

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
<code>project_id</code>	A unique identifier for the proposed project. Example: p036502
<code>project_title</code>	Title of the project. Examples: <ul style="list-style-type: none">• Art Will Make You Happy!• First Grade Fun
<code>project_grade_category</code>	Grade level of students for which the project is targeted. One of the following enumerated values: <ul style="list-style-type: none">• Grades PreK-2• Grades 3-5• Grades 6-8• Grades 9-12
<code>project_subject_categories</code>	One or more (comma-separated) subject categories for the project from the following enumerated list of values: <ul style="list-style-type: none">• Applied Learning• Care & Hunger• Health & Sports• History & Civics• Literacy & Language• Math & Science• Music & The Arts• Special Needs• Warmth Examples: <ul style="list-style-type: none">• Music & The Arts• Literacy & Language, Math & Science
<code>school_state</code>	State where school is located (Two-letter U.S. postal code). Example: WY
<code>project_subject_subcategories</code>	One or more (comma-separated) subject subcategories for the project. Examples: <ul style="list-style-type: none">• Literacy

Feature	Description
<code>project_resource_summary</code>	An explanation of the resources needed for the project. Example: <ul style="list-style-type: none"> My students need hands on literacy materials to manage sensory needs!
<code>project_essay_1</code>	First application essay*
<code>project_essay_2</code>	Second application essay*
<code>project_essay_3</code>	Third application essay*
<code>project_essay_4</code>	Fourth application essay*
<code>project_submitted_datetime</code>	Datetime when project application was submitted. Example: 2016-04-28 12:43:56.245
<code>teacher_id</code>	A unique identifier for the teacher of the proposed project. Example: bdf8baa8fedef6bfeec7ae4ff1c15c56
<code>teacher_prefix</code>	Teacher's title. One of the following enumerated values: <ul style="list-style-type: none"> nan Dr. Mr. Mrs. Ms. Teacher.
<code>teacher_number_of_previously_posted_projects</code>	Number of project applications previously submitted by the same teacher. Example: 2

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

Note: Many projects require multiple resources. The `id` value corresponds to a `project_id` 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
<code>project_is_approved</code>	A binary flag indicating whether DonorsChoose approved the project. A value of 0 indicates the project was not 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"

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.

In [1]:

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

import sqlite3
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
```

1.1 Reading Data

In [2]:

```
project_data = pd.read_csv('train_data.csv')
resource_data = pd.read_csv('resources.csv')
```

In [3]:

```
print("Number of data points in train data", project_data.shape)
print('-'*50)
print("The attributes of data :", project_data.columns.values)
```

Number of data points in train data (109248, 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']
```

In [4]:

```
from sklearn.utils import resample
project_data=resample(project_data,n_samples=50000)
```

In [5]:

```
print("Number of data points in train data", resource_data.shape)
print(resource_data.columns.values)
resource_data.head(2)
```

Number of data points in train data (1541272, 4)
['id' 'description' 'quantity' 'price']

Out[5]:

	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

In [6]:

```
price_data = resource_data.groupby('id').agg({'price':'sum', 'quantity':'sum'}).reset_index()

# join two dataframes in python:
project_data = pd.merge(project_data, price_data, on='id', how='left')
```

1.2 preprocessing of project_subject_categories

In [7]:

```
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 & Hunger"]
        if 'The' in j.split(): # this will split each of the category based on space "Math & Science"=> "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 replacing 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('&','_') # 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]))
```

1.3 preprocessing of project subject subcategories

In [8]:

```
sub_categories = 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_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 & H
unger"]
        if 'The' in j.split(): # this will split each of the category based on space "Math & Scienc
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 placing 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]))
```

1.3 Text preprocessing

In [9]:

```
# 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 [10]:

```
project_data.head(2)
```

Out[10]:

	Unnamed: 0	id	teacher_id	teacher_prefix	school_state	project_submitted_datetime	proj
0	23360	p115962	115331e4a8268e26151cbcf55b6f9e30	Ms.	CA	2016-11-06 14:02:10	Gra
1	162046	p059531	a1f80c84d1beadc2c4e57344cebb085e	Mrs.	IL	2016-08-16 23:30:24	Gra

Unnamed: 0	id	teacher_id	teacher_prefix	school_state	project_submitted_datetime	project_data
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In [11]:

```
#### 1.4.2.3 Using Pretrained Models: TFIDF weighted W2V
```

In [12]:

