost_round
                                                           - 0.90
                                                                                                                                    - 0.680
                                     0.9425
                                               0.9985
                                                           - 0.84
                                                                                          0.6889
                                                                                                                                    - 0.672
 param_
15
                                                           - 0.78
                                               0.9999
                                                                                          0.6942
                                                                                                    0.6874
                                     0.9696
                                                                                                                        0.6611
                                                                                                                                     - 0 664
                                                            0.72
                                                                                                                                     0.656
                                     0.9863
                                                                                          0.6973
                                                                                                                        0.6618
  2
                                                                            2
         2
                                                10
                                                                                                                         10
                       param max depth
                                                                                                param_max_depth
Best parameter
In [0]:
print(clf7.score(X_set5_train,y_train))
print(clf7.score(X_set5_test,y_test))
print(clf7.best_params_)
print(clf7.best_score_)
0.7544784550151872
0.7006129900610101
{'max_depth': 3, 'num_boost_round': 20}
0.6972502897659504
In [0]:
```

#Best tune parameters

https://scikit-

#https://scikit-

sifier.decision function

from sklearn.metrics import roc_curve, auc
from sklearn.tree import DecisionTreeClassifier
from sklearn.metrics import roc_curve, auc

y_train_pred1 = clf11.predict_proba(X_set5_train) [:,1]
y_test_pred1 = clf11.predict_proba(X_set5_test) [:,1]

train_fpr1, train_tpr1, tr_thresholds1 = roc_curve(y_train, y_train_pred1)
test fpr1, test tpr1, te thresholds1 = roc curve(v test, v test pred1)

clf11.fit(X_set5_train, y_train)

In [0]:

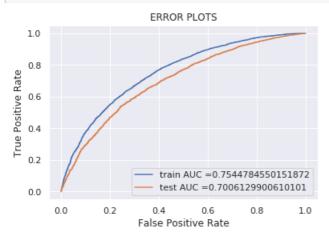
best_tune_parameters=[{'num_boost_round': [20], 'max_depth':[3] }]

 $learn.org/stable/modules/generated/sklearn.metrics.roc_curve.html \# sklearn.metrics.roc_curve.html \# sklearn.metrics.html \# sklearn.metrics.html$

clf11 = GridSearchCV(XGBoostClassifier(num_class = 2, nthread = 4),best_tune_parameters)

learn.org/stable/modules/generated/sklearn.linear model.SGDClassifier.html#sklearn.linear model.SGDClassifier.html#sklearn.html#s

```
plt.plot(train_fpr1, train_tpr1, label="train AUC ="+str(auc(train_fpr1, train_tpr1)))
plt.plot(test_fpr1, test_tpr1, label="test AUC ="+str(auc(test_fpr1, test_tpr1)))
plt.legend()
plt.xlabel("False Positive Rate")
plt.ylabel("True Positive Rate")
plt.title("ERROR PLOTS")
plt.grid(True)
plt.show()
```



Confusion matrix:

In [0]:

```
#https://www.quantinsti.com/blog/creating-heatmap-using-python-seaborn
import seaborn as sns; sns.set()

con m_train = confusion_matrix(y_train, predict(y_train_pred1, tr_thresholds1, train_fpr1, train_tp
r1))

con_m_test = confusion_matrix(y_test, predict(y_test_pred1, te_thresholds1, test_fpr1, test_tpr1))

key = (np.asarray([['TN','FP'], ['FN', 'TP']]))

fig, ax = plt.subplots(1,2, figsize=(12,5))

labels_train = (np.asarray(["{0} = {1:.2f}" .format(key, value) for key, value in zip(key.flatten(), con_m_train.flatten())])).reshape(2,2)

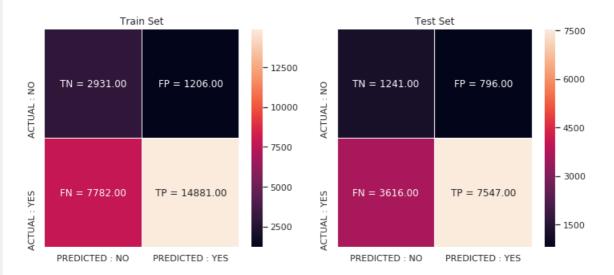
labels_test = (np.asarray(["{0} = {1:.2f}" .format(key, value) for key, value in zip(key.flatten(), con_m_test.flatten())])).reshape(2,2)

sns.heatmap(con_m_train, linewidths=.5, xticklabels=['PREDICTED : No', 'PREDICTED : YES'],
yticklabels=['ACTUAL : No', 'ACTUAL : YES'], annot = labels_train, fmt = '', ax=ax[0])
sns.heatmap(con_m_test, linewidths=.5, xticklabels=['PREDICTED : No', 'PREDICTED : YES'],
yticklabels=['ACTUAL : No', 'ACTUAL : YES'], annot = labels_test, fmt = '', ax=ax[1])

av[0] set title('Train_Set')
```

```
ax[1].set_title('Test Set')
plt.show()
```

the maximum value of tpr*(1-fpr) 0.47 for threshold 0.85 the maximum value of tpr*(1-fpr) 0.42 for threshold 0.84



3. Conclusion

In [0]:

Vectorizer	 	Model	•	num_boost_round	•		•	•
wordtovec	 	XgBoost	 	20	 	3	 -	70 j
+	+		Τ		+		+-	