



Image Source: [https://www.google.com/search?](https://www.google.com/search?biw=1366&bih=637&tbm=isch&sxsr=ACYBGNTtmKSa8sZ9msdRhqKngmdQh_XDkg%3A1577293927022&sa=1,wiz-img.....35i39j0i24j0i10i24.TlwH9NXuj1s#imgsrc=ffQstrgGmatx7M_(https://www.google.com/search?biw=1366&bih=637&tbm=isch&sxsr=ACYBGNTtmKSa8sZ9msdRhqKngmdQh_XDkg%3A1577293927022&sa=1,wiz-img.....35i39j0i24j0i10i24.TlwH9NXuj1s#imgsrc=ffQstrgGmatx7M):)

[biw=1366&bih=637&tbm=isch&sxsr=ACYBGNTtmKSa8sZ9msdRhqKngmdQh_XDkg%3A1577293927022&sa=1,wiz-img.....35i39j0i24j0i10i24.TlwH9NXuj1s#imgsrc=ffQstrgGmatx7M_\(https://www.google.com/search?biw=1366&bih=637&tbm=isch&sxsr=ACYBGNTtmKSa8sZ9msdRhqKngmdQh_XDkg%3A1577293927022&sa=1,wiz-img.....35i39j0i24j0i10i24.TlwH9NXuj1s#imgsrc=ffQstrgGmatx7M\):](https://www.google.com/search?biw=1366&bih=637&tbm=isch&sxsr=ACYBGNTtmKSa8sZ9msdRhqKngmdQh_XDkg%3A1577293927022&sa=1,wiz-img.....35i39j0i24j0i10i24.TlwH9NXuj1s#imgsrc=ffQstrgGmatx7M_(https://www.google.com/search?biw=1366&bih=637&tbm=isch&sxsr=ACYBGNTtmKSa8sZ9msdRhqKngmdQh_XDkg%3A1577293927022&sa=1,wiz-img.....35i39j0i24j0i10i24.TlwH9NXuj1s#imgsrc=ffQstrgGmatx7M):)

Grasping the data and datasource - 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	
<code>project_id</code>	A unique identifier for the proposed project. Example: 123456789
<code>project_title</code>	Title of the project • Art Will Make You • First
<code>project_grade_category</code>	Grade level of students for which the project is targeted. One of the following enumerated categories: • Kindergarten • Grade 1 • Grade 2 • Grade 3 • Grade 4 • Grade 5 • Grade 6 • Grade 7 • Grade 8 • Grade 9 • Grade 10 • Grade 11 • Grade 12
<code>project_subject_categories</code>	One or more (comma-separated) subject categories for the project. The following enumerated list of categories is provided: • Applied • Care • Health • History • Literacy & • Math • Music & • Spec
<code>school_state</code>	State where school is located (Two-letter U.S. state abbreviations) (https://en.wikipedia.org/wiki/List of U.S. state abbreviations#Positional abbreviations) Example: CA
<code>project_subject_subcategories</code>	One or more (comma-separated) subject subcategories for the project. • • Literature & Writing, Social
<code>project_resource_summary</code>	An explanation of the resources needed for the project. • My students need hands on literacy materials and sensory nee
<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: 2016-01-12T12:00:00
<code>teacher_id</code>	A unique identifier for the teacher of the proposed project. bdf8baa8fedef6bfeec7ae4

Feature

Teacher's title. One of the following enumerations:

`teacher_prefix`

-
-
-
-
-
-

`teacher_number_of_previously_posted_projects`

Number of project applications previously submitted by the teacher

* 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 <code>0</code> indicates the project was not approved, and a value of <code>1</code> 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_4:** "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.