```
# printing some random reviews
print(project_data['essay'].values[0])
print("="*50)
print(project_data['essay'].values[150])
print("="*50)
print(project_data['essay'].values[1000])
print("="*50)
print(project_data['essay'].values[20000])
print("="*50)
#print(project_data['essay'].values[99999])
print("="*50)
```

Our school is in a unique mix of residential and high-tech industrial area. Many of our students are brand new residents of the United States, as their families have moved here for the opportunity to work for some of these large tech companies. These students are coming into fourth grade with little to no English language knowledge and they give their absolute best every day to progress to fluency and mastery of fourth grade concepts. It is an amazingly inspirational experience I get to observe and help facilitate. \r\n\r\nIn essence, the wonderful community of learners I get to see every day is comprised of members of our FUTURE. This idea and the perseverance I see in my students is what keeps me motivated to provide the best I possibly can for them. \r\n\r\n\r\nIn the past, I've had my doubts about every student needing an iPad or a Chromebook. I feel that our society relies too heavily on devices to avoid face-to-face interaction and replace the skills of parenting, among other things. However, after attending many events and being in various arenas where technology in education is the focus, I've been exposed to the great potential for technology to be an enhancement of education.\r\n\r\n\r\nIn fact, after these experiences, I believe that the MOST important place for technology to be prevalently used is within the education system. It's truthfully difficult to portray the plethora of beneficial experiences technology can bring into the classroom but to list a few:\r\n\r\nQR codes: scannable barcodes that link to any pre-selected content. Example of one use - (teacher cloning) teacher records him/herself teaching several different lessons targeting same skill but with varying levels of difficulty. Students groups scan the barcode and hear the lesson taught at a level that is accessible to them, all within the same class period.\r\n\r\nThingLink: students can virtually annotate a photo or video. Example of one use - (research and presentation skills) students research a topic, upload a photo of their work, then select various parts of the photo to attach links to \"cite\" their sources.\r\n\r\nMovieMaker: there are many apps that allow students to create stop motion videos and do some \"post-production\" editing. Example of one use - (demonstration of understanding) students learn about the rock cycle, then use drawings or objects to make a video of the process, while narrating with a list of important vocabulary that must be integrated.\r\n\r\nClassDojo: this is one of many apps that allow students to create online portfolios. Example of one use: digital portfolios to keep parents updated on what we are producing.nannan

Kindergarten students are adorable and like to show off everything they know. 1st grade students try to contain their excitement about art but need lots of routine practice. Second grade students are attentive and willing to try anything, confident, and well behaved. Most Third grade students are chatty, annoyed, and distracted by each other. I do my best to engage with their interests and build their confidence. Fourth grade students are generally trying to be super cool and pretend they already know everything; they need to be pushed a lot. Fifth grade students are great because none of them give up, they usually try their best and are resourceful and help each other. I have two mixed grade special education classes I enjoy because they challenge me to adapt and try new lessons. There is a wide range of personalities and skill levels in all of my classes. My Students need to experience and use more than 8 colors, our class packs of crayons and colored pencils provided by the school have a limited color palette - 8 colors. Our world is not made up of only 8 colors and my students need to see, and identify, and use - light brown, dark brown, sand, peach, tan, as skin tones or neutral tones. \r\n\r\n\r\nHaving a variety of color choices gets students to identify and describe specific colors and color schemes such as warm colors, cool colors, lights, darks, neutral, pastel, metallic. Texture plates will be used to teach the element of texture and create textures in student artworks. I will use the display easel for the top of my cart I use to teach 2 days a week.nannan

My students are a diverse group of kiddos from East Oakland, California. Many are learning English as a second language while soaking up all of the knowledge that first grade has to offer. \r\n\r\nThese students come to school each day ready to challenge themselves and use their growth mindset, despite the many struggles they may face at home or in their surrounding neighborhoods. \r\n\r\nWe are a Title 1 School and 100% of my student population qualifies for free and reduced school lunch. Many of our students have experienced trauma in and around their homes and neighborhoods and our school emphasizes a strong culture of social emotional learning. Our public charter school be

believes that every student has the power to go out and change the world and we emphasize this in our classrooms and school yard every day. \r\nHaving a small group set of iPod Shuffles will allow my students independent access to a variety of texts and in turn increase their reading ability, comprehension and vocabulary. They will be able to listen to a variety of 1st grade appropriate text that they may not have access to on their own or at home. \r\nMany of my students are struggling readers and providing them with the opportunity to listen to a favorite book or a story without the aid of a teacher or classmate will allow them to be curious and excited about reading without the anxiety of not being able to access the text. \r\nWith the iPod shuffle as one of our Guided Reading stations all students will be able to participate in reading a variety of grade level texts each day and building their skills as confident readers.nannan

I have an eager group of students who are so excited to come to school! Many of my students do not attend preschool so kindergarten is their first school experience. Students come in with varying abilities from being able to read and solve math problems to knowing a few letters of the alphabet and counting to 10. \r\n\r\nEvery day they come to school ready for more fun! They love hands-on activities to help them learn and practice new skills. I am hoping to be able to provide them with more centers and independent activities to continue their love of learning.We have done guided reading groups for years to meet the individual needs of ALL students. Now it is time to implement guided math groups. Just like in reading, kids come to school with varying math abilities. \r\n\r\nTo meet the needs of all of my students we need a variety of math games and activities that students can complete independently while I am working with individual or small groups of students. The items we are requesting will coincide with what the student(s) are being taught so that they can feel successful while working on a math concept/skill. \r\n\r\nI am so excited to implement guided math so that I can meet my students at their level and they can feel successful while working independently or with a partner.nannan

In [13]:

```
# 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"\ 've", " have", phrase)
    phrase = re.sub(r"\ 'm", " am", phrase)
    return phrase
```

In [14]:

```
sent = decontracted(project_data['essay'].values[20000])
print(sent)
print("="*50)
```

I have an eager group of students who are so excited to come to school! Many of my students do not attend preschool so kindergarten is their first school experience. Students come in with varying abilities from being able to read and solve math problems to knowing a few letters of the alphabet and counting to 10. \r\n\r\nEvery day they come to school ready for more fun! They love hands-on activities to help them learn and practice new skills. I am hoping to be able to provide them with more centers and independent activities to continue their love of learning.We have done guided reading groups for years to meet the individual needs of ALL students. Now it is time to implement guided math groups. Just like in reading, kids come to school with varying math abilities. \r\n\r\nTo meet the needs of all of my students we need a variety of math games and activities that students can complete independently while I am working with individual or small groups of students. The items we are requesting will coincide with what the student(s) are being taught so that they can feel successful while working on a math concept/skill. \r\n\r\nI am so excited to implement guided math so that I can meet my students at their level and they can feel successful while working independently or with a partner.nannan

In [15]:

```
# \r\n\r\n\t remove from string python: http://texthandler.com/info/remove-line-breaks-python/
```

```
# If you're removing from string python: http://cexandrier.com/info/remove-line-breaks-python/
sent = sent.replace('\\r', ' ')
sent = sent.replace('\\n', ' ')
sent = sent.replace('\\n', ' ')
print(sent)
```

I have an eager group of students who are so excited to come to school! Many of my students do not attend preschool so kindergarten is their first school experience. Students come in with varying abilities from being able to read and solve math problems to knowing a few letters of the alphabet and counting to 10. Every day they come to school ready for more fun! They love hands-on activities to help them learn and practice new skills. I am hoping to be able to provide them with more centers and independent activities to continue their love of learning. We have done guided reading groups for years to meet the individual needs of ALL students. Now it is time to implement guided math groups. Just like in reading, kids come to school with varying math abilities. To meet the needs of all of my students we need a variety of math games and activities that students can complete independently while I am working with individual or small groups of students. The items we are requesting will coincide with what the student(s) are being taught so that they can feel successful while working on a math concept/skill. I am so excited to implement guided math so that I can meet my students at their level and they can feel successful while working independently or with a partner. nannan

In [16]:

```
#remove spacial character: https://stackoverflow.com/a/5843547/4084039
sent = re.sub('[^A-Za-z0-9]+', ' ', sent)
print(sent)
```

I have an eager group of students who are so excited to come to school Many of my students do not attend preschool so kindergarten is their first school experience Students come in with varying abilities from being able to read and solve math problems to knowing a few letters of the alphabet and counting to 10 Every day they come to school ready for more fun They love hands on activities to help them learn and practice new skills I am hoping to be able to provide them with more centers and independent activities to continue their love of learning We have done guided reading groups for years to meet the individual needs of ALL students Now it is time to implement guided math groups Just like in reading kids come to school with varying math abilities To meet the needs of all of my students we need a variety of math games and activities that students can complete independently while I am working with individual or small groups of students The items we are requesting will coincide with what the student s are being taught so that they can feel successful while working on a math concept skill I am so excited to implement guided math so that I can meet my students at their level and they can feel successful while working independently or with a partner nannan

In [17]:

```
# 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", 'your', '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', 'each', \
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', "doesn't", 'hadn', \
esn't", '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"]
```



```
# Combining all the above stundents
from tqdm import tqdm
preprocessed_essays = []
# tqdm is for printing the status bar
for sentence in tqdm(project_data['essay'].values):
    sent = decontracted(sentence)
    sent = sent.replace('\\r', ' ')
    sent = sent.replace('\\n', ' ')
    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 not in stopwords)
    preprocessed_essays.append(sent.lower().strip())
```

In [19]:

Out [19] :

'i eager group students excited come school many students not attend preschool kindergarten first school experience students come varying abilities able read solve math problems knowing letters alphabet counting 10 every day come school ready fun they love hands activities help learn practice new skills i hoping able provide centers independent activities continue love learning we done guided reading groups years meet individual needs all students now time implement guided math groups just like reading kids come school varying math abilities to meet needs students need variety math games activities students complete independently i working individual small groups students the items requesting coincide student taught feel successful working math concept skill i excited implement guided math i meet students level feel successful working independently partner nannan'

```
project_data['clean_essays'] = preprocessed_essays
project_data.drop(['project_essay_1'], axis=1, inplace=True)
project_data.drop(['project_essay_2'], axis=1, inplace=True)
project_data.drop(['project_essay_3'], axis=1, inplace=True)
project_data.drop(['project_essay_4'], axis=1, inplace=True)
```

In [21]:

```
# similarly you can preprocess the titles also
from tqdm import tqdm
preprocessed_title= []
# tqdm is for printing the status bar
for sentence in tqdm(project_data['project_title'].values):
    sent = decontracted(sentence)
    sent = sent.replace('\r', ' ')
    sent = sent.replace('\n', ' ')
    sent = sent.replace('\^n', ' ')
    sent = sent.replace('\nan', ' ')
    sent = re.sub('[^A-Za-z0-9]+', ' ', sent)
    # https://gist.github.com/sebleier/554280
    sent = ' '.join(e for e in sent.split() if e not in stopwords)
    preprocessed_title.append(sent.lower().strip())
```

In [23]:

```
project_data['clean_title'] = preprocessed_title
```

```
project_data['clean_title'] = preprocessed_title
```

In [27]:

```
y = project_data['project_is_approved'].values  
project_data.drop(['project_is_approved'], axis=1, inplace=True)  
X = project_data
```

In [28]:

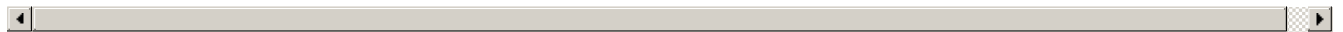
```
from sklearn.model_selection import train_test_split  
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.33, stratify=y)  
X_train, X_cv, y_train, y_cv = train_test_split(X_train, y_train, test_size=0.33, stratify=y_train)
```

In [29]:

```
print(X_train.shape, y_train.shape)  
print(X_cv.shape, y_cv.shape)  
print(X_test.shape, y_test.shape)  
  
print("="*100)
```

```
(22445, 17) (22445,)  
(11055, 17) (11055,)  
(16500, 17) (16500,)
```

=====



In [30]:

```
i="Divya is nan a good girl"  
list_i = i.replace('^nan',' ')  
print(list_i)
```

Divya is nan a good girl

Checking for Sample data

In [31]:

```
list_word = ["ABC", "PQR", "DEF"]  
combined_essay = ["ABC DEF IJK PQR", "PQR KLM OPQ", "LMN PQR XYZ ABC DEF PQR ABC"]
```

In [32]:

```
arr = []  
for i in combined_essay:  
    arr.append(i.split())
```

In [33]:

```
arr
```

Out[33]:

```
[['ABC', 'DEF', 'IJK', 'PQR'],  
 ['PQR', 'KLM', 'OPQ'],  
 ['LMN', 'PQR', 'XYZ', 'ABC', 'DEF', 'PQR', 'ABC']]
```

In [34]:

```
co_variance = pd.DataFrame(np.zeros(9).reshape(3,3))  
co_variance.columns = list_word  
co_variance.index = list_word
```

In [35]:

```
co_variance
```

Out[35]:

	ABC	PQR	DEF
ABC	0.0	0.0	0.0
PQR	0.0	0.0	0.0
DEF	0.0	0.0	0.0

In [36]:

```
%%time
for i,w in enumerate(list_word):
    #print(w,"word")
    for line in arr:
        #print(line,"line")

        if w in line:
            list_search = list_word[i+1:3]
            #print(list_search,"list_search")
            #print("It is in list")
            for search in list_search:
                #print(search)
                if search in line:
                    count = 0
                    flag = 0
                    indices_search = [i for i, x in enumerate(line) if x == search]
                    indices_word = [i for i, x in enumerate(list_word) if x == w]
                    #print(indices_search , indices_word)
                    for j in indices_search:
                        for k in indices_word:
                            if abs(j-k)<=2:
                                count += 1
                                flag=1

                    if flag==1:
                        co_variance.loc[w ,search] += count
                        co_variance.loc[search ,w] += count
            #print(co_variance)
```

Wall time: 0 ns

In [37]:

```
co_variance
```

Out[37]:

	ABC	PQR	DEF
ABC	0.0	3.0	3.0
PQR	3.0	0.0	2.0
DEF	3.0	2.0	0.0

In [38]:

```
print(len(preprocessed_essays))
N =(len(preprocessed_title))
```

50000

In [39]:

```
DF = {}

for i in range(N):
    tokens = preprocessed_essays[i]
    for w in tokens:
        try:
            DF[w].add(i)
        except:
            DF[w] = {i}

    tokens = preprocessed_title[i]
    for w in tokens:
        try:
            DF[w].add(i)
        except:
            DF[w] = {i}
for i in DF:
    DF[i] = len(DF[i])
```

In [40]:

```
train_essays = X_train['clean_essays']
train_title = X_train['clean_title']
test_essays = X_test['clean_essays']
test_title = X_test['clean_title']
cv_essays = X_cv['clean_essays']
cv_title = X_cv['clean_title']
```

In [41]:

```
train_essays.iloc[100].find("nan")
```

Out[41]:

909

2.1 Selecting top 2000 words from `essay` and `project_title`

In [42]:

```
train_essaycombined = train_essays + train_title
test_essaycombined = test_essays + test_title
cv_essaycombined = cv_essays + cv_title
```

In [45]:

```
%%time
# reference https://towardsdatascience.com/tf-idf-for-document-ranking-from-scratch-in-python-on-real-world-dataset-796d339a4089

def word_vector(processed_combined):
    DF = {}
    for i,w in enumerate(processed_combined):
        list_words = w.split()
        #print(len(list_words))
        tokens = list(set(list_words))
        #print(tokens)
        for j in tokens:
            #print(j)
            try:
                DF[j].add(i)
            except:
                DF[j] = {i}
    idf = {}
```

```

for i in DF:
    DF[i] = len(DF[i])
    df = DF[i]
    idf[i] = np.log((N+1)/(df+1))
#print((idf))
print(len(idf))
dictionary_idf =sorted(idf.items(),reverse=True, key=lambda item: item[1])
dictionary_idf = dictionary_idf[:2000]
list_word =[]
for i in dictionary_idf:
    list_word .append(i[0])
print(len(list_word))
return list_word

#print(DF)

```

Wall time: 0 ns

In [46]:

```

train_essaycombined = [re.sub(r"nan" , "",i) for i in train_essaycombined ]
cv_essaycombined = [re.sub(r"nan" , "",i) for i in cv_essaycombined ]
test_essaycombined = [re.sub(r"nan" , "",i) for i in test_essaycombined ]

```

In [47]:

```
test_essaycombined[100].find("nan")
```

Out[47]:

-1

In [48]:

```

trainessays_words= word_vector(train_essaycombined)
testessays_words= word_vector(test_essaycombined)
cvessays_words= word_vector(cv_essaycombined)

```

```

31180
2000
28004
2000
23706
2000

```

In [49]:

```

# removing nan from words
trainessays_words = [re.sub(r"nan" , "",i) for i in trainessays_words ]
cvessays_words = [re.sub(r"nan" , "",i) for i in cvessays_words ]
testessays_words = [re.sub(r"nan" , "",i) for i in testessays_words ]

```

In [50]:

```
trainessays_words[:10]
```

Out[50]:

```

['northridge',
 'guin',
 'distractive',
 'skipper',
 'ballance',
 'emboldens',
 'absorption',
 'enlish',
 'haitain',
 'mementos']

```

2.2 Computing Co-occurrence matrix

In [51]:

```
def create_covariance(list_word ,essaycombined):
    co_variance = pd. DataFrame (np.zeros(4000000).reshape(2000,2000))
    co_variance .columns = list_word
    co_variance . index = list_word
    arr =[]
    for i in essaycombined:
        arr .append (i.split())
    for i,w in enumerate(list_word):
        for line in arr:
            if w in line:
                list_search = list_word[i+1:2000]
                for search in list_search:
                    if search in line:
                        count = 0
                        flag = 0
                        indices_search = [i for i, x in enumerate(line) if x == search]
                        indices_word = [i for i, x in enumerate(line) if x == w]
                        #print(indices_search , indices_word)
                        for j in indices_search:
                            for k in indices_word:
                                if abs(j-k)<=5:
                                    count += 1
                                    flag=1

                if flag==1:
                    co_variance .loc[w ,search] += count
                    co_variance.loc[search ,w] += count

    return co_variance
```

In [52]:

```
%%time
X_train_essays = create_covariance(trainessays_words , train_essaycombined)
X_cv_essays = create_covariance(cvessays_words , cv_essaycombined)
X_test_essays = create_covariance(testessays_words , test_essaycombined)
```

Wall time: 4min 13s

Truncated SVD matrix

In [53]:

```
from sklearn.decomposition import TruncatedSVD
from scipy import sparse
variance_list=[]
list_i=[10,100,200,300,500,750,1000,1300,1500,1750]
variance_list_test=[]
for i in (list_i):
    svd = TruncatedSVD(n_components=i, n_iter=3, random_state=42)
    svd.fit(sparse.csr_matrix(X_train_essays) )
    variance_list .append(svd.explained_variance_.sum())
    print(i,svd.explained_variance_.sum())
for i in (list_i):
    svd = TruncatedSVD(n_components=i, n_iter=3, random_state=42)
    svd.fit(sparse.csr_matrix(X_test_essays) )
    variance_list_test .append(svd.explained_variance_.sum())
    print(i,svd.explained_variance_.sum())
```

```
10 0.13666123135623723
100 0.2917780274606508
200 0.3428743916414434
300 0.3929614033742139
500 0.43656449999999998
750 0.43656450000000013
1000 0.43656450000000003
1300 0.43656450000000074
1500 0.43656449999999981
```

```

1750 0.43656450000000017
10 0.0521529563476717
100 0.1622770658693448
200 0.21253685110080944
300 0.2625736715045041
500 0.3267585
750 0.32675849999999994
1000 0.32675850000000006
1300 0.32675850000000008
1500 0.32675849999999923
1750 0.32675849999999996

```

In [54]:

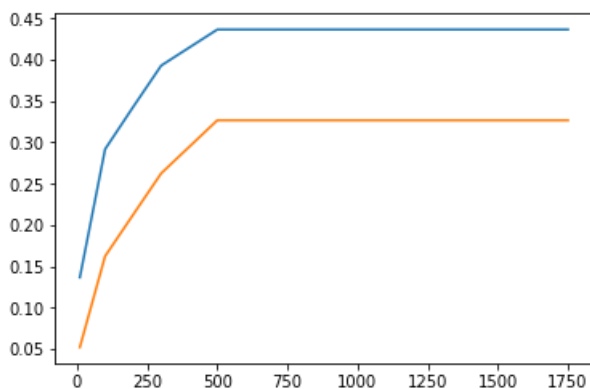
```

plt.plot(list_i, variance_list)
plt.plot(list_i, variance_list_test)

```

Out[54]:

[<matplotlib.lines.Line2D at 0x2a6f96c6da0>]



In [55]:

```

svd = TruncatedSVD(n_components=500, n_iter=3, random_state=18)
X_train_essays_SVD=svd.fit_transform(sparse.csr_matrix(X_train_essays))
X_test_essays_SVD=svd.fit_transform(sparse.csr_matrix(X_test_essays))

```

In [56]:

```
type(X_train_essays_SVD)
```

Out[56]:

numpy.ndarray

In [57]:

```
X_test_essays_SVD.shape
```

Out[57]:

(2000, 500)

In [58]:

```

list_holding = []
for i in train_essaycombined:
    list_holding.append(len(i.split()))
X_train['word_count'] = list_holding

list_holding = []
for i in test_essaycombined:
    list_holding.append(len(i.split()))
X_test['word_count'] = list_holding

```

```

list_holding = []
for i in cv_essaycombined:
    list_holding.append(len(i.split()))
X_cv['word_count'] = list_holding

list_holding = []
for i in train_title:
    list_holding.append(len(i.split()))
X_train['title_count'] = list_holding

list_holding = []
for i in test_title:
    list_holding.append(len(i.split()))
X_test['title_count'] = list_holding

list_holding = []
for i in cv_title:
    list_holding.append(len(i.split()))
X_cv['title_count'] = list_holding

```

In [59]:

```

print((X_train['word_count'][:100]))
print((X_train['title_count'][:100]))

```

```

42636    116
19749    147
17400    149
43843    192
15202    178
29114    150
17774    126
46556    190
21883    137
33785    126
41005    111
47845    150
28380    138
29334    119
11711    190
6220     129
40424    126
3939     107
40173    145
49046    246
21188    183
7867     242
18012    143
36102    123
23345    146
45308    101
30942    145
44096    200
10916    224
5398     136
...
7067     155
43154    215
7378     106
20137    147
40389    126
9532     128
34222    149
30302    135
46708    168
30193    240
738      173
47614    195
44740    116
1468     102
32299    113
30608    132
28503    143
9117     142
1689     241

```