- Understanding the Businessreal world problem
 - Loading the data
 - Preprocessing the data(based on the type of data = categorical , text, Numarical)
 - Preprocessing data includes (removing outliers, impute missung values, cleaning data,etc..)
 - Split the data into train, cv, test
 - Vectorization data (one hot encoding)
 - Vectorizing text data
 - Normalizing
 - Contactinating all the type of features(cat + text)
 - Caluculating the idf values,
 - sorting vocab based on idf value and selecting top 2000 words,
 - Applying TruncatedSVD on co-occurence matrix for dimation reduction
 - Ploting n_components v/s Explained Variance
 - Sentiment scores for each essay
 - Merge matrices
 - Apply XGBoost on the Final Features
 - Hyperparameter tuning to find th best estimator(GridSearchCV)
 - Train the XGBoost model using best hyperparameter and plotting auc roc-curve
 - Plot Confusion Matrix
 - Observation on overall model performances
 - Ploting the performances by tableau format.
-

```
C:\Users\Ramesh Battu> import required libraries
```

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
```

1.1 Reading Data

In [9]:

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

In [10]:

```
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 [11]:

```
# 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)
project_data.sort_values(by=['Date'], inplace=True)

# how to reorder columns pandas python: https://stackoverflow.com/a/13148611/4084039
project_data = project_data[cols]
```

In [12]:

```
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[12]:

	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

1.1.1 preprocessing of project_subject_categories

In [13]:

```
# remove special characters from list of strings python: https://stackoverflow.com/a/47.
# 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
categories = list(project_data['project_subject_categories'].values)

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
        if 'The' in j.split(): # this will split each of the category based on space "Math
            j=j.replace('The','') # if we have the words "The" we are going to replace
        j = j.replace(' ','') # we are placing all the ' '(space) with ''(empty) ex:"Math
        temp+=j.strip()+" " #" abc ".strip() will return "abc", remove the trailing space
        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]))
```

1.1.2 preprocessing of project_subject_subcategories

In [14]:

```
sub_catogories = list(project_data['project_subject_subcategories'].values)
# remove special characters from list of strings python: https://stackoverflow.com/a/47
# 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
        if 'The' in j.split(): # this will split each of the catogory based on space "M
            j=j.replace('The','') # if we have the words "The" we are going to replace
        j = j.replace(' ', '') # we are placeing all the ' '(space) with ''(empty) ex:"M
        temp +=j.strip()+" #" "abc ".strip() will return "abc", remove the trailing spa
        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]))
```

1.1.3 preprocessing of school_state

In [15]:

```
# remove special characters from list of strings python: https://stackoverflow.com/a/47.
# 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
school_state_categories = list(project_data['school_state'].values)
cat_list = []
for i in school_state_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
        if 'The' in j.split(): # this will split each of the category based on space "Math
            j=j.replace('The','') # if we have the words "The" we are going to replace
            j = j.replace(' ','') # we are replacing all the ' ' (space) with '' (empty) ex: "Math
            temp+=j.strip()+" " #" abc ".strip() will return "abc", remove the trailing space
            temp = temp.replace('&','_') # we are replacing the & value into
    cat_list.append(temp.strip())
project_data['school_state'] = cat_list

from collections import Counter
my_counter = Counter()
for word in project_data['school_state'].values:
    my_counter.update(word.split())

cat_dict = dict(my_counter)
sorted_school_state_dict = dict(sorted(cat_dict.items(), key=lambda kv: kv[1]))
```

1.1.4 Preprocessing of teacher_prefix

In [152]:

```
# citation code :https://www.datacamp.com/community/tutorials/categorical-data
project_data = project_data.fillna(project_data['teacher_prefix'].value_counts().index[0])
teacher_prefix_categories = list(project_data['teacher_prefix'].values)
# Citation code : https://stackoverflow.com/questions/39303912/tfidfvectorizer-in-scikit-learn
# To convert the data type object to unicode string : used ""astype('U')"" code from
# count of all the words in corpus python: https://stackoverflow.com/a/22898595/4084039
# remove special characters from list of strings python: https://stackoverflow.com/a/4714111/4084039
# 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 teacher_prefix_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"
            j=j.replace('The','') # if we have the words "The" we are going to replace it with ''
        j = j.replace(' ', '') # we are placing all the ' ' (space) with '' (empty) ex: "Math & Science"
        temp+=j.strip()+" " # " abc ".strip() will return "abc", remove the trailing space
        temp = temp.replace('&','_') # we are replacing the & value into _
        temp =temp.replace('.', '') # we are removing dot(.)
    cat_list.append(temp.strip())

project_data['teacher_prefix'] = cat_list

from collections import Counter
my_counter = Counter()
for word in project_data['teacher_prefix'].values:
    word = str(word)
    my_counter.update(word.split())