```

...
14699    256
8444     138
37105    136
11551    166
13322    223
14218    209
40192    168
42440    220
37366    118
15465    200
31227    177
Name: word_count, Length: 100, dtype: int64
42636     3
19749     9
17400     5
43843    10
15202    10
29114     4
17774     3
46556     6
21883     3
33785     5
41005     3
47845     5
28380     2
29334     4
11711     4
6220      5
40424     7
3939      2
40173     3
49046     8
21188     5
7867      6
18012     4
36102     4
23345     3
45308     2
30942     4
44096     6
10916     5
5398      5
...
7067      4
43154     5
7378      3
20137     5
40389     7
9532      4
34222     2
30302     7
46708     5
30193     5
738       7
47614     3
44740     3
1468      3
32299     3
30608     4
28503     5
9117      4
1689      4
14699    10
8444      3
37105     5
11551     5
13322     5
14218     5
40192     5
42440     6
37366     9
15465     4
31227     4
Name: title_count, Length: 100, dtype: int64

```

In [60]:

```
project_data.columns
```

Out[60]:

```
Index(['Unnamed: 0', 'id', 'teacher_id', 'teacher_prefix', 'school_state',  
      'project_submitted_datetime', 'project_grade_category', 'project_title',  
      'project_resource_summary',  
      'teacher_number_of_previously_posted_projects', 'price', 'quantity',  
      'clean_categories', 'clean_subcategories', 'essay', 'clean_essays',  
      'clean_title'],  
      dtype='object')
```

In [61]:

```
from sklearn.preprocessing import OneHotEncoder, LabelEncoder  
def OneHotEncoder_find(category):  
    le = LabelEncoder()  
    enc = OneHotEncoder()  
    train_ohe = enc.fit_transform(le.fit_transform(X_train[category].astype(str).values).reshape(-1,  
    , 1))  
    test_ohe = enc.fit_transform(le.fit_transform(X_test[category].astype(str).values).reshape(-1,  
    1))  
    #train_ohe = enc.fit_transform(X_train[category])  
    #test_ohe = enc.fit_transform(X_test[category])  
    return train_ohe ,test_ohe
```

In [62]:

```
X_train_clean_cat,X_te_clean_cat = OneHotEncoder_find('clean_categories')  
X_train_clean_subcat,X_te_clean_subcat = OneHotEncoder_find('clean_subcategories')  
X_train_grade ,X_test_grade = OneHotEncoder_find('project_grade_category')  
X_train_teacher,X_test_teacher = OneHotEncoder_find('teacher_prefix')  
X_train_state,X_test_state = OneHotEncoder_find('school_state')
```

In [63]:

```
print((X_train_clean_cat).shape)  
print(X_te_clean_cat.shape)  
print(X_test.shape)
```

```
(22445, 49)  
(16500, 47)  
(16500, 19)
```

In [75]:

```
essays =X_train['clean_essays']  
essays = essays[:2000]  
print(type(essays))  
X_train = X_train[:2000]  
X_test = X_test[:2000]
```

```
<class 'pandas.core.series.Series'>
```

Computing Sentiment Scores

In [94]:

```
%%time  
import nltk  
from nltk.sentiment.vader import SentimentIntensityAnalyzer  
list_neg = []  
list_neu = []  
list_pos = []  
list_comp = []  
essays = X_train['clean_essays']  
for i in essays[:2000]:  
  
    sid = SentimentIntensityAnalyzer()
```

```

ss = sid.polarity_scores(i)

for k in ss:
    #print('{0}: {1}'.format(k, ss[k]), end='')
    #neg: 0.01, neu: 0.745, pos: 0.245, compound: 0.9975
    if k=='neg':
        list_neg.append(ss[k])
    elif k=='neu':
        list_neu.append(ss[k])
    elif k=='pos':
        list_pos.append(ss[k])
    elif k=='compound':
        list_comp.append(ss[k])
X_train['sent_neg'] = list_neg
X_train['sent_neu'] = list_neu
X_train['sent_pos'] = list_pos
X_train['sent_comp'] = list_comp

list_neg=[]
list_neu=[]
list_pos=[]
list_comp=[]
essays = X_test['clean_essays']
for i in essays[:2000]:

    sid = SentimentIntensityAnalyzer()
    ss = sid.polarity_scores(i)

    for k in ss:
        #print('{0}: {1}'.format(k, ss[k]), end='')
        #neg: 0.01, neu: 0.745, pos: 0.245, compound: 0.9975
        if k=='neg':
            list_neg.append(ss[k])
        elif k=='neu':
            list_neu.append(ss[k])
        elif k=='pos':
            list_pos.append(ss[k])
        elif k=='compound':
            list_comp.append(ss[k])
X_test['sent_neg'] = list_neg
X_test['sent_neu'] = list_neu
X_test['sent_pos'] = list_pos
X_test['sent_comp'] = list_comp

```

Wall time: 51.3 s

In [86]:

```
X_train.columns
```

Out[86]:

```

Index(['Unnamed: 0', 'id', 'teacher_id', 'teacher_prefix', 'school_state',
      'project_submitted_datetime', 'project_grade_category', 'project_title',
      'project_resource_summary',
      'teacher_number_of_previously_posted_projects', 'price', 'quantity',
      'clean_categories', 'clean_subcategories', 'essay', 'clean_essays',
      'clean_title', 'word_count', 'title_count', 'sent_neg', 'sent_neu',
      'sent_pos', 'sent_comp'],
      dtype='object')

```

In [104]:

```

from sklearn.preprocessing import StandardScaler
standard_vec = StandardScaler(with_mean = False)
# this will rise an error Expected 2D array, got 1D array instead:
# array=[105.22 215.96 96.01 ... 368.98 80.53 709.67].
# Reshape your data either using
# array.reshape(-1, 1) if your data has a single feature
# array.reshape(1, -1) if it contains a single sample.
standard_vec.fit(X_train['quantity'].values.reshape(-1,1))

X_train_qty_std = standard_vec.transform(X_train['quantity'].values.reshape(-1,1))

```

```
X_cv_qty_std = standard_vec.transform(X_cv['quantity'].values.reshape(-1,1))
X_test_qty_std = standard_vec.transform(X_test['quantity'].values.reshape(-1,1))

print("After vectorizations")
print(X_train_qty_std.shape, y_train.shape)
print(X_cv_qty_std.shape, y_cv.shape)
print(X_test_qty_std.shape, y_test.shape)
print("=="*100)
```

After vectorizations
(2000, 1) (22445,)
(11055, 1) (11055,)
(2000, 1) (16500,)

In [105]:

```
from sklearn.preprocessing import StandardScaler
standard_vec = StandardScaler(with_mean = False)
# this will rise an error Expected 2D array, got 1D array instead:
# array=[105.22 215.96 96.01 ... 368.98 80.53 709.67].
# Reshape your data either using
# array.reshape(-1, 1) if your data has a single feature
# array.reshape(1, -1) if it contains a single sample.
standard_vec.fit(X_train['word_count'].values.reshape(-1,1))

X_train_word_std = standard_vec.transform(X_train['word_count'].values.reshape(-1,1))
X_cv_word_std = standard_vec.transform(X_cv['word_count'].values.reshape(-1,1))
X_test_word_std = standard_vec.transform(X_test['word_count'].values.reshape(-1,1))

print("After vectorizations")
print(X_train_word_std.shape, y_train.shape)
print(X_cv_word_std.shape, y_cv.shape)
print(X_test_word_std.shape, y_test.shape)
```

After vectorizations
(2000, 1) (22445,)
(11055, 1) (11055,)
(2000, 1) (16500,)

In [106]:

```
from sklearn.preprocessing import StandardScaler
standard_vec = StandardScaler(with_mean = False)
# this will rise an error Expected 2D array, got 1D array instead:
# array=[105.22 215.96 96.01 ... 368.98 80.53 709.67].
# Reshape your data either using
# array.reshape(-1, 1) if your data has a single feature
# array.reshape(1, -1) if it contains a single sample.
standard_vec.fit(X_train['word_count'].values.reshape(-1,1))

X_train_title_std = standard_vec.transform(X_train['title_count'].values.reshape(-1,1))
X_cv_title_std = standard_vec.transform(X_cv['title_count'].values.reshape(-1,1))
X_test_title_std = standard_vec.transform(X_test['title_count'].values.reshape(-1,1))

print("After vectorizations")
print(X_train_title_std.shape, y_train.shape)
print(X_cv_title_std.shape, y_cv.shape)
print(X_test_title_std.shape, y_test.shape)
```

After vectorizations
(2000, 1) (22445,)
(11055, 1) (11055,)
(2000, 1) (16500,)

In [95]:

```
from sklearn.preprocessing import StandardScaler
standard_vec = StandardScaler(with_mean = False)

standard_vec.fit(X_train['sent_neg'].values.reshape(-1,1))
#print(X_train['sent_neg'])
```

```
#print(X_train['sent_neg'])

X_train_sent_neg_std = standard_vec.transform(X_train['sent_neg'].values.reshape(-1,1))
#X_cv_title_std = standard_vec.transform(X_cv['sent_neg'].values.reshape(-1,1))
X_test_sent_neg_std = standard_vec.transform(X_test['sent_neg'].values.reshape(-1,1))

print("After vectorizations")
print(X_train_sent_neg_std.shape, y_train.shape)
#print(X_cv_title_std.shape, y_cv.shape)
print(X_test_sent_neg_std.shape, y_test.shape)
```

After vectorizations

```
(2000, 1) (22445,)
(2000, 1) (16500,)
```

In [96]:

```
from sklearn.preprocessing import StandardScaler
standard_vec = StandardScaler(with_mean = False)
# this will rise an error Expected 2D array, got 1D array instead:
# array=[105.22 215.96 96.01 ... 368.98 80.53 709.67].
# Reshape your data either using
# array.reshape(-1, 1) if your data has a single feature
# array.reshape(1, -1) if it contains a single sample.
standard_vec.fit(X_train['sent_neu'].values.reshape(-1,1))

X_train_sent_neu_std = standard_vec.transform(X_train['sent_neu'].values.reshape(-1,1))
#X_cv_title_std = standard_vec.transform(X_cv['sent_neg'].values.reshape(-1,1))
X_test_sent_neu_std = standard_vec.transform(X_test['sent_neu'].values.reshape(-1,1))

print("After vectorizations")
print(X_train_sent_neu_std.shape, y_train.shape)
#print(X_cv_title_std.shape, y_cv.shape)
print(X_test_sent_neu_std.shape, y_test.shape)
```

After vectorizations

```
(2000, 1) (22445,)
(2000, 1) (16500,)
```

In [97]:

```
from sklearn.preprocessing import StandardScaler
standard_vec = StandardScaler(with_mean = False)
# this will rise an error Expected 2D array, got 1D array instead:
# array=[105.22 215.96 96.01 ... 368.98 80.53 709.67].
# Reshape your data either using
# array.reshape(-1, 1) if your data has a single feature
# array.reshape(1, -1) if it contains a single sample.
standard_vec.fit(X_train['sent_pos'].values.reshape(-1,1))

X_train_sent_pos_std = standard_vec.transform(X_train['sent_pos'].values.reshape(-1,1))
#X_cv_title_std = standard_vec.transform(X_cv['sent_neg'].values.reshape(-1,1))
X_test_sent_pos_std = standard_vec.transform(X_test['sent_pos'].values.reshape(-1,1))

print("After vectorizations")
print(X_train_sent_pos_std.shape, y_train.shape)
#print(X_cv_title_std.shape, y_cv.shape)
print(X_test_sent_pos_std.shape, y_test.shape)
```

After vectorizations

```
(2000, 1) (22445,)
(2000, 1) (16500,)
```

In [98]:

```
from sklearn.preprocessing import StandardScaler
standard_vec = StandardScaler(with_mean = False)
# this will rise an error Expected 2D array, got 1D array instead:
# array=[105.22 215.96 96.01 ... 368.98 80.53 709.67].
# Reshape your data either using
# array.reshape(-1, 1) if your data has a single feature
# array.reshape(1, -1) if it contains a single sample.
standard_vec.fit(X_train['sent_comp'].values.reshape(-1,1))
```

```

X_train_sent_comp_std = standard_vec.transform(X_train['sent_comp'].values.reshape(-1,1))
#X_cv_title_std = standard_vec.transform(X_cv['sent_neg'].values.reshape(-1,1))
X_test_sent_comp_std = standard_vec.transform(X_test['sent_comp'].values.reshape(-1,1))

print("After vectorizations")
print(X_train_sent_comp_std.shape, y_train.shape)
#print(X_cv_title_std.shape, y_cv.shape)
print(X_test_sent_comp_std.shape, y_test.shape)

```

After vectorizations
(2000, 1) (22445,)
(2000, 1) (16500,)

In [107]:

```

from sklearn.preprocessing import StandardScaler
standard_vec = StandardScaler(with_mean = False)
# this will rise an error Expected 2D array, got 1D array instead:
# array=[105.22 215.96 96.01 ... 368.98 80.53 709.67].
# Reshape your data either using
# array.reshape(-1, 1) if your data has a single feature
# array.reshape(1, -1) if it contains a single sample.
standard_vec.fit(X_train['price'].values.reshape(-1,1))

X_train_price_std = standard_vec.transform(X_train['price'].values.reshape(-1,1))
#X_cv_title_std = standard_vec.transform(X_cv['sent_neg'].values.reshape(-1,1))
X_test_price_std = standard_vec.transform(X_test['price'].values.reshape(-1,1))

print("After vectorizations")
print(X_train_price_std.shape, y_train.shape)
#print(X_cv_title_std.shape, y_cv.shape)
print(X_test_price_std.shape, y_test.shape)

```

After vectorizations
(2000, 1) (22445,)
(2000, 1) (16500,)

In [113]:

```

from sklearn.preprocessing import StandardScaler
standard_vec = StandardScaler(with_mean = False)
# this will rise an error Expected 2D array, got 1D array instead:
# array=[105.22 215.96 96.01 ... 368.98 80.53 709.67].
# Reshape your data either using
# array.reshape(-1, 1) if your data has a single feature
# array.reshape(1, -1) if it contains a single sample.
standard_vec.fit(X_train['teacher_number_of_previously_posted_projects'].values.reshape(-1,1))

X_train_projects_std =
standard_vec.transform(X_train['teacher_number_of_previously_posted_projects'].values.reshape(-1,1))
#X_cv_title_std = standard_vec.transform(X_cv['sent_neg'].values.reshape(-1,1))
X_test_projects_std = standard_vec.transform(X_test['teacher_number_of_previously_posted_projects'].values.reshape(-1,1))

print("After vectorizations")
print(X_train_projects_std.shape, y_train.shape)
#print(X_cv_title_std.shape, y_cv.shape)
print(X_test_projects_std.shape, y_test.shape)

```

After vectorizations
(2000, 1) (22445,)
(2000, 1) (16500,)

In [114]:

```
X_test_essays_SVD .shape
```

Out[114]:
(2000, 500)

In [115]:

```
X_train = X_train[:2000]
X_test = X_test[:2000]
```

In [116]:

```
X_train_essays_SVD
```

Out[116]:

```
array([[ -3.33887652e-19,  4.44503951e-16,  5.67552032e-18, ...,
         1.80933192e-19, -1.06062521e-19, -8.59491669e-20],
       [ -3.03959029e-19,  3.41863380e-19,  4.63179042e-18, ...,
        -2.27334821e-19,  2.32524610e-19,  1.14034634e-20],
       [ -3.60316517e-17, -4.00849435e-17,  5.87449950e-16, ...,
        -1.38302694e-18,  5.69065557e-18, -2.30140231e-18],
       ...,
       [ 0.00000000e+00,  0.00000000e+00, -0.00000000e+00, ...,
        -0.00000000e+00,  0.00000000e+00, -0.00000000e+00],
       [ 2.89290703e-19,  2.31645313e-18,  2.76602866e-17, ...,
        -1.44411938e-33, -8.30367625e-34,  1.57348462e-33],
       [ 1.62935225e-18,  9.69692698e-17, -4.59452045e-17, ...,
        -5.70982244e-34, -6.85356917e-34,  2.09024467e-33]])
```

In [117]:

```
X_test_essays_SVD
```

Out[117]:

```
array([[ -4.46884382e-16,  1.28697792e-16,  7.64898328e-17, ...,
        -2.78663516e-33,  1.33079768e-33,  2.29972696e-33],
       [ -2.21567249e-15, -3.26383705e-16,  2.39813310e-16, ...,
        2.36635427e-33,  4.67074814e-34,  1.56854383e-33],
       [ 2.21020774e-16, -7.51967564e-17, -4.16496121e-16, ...,
        8.42614380e-18, -9.43694108e-19,  1.02335159e-18],
       ...,
       [ 6.86393716e-17, -9.98540081e-17,  4.04165417e-17, ...,
        1.24315853e-33,  8.72804757e-35, -1.90206103e-33],
       [ -0.00000000e+00, -0.00000000e+00, -0.00000000e+00, ...,
        -0.00000000e+00,  0.00000000e+00,  0.00000000e+00],
       [ 1.88488661e-16,  7.26874525e-17, -1.18725103e-16, ...,
        3.66197206e-33,  2.15256608e-33, -8.84373585e-34]])
```

Concatinating all features

In [154]:

```
X_train_clean_cat = (X_train_clean_cat[:2000,:47])
X_train_clean_subcat = X_train_clean_subcat[:2000,:319]
X_train_teacher = X_train_teacher[:2000,:5]
```

In [155]:

```
from scipy.sparse import hstack
from scipy.sparse import hstack
X_tr = hstack((X_train_essays_SVD,X_train_clean_cat[:2000,:],X_train_clean_subcat[:2000,:], X_train_state[:2000,:], X_train_teacher[:2000,:], X_train_grade[:2000,:], X_train_price_std[:2000,:],X_train_projects_std[:2000,:],X_train_qty_std[:2000,:],X_train_word_std[:2000,:],X_train_title_std[:2000,:],X_train_sent_comp_std,X_train_sent_pos_std,X_train_sent_neu_std,X_train_sent_neg_std)).tocsr()
#X_cr =
hstack((X_cv_essays_SVD,X_cv_titles_tfidf,X_cv_summary_tfidf,X_cv_clean_cat_ohe,X_cv_clean_subcat_ohe,X_cv_state_ohe, X_cv_teacher_ohe, X_cv_grade_ohe,
X_cv_price_std,X_cv_projects_std,X_cv_qty_std)).tocsr()
X_te = hstack((X_test_essays_SVD,X_te_clean_cat[:2000,:],X_te_clean_subcat[:2000,:], X_test_state[:2000,:], X_test_teacher[:2000,:], X_test_grade[:2000,:], X_test_price_std[:2000,:],X_test_projects_std[:2000,:],X_test_qty_std[:2000,:],X_test_word_std[:2000,:],X_test_title_std[:2000,:],X_test_s
```

```
ent_comp_std,X_test_sent_pos_std,X_test_sent_neu_std,X_test_sent_neg_std)).tocsr()
```

```
y_train = y_train[:2000]
y_test = y_test[:2000]
print("Final Data matrix")
print(X_tr.shape, y_train.shape)
#print(X_cr.shape, y_cv.shape)
print(X_te.shape, y_test.shape)
print("="*100)
```

```
Final Data matrix
(2000, 935) (2000,)
(2000, 935) (2000,)
```

Assignment 11: TruncatedSVD

- **step 1** Select the top 2k words from essay text and project_title (concatenate essay text with project title and then find the top 2k words) based on their `idf`` values
- **step 2** Compute the co-occurrence matrix with these 2k words, with window size=5 ([ref](#))
- **step 3** Use [TruncatedSVD](#) on calculated co-occurrence matrix and reduce its dimensions, choose the number of components (`n_components`) using [elbow method](#)
 - 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

2.5 Apply XGBoost on the Final Features from the above section

In [157]:

```
%%time
import warnings
warnings.filterwarnings("ignore")
# https://scikit-learn.org/stable/modules/generated/sklearn.model_selection.GridSearchCV.html
from sklearn.model_selection import GridSearchCV
import math
import xgboost as xgb
from sklearn.metrics import roc_auc_score
from sklearn import metrics
from sklearn import cross_validation
```


Wall time: 0 ns

In [158]:

```
X_tr = X_tr.toarray()
X_te = X_te.toarray()
```

In [159]:

```
type(y_train)
```

Out[159]:

```
numpy.ndarray
```

In [163]:

```
import sys
import math

import numpy as np
from sklearn.grid_search import GridSearchCV
from sklearn.metrics import roc_auc_score

# you might need to install this one
import xgboost as xgb

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,
                             verbose_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
```

In [185]:

```
%%time
clf = XGBoostClassifier(eval_metric = 'auc', num_class = 2, nthread = 4,)
#####
#           Change from here                                     #
#####
```