# dict sort by value python: https://stackoverflow.com/a/613218/4084039
teacher_prefix_dict = dict(my_counter)
sorted_teacher_prefix_dict = dict(sorted(teacher_prefix_dict.items(), key=lambda kv: kv[1], reverse=True))
```

In [153]:

```
from sklearn.feature_extraction.text import CountVectorizer
vectorizer_tp = CountVectorizer(vocabulary=list(sorted_teacher_prefix_dict.keys()), lowercase=False)

teacher_prefix_one_hot = vectorizer_tp.fit_transform(project_data['teacher_prefix'].values)
print("vectorizer of project_subject_subcategories feature names :", vectorizer_tp.get_feature_names())

vectorizer of project_subject_subcategories feature names : ['Dr', 'Teacher', 'Mr', 'Ms', 'Mrs']
```

1.1.5 Preprocessing of project_grade_category

In [154]:

```
project_grade_categories = list(project_data['project_grade_category'].values)
# remove special characters from list of strings python: https://stackoverflow.com/a/47
# 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 project_grade_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
        if 'The' in j.split(): # this will split each of the category based on space "Math
            j=j.replace('The','') # if we have the words "The" we are going to replace
        j = j.replace(' ','') # we are replacing all the ' '(space) with ''(empty) ex:"Math
        temp+=j.strip()+" " #" abc ".strip() will return "abc", remove the trailing space
        temp = temp.replace('&','_') # we are replacing the & value into
        temp = temp.replace('-', '_') # we are replacing '-' with '_'
    cat_list.append(temp.strip())

project_data['project_grade_category'] = cat_list

#Link : https://www.datacamp.com/community/tutorials/categorical-data
project_data = project_data.fillna(project_data['project_grade_category'].value_counts(

# count of all the words in corpus python: https://stackoverflow.com/a/22898595/4084039
from collections import Counter
my_counter = Counter()
for word in project_data['project_grade_category'].values:
    word = str(word)
    my_counter.update(word.split())

# dict sort by value python: https://stackoverflow.com/a/613218/4084039
project_grade_category_dict = dict(my_counter)
sorted_project_grade_category_dict = dict(sorted(project_grade_category_dict.items(), key=
```

1.2. Text Preprocessing

1.2.1 Text Preprocessing of essay

In [18]:

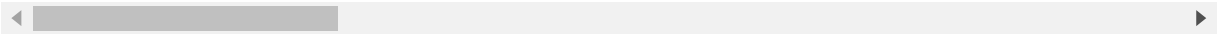
```
# merge two column text dataframe:
project_data["essay"] = project_data["project_essay_1"].map(str) + \
    project_data["project_essay_2"].map(str) + \
    project_data["project_essay_3"].map(str) + \
    project_data["project_essay_4"].map(str)
```

In [0]:

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

Out[8]:

	Unnamed: 0	id	teacher_id	teacher_prefix	school_state	project
0	160221	p253737	c90749f5d961ff158d4b4d1e7dc665fc	Mrs.	IN	
1	140945	p258326	897464ce9ddc600bced1151f324dd63a	Mr.	FL	



In [19]:

```
# printing some random reviews
print(project_data['essay'].values[20])
print("="*50)
print(project_data['essay'].values[120])
print("="*50)
print(project_data['essay'].values[1020])
print("="*50)
print(project_data['essay'].values[20020])
print("="*50)
print(project_data['essay'].values[99920])
print("="*50)
```

Throughout this school year, I hope to enable my students to develop a love for reading, discovering, and creating. Students in my classroom have the opportunity to explore learning through large group activities, small group learning times, and interactive play with other students and teachers. My school is a Pre-K through 8 school with teachers and staff who are determined to help children in Detroit succeed. Students in my classroom are just beginning their education in Pre-K and I strive to enable my students to shine brightly as they learn using the High Scope Curriculum. The shopping carts would be a wonderful addition to our dramatic play area. With them, the children will be able to work on their social skills. They will learn about sharing, purchasing items, and the cost of different items. Similarly, the parachute will be used to encourage cooperative play. The children will have fun learning and exercising. I'm looking forward to adding the beads to our art area in the classroom to encourage both creativity as well as fine motor skills as the children work on stringing them. Donations to this project will allow me to encourage my students' curiosity and develop a love for learning in them. \n\nWith these resources, my students will start off on the right path in their education

=====

Every afternoon we extend our learning through creative play. Creative play prepares kids to be next generation innovators! My students' location does not define their ability or determine their future. Our city has been voted the "Most Dangerous" and "Poorest" city in the United States consistently in the past 30 years. Fortunately these labels are not actual descriptions of who we are. My students are innocent, hardworking, energetic, inquisitive and zealous for knowledge. They are young leaders fighting for a 21st century education. \n\nIn my school, 100% of our eager students receive free breakfast and lunch. 100% of our dedicated students are thirsty to learn. 100% of our teachers are driving our students to grow beyond a year's growth. We are in the fight and we will win. We are in need of resources that will spark our creativity and open our minds to possibilities. The light table will supply a different way to view material. While the kinetic sand and water bead will allow them to use their senses to create new artifacts that extend their thinking and imagine boundless ideas. These tools will encourage discussion and team building techniques. These resources will fuel our learning as they open their minds to what's possible. They will encourage my students to take change, problem solve, and collaborate. They will become creative thinkers and doers. It will also help my students cope in their everyday lives.

=====

Over 25% of Kindergarten have significant poor attention and focus. Sensory tools can help kids focus, so they can learn. Movement and physical activity are imperative to brain development, academic performance and self-regulation. These tools increase teaching time and decrease redirection time. Our students are from a school district where about 50% of the

kids qualify for the free lunch program. Many of the kids that need sensory tools are boys. Our kids need to move to pay attention. Providing the kids with sensory tools can give them a well deserved advantage in the classroom. Many of our kids are in under served population groups, whose families struggle to put food on the table and have few resources to support their child's education. The kids need our help to increase focus and attention in the classroom. The teachers need easily accessible tools they can use during instruction to provide needed movement to keep kids focused and learning. The sensory toolbox items will be available to the students as needed. When the teacher sees that "Susie" is off task, she will walk around to her and offer her a fidget toy from the fidget basket. Susie chooses the squishy stress ball and starts squeezing it. Soon, Susie is back on task and answering the next question correctly. When the teacher sees Joey tapping his foot, slouching in his chair or standing frequently at his desk, she carries a move and sit cushion to Joey to sit on. Joey proceeds to wiggle on the seat cushion and is able to stay in his seat, yet receive the sought after movement. Both these examples demonstrate how attention and focus are improved without the use of disciplinary actions or yelling. Using this proactive approach increases confidence, intelligence, and overall learning. Donations towards my project to fund sensory tools for the classrooms will empower teachers to make a difference in how well students can attend and focus, while at the same time reducing the need for disciplinary actions related to poor attention and focus in the classroom. It's a win-win situation. Teachers can teach, and children can learn by keeping them moving and fidgeting.

=====

I am a second year teacher and I have the pleasure of teaching a kindergarten class that is filled with 21 sweet South Carolina students. Part of my job as a kindergarten teacher is to make sure my students have a safe and fun environment in which to learn so they will have a good outlook about school. I have the privilege to help change the lives of children and I want nothing but the best for my class. My students are between 5 and 6 years old and they are learning some of the hardest things they will ever learn: reading, writing, and sharing. They love being at school and having a fun environment to learn in. My students come from all walks of life and I want each student to have the best chance to succeed. I am a kindergarten teacher at a public charter school that will be open for its third year this month. My students last year inspired me to create this project, since everyday I saw some sit and stand instead of playing and having fun. My students love to play football and last year they played it using their imagination, since we did not have a football. There is a big need for playground equipment that allows them to move and play. We have a basketball goal but the balls are needing replaced. The bowling ball set and hula hoops will be great for those that don't want to play a contact game. The mobile cart will be a lifesaver with moving the equipment to and from the playground each day. \r\nMost of all my students will not stand around during recess and not play when they get these wonderful items! Mrs. Mrs.

=====

I have a large population of Haitian students in our beautiful classroom in P.H. Elementary in Florida. Many enter my class struggling with the English language and we work very hard to keep up with the Common Core Standards. I work in a Title I school and my students live in high poverty. It's a very old school building and needless to say supplies of any kind are short in demand. Luckily love and determination are in abundance in my young hard workers. We don't have such beautiful supplies that can quickly engage the mind of the learner and show them that there are so many ways to learn through play. We have Center-Time in our classrooms which must show them to learn independently and these types of supplies would push them to create exemplary center work. These games will allow th

em to grow academically while actually playing a game. What is now undeniably clear in the 21st century is that play is essential, vital, critical, and fundamental to a child's social, emotional, physical, and intellectual development. These types of supplies hold these words to be true in a learning classroom.Mrs.Mrs.

=====

In [20]:

```
# 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 [22]:

```
sent = decontracted(project_data['essay'].values[20020])
```

```
print(sent)
```

```
print("="*80)
```

I am a second year teacher and I have the pleasure of teaching a kindergarten class that is filled with 21 sweet South Carolina students. Part of my job as a kindergarten teacher is to make sure my students have a safe and fun environment in which to learn so they will have a good outlook about school. I have the privilege to help change the lives of children and I want nothing but the best for my class. My students are between 5 and 6 years old and they are learning some of the hardest things they will ever learn: reading, writing, and sharing. They love being at school and having a fun environment to learn in. My students come from all walks of life and I want each student to have the best chance to succeed.I am a kindergarten teacher at a public charter school that will be open for its third year this month. My students last year inspired me to create this project, since everyday I saw some sit and stand instead of playing and having fun. My students love to play football and last year they played it using their imagination, since we did not have a football. There is a big need for playground equipment that allows them to move and play. We have a basketball goal but the balls are needing replaced. The bowling ball set and hula hoops will be great for those that do not want to play a contact game. The mobile cart will be a lifesaver with moving the equipment to and from the playground each day. \r\nMost of all my students will not stand around during recess and not play when they get these wonderful items!Mrs.Mrs.

=====

=====

In [23]:

```
# \r \n \t remove from string python: http://texthandler.com/info/remove-line-breaks-py
sent = sent.replace('\r', ' ')
sent = sent.replace('\n', ' ')
sent = sent.replace('\t', ' ')
print(sent)
```

I am a second year teacher and I have the pleasure of teaching a kindergarten class that is filled with 21 sweet South Carolina students. Part of my job as a kindergarten teacher is to make sure my students have a safe and fun environment in which to learn so they will have a good outlook about school. I have the privilege to help change the lives of children and I want nothing but the best for my class. My students are between 5 and 6 years old and they are learning some of the hardest things they will ever learn: reading, writing, and sharing. They love being at school and having a fun environment to learn in. My students come from all walks of life and I want each student to have the best chance to succeed. I am a kindergarten teacher at a public charter school that will be open for its third year this month. My students last year inspired me to create this project, since everyday I saw some sit and stand instead of playing and having fun. My students love to play football and last year they played it using their imagination, since we did not have a football. There is a big need for playground equipment that allows them to move and play. We have a basketball goal but the balls are needing replaced. The bowling ball set and hula hoops will be great for those that do not want to play a contact game. The mobile cart will be a lifesaver with moving the equipment to and from the playground each day. Most of all my students will not stand around during recess and not play when they get these wonderful items! Mrs. Mrs.

In [24]:

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

I am a second year teacher and I have the pleasure of teaching a kindergarten class that is filled with 21 sweet South Carolina students Part of my job as a kindergarten teacher is to make sure my students have a safe and fun environment in which to learn so they will have a good outlook about school I have the privilege to help change the lives of children and I want nothing but the best for my class My students are between 5 and 6 years old and they are learning some of the hardest things they will ever learn reading writing and sharing They love being at school and having a fun environment to learn in My students come from all walks of life and I want each student to have the best chance to succeed I am a kindergarten teacher at a public charter school that will be open for its third year this month My students last year inspired me to create this project since everyday I saw some sit and stand instead of playing and having fun My students love to play football and last year they played it using their imagination since we did not have a football There is a big need for playground equipment that allows them to move and play We have a basketball goal but the balls are needing replaced The bowling ball set and hula hoops will be great for those that do not want to play a contact game The mobile cart will be a lifesaver with moving the equipment to and from the playground each day Most of all my students will not stand around during recess and not play when they get these wonderful items Mrs Mrs

In [25]:

```
# https://gist.github.com/sebleier/554280
# we are removing the words from the stop words list: 'no', 'nor', 'not'
stopwords= ["a", "about", "above", "after", "again", "against", "ain", "all", "am", "an", "and", "a",
"as", "at", "be", "because", "been", "before", "being", "below", "between", "both",
"d", "did", "didn", "didn't", "do", "does", "doesn", "doesn't", "doing", "don", "don",
"for", "from", "further", "had", "hadn", "hadn't", "has", "hasn", "hasn't", "have",
"here", "hers", "herself", "him", "himself", "his", "how", "i", "if", "in", "into", "i",
"itself", "just", "ll", "m", "ma", "me", "mightn", "mightn't", "more", "most", "must",
"needn't", "no", "nor", "not", "now", "o", "of", "off", "on", "once", "only", "or", "o",
"out", "over", "own", "re", "s", "same", "shan", "shan't", "she", "she's", "should",
"so", "some", "such", "t", "than", "that", "that'll", "the", "their", "theirs", "the",
"these", "they", "this", "those", "through", "to", "too", "under", "until", "up", "v",
"we", "were", "weren", "weren't", "what", "when", "where", "which", "while", "who",
"won't", "wouldn", "wouldn't", "y", "you", "you'd", "you'll", "you're", "you've", "y",
"yourselves", "could", "he'd", "he'll", "he's", "here's", "how's", "i'd", "i'll", "i",
"she'd", "she'll", "that's", "there's", "they'd", "they'll", "they're", "they've",
"what's", "when's", "where's", "who's", "why's", "would", "able", "abst", "accord",
"across", "act", "actually", "added", "adj", "affected", "affecting", "affects", "a",
"along", "already", "also", "although", "always", "among", "amongst", "announce",
"anymore", "anyone", "anything", "anyway", "anyways", "anywhere", "apparently", "a",
"around", "aside", "ask", "asking", "auth", "available", "away", "awfully", "b", "b",
"becoming", "beforehand", "begin", "beginning", "beginnings", "begins", "behind",
"beyond", "biol", "brief", "briefly", "c", "ca", "came", "cannot", "can't", "cause",
"co", "com", "come", "comes", "contain", "containing", "contains", "couldnt", "date",
"due", "e", "ed", "edu", "effect", "eg", "eight", "eighty", "either", "else", "elsew",
"especially", "et", "etc", "even", "ever", "every", "everybody", "everyone", "every",
"f", "far", "ff", "fifth", "first", "five", "fix", "followed", "following", "follow",
"found", "four", "furthermore", "g", "gave", "get", "gets", "getting", "give", "give",
"gone", "got", "gotten", "h", "happens", "hardly", "hed", "hence", "hereafter", "he",
"hes", "hi", "hid", "hither", "home", "howbeit", "however", "hundred", "id", "ie", "i",
"importance", "important", "inc", "indeed", "index", "information", "instead", "in",
"it'll", "j", "k", "keep", "keeps", "kept", "kg", "km", "know", "known", "knows", "l",
"later", "latter", "latterly