```

parameters = {
    #'eta': [0.05,0.6,0.8, 0.1,0.2,0.5,.4, 0.3],
    'learning_rate': [0.1, 0.8, 0.4, 0.5 ,0.6,0.2],
    #'num_boost_round': [100, 250, 500],
    #'max_depth': [3, 5 ,6, 9, 12],
    'subsample': [0.1,0.5,0.7,0.9, 1.0],
}

clf = GridSearchCV(clf, parameters,cv=3, scoring='roc_auc')
#X = np.array([[1,2], [3,4], [2,1], [4,3], [1,0], [4,5]])
#Y = np.array([0, 1, 0, 1, 0, 1])
clf.fit(X_tr, y_train)

# print(clf.grid_scores_)
best_parameters, score, _ = max(clf.grid_scores_, key=lambda x: x[1])
print('score:', score)
for param_name in sorted(best_parameters.keys()):
    print("%s: %r" % (param_name, best_parameters[param_name]))

```

score: 0.6148373470648865
 learning_rate: 0.2
 subsample: 0.9
 Wall time: 2min 14s

HeatMap using Seaborn

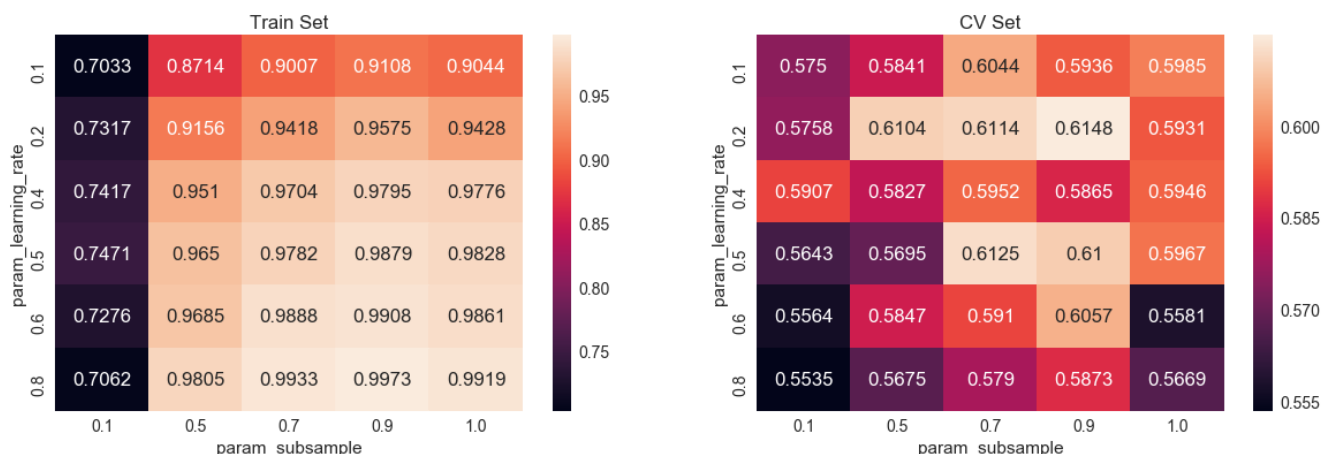
In [186]:

```

import seaborn as sns
import pandas as pd
#reference https://seaborn.pydata.org/generated/seaborn.heatmap.html ,
https://blog.exploratory.io/quick-introduction-to-heatmap-c21a9f9e4644?gi=cbae67554962

max_scores1 = pd.DataFrame(clf.cv_results_).groupby(['param_learning_rate',
'param_subsample']).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()

```



XGBoost with Optimal Parameters

In [182]:

```

from sklearn.model_selection import GridSearchCV
clf = XGBoostClassifier(eval_metric = 'auc', num_class = 2, nthread = 4)
best_parameters= {'learning_rate':[0.2], 'subsample' :[0.9]}
clf = GridSearchCV(clf, best_parameters,cv=3, scoring='roc_auc')
clf.fit(X_tr, y_train)

```

```

print(clf.predict_proba(X_tr)[:,-1])
y_train_pred= clf.predict_proba(X_tr)[:,-1]
y_test_pred= clf.predict_proba(X_te)[:,-1]
#print(X_tr , X_te)
train_fpr1, train_tpr1, tr_thresholds1 = roc_curve(y_train, y_train_pred)
test_fpr1, test_tpr1, te_thresholds1 = roc_curve(y_test, y_test_pred)
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()

```

[0.7337475 0.8135566 0.3237442 ... 0.794764 0.88391733 0.89353365]



Confusion Matrix for test and train data

In [173]:

```

def get_confusion_matrix(y_test,y_pred):
    df_cm = pd.DataFrame(confusion_matrix(y_test, y_pred), range(2),range(2))
    df_cm.columns = ['Predicted NO','Predicted YES']
    df_cm = df_cm.rename({0: 'Actual NO', 1: 'Actual YES'})
    sns.set(font_scale=1.4)#for label size
    sns.heatmap(df_cm, annot=True,annot_kws={"size": 16}, fmt='g')

```

In [176]:

```

predbow = (clf.predict(X_te))
predbow_train=(clf.predict(X_tr))

```

In [177]:

```

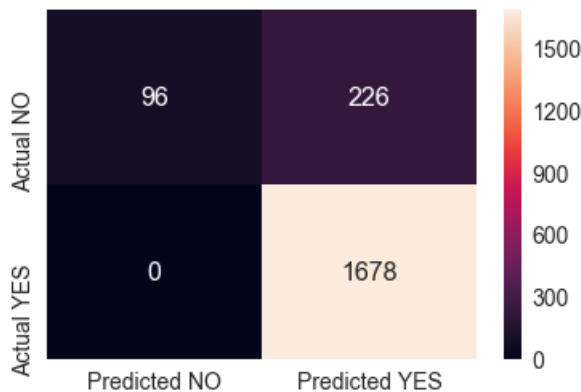
get_confusion_matrix(y_test,predbow)

```



In [178]:

```
get_confusion_matrix(y_train,predbow_train)
```



In [179]:

```
from sklearn.metrics import classification_report
print(classification_report(y_test ,predbow))
```

	precision	recall	f1-score	support
0	0.16	0.03	0.04	307
1	0.85	0.97	0.91	1693
avg / total	0.74	0.83	0.77	2000

3. Conclusion

Observations:

1. For creating top features depending on IDF value , instead of using existing techniques likes TFIDF,AVG W2V , function which calculates IDF vaues of each word in al documnets was created
2. Creation of covariance matrix is done using pandas dataframe , as it would be easier to assign row and column names as top 2000 features.
- 3.Inorder to find index of each word index() was first used and observed that it returned i ndex of only first occurence of word.Then another function was used to calcuate a indices of word.Inorder to find word in forward and backward direction abs() is used so that no additiona checks are required at boundaries.
- 4.Applying truncated SVD was easier and took less time and found that input for this shoud be sparse matrix.
- 5.Combining features in both train and test also was easy but there were some issues in bri nging all data to same shape while applying hstack()
- 6.Atlast applying XGboost function took very less time and it was very fast compared to oth er algorithms.Two parameters were considered 'max_depth' and 'eta'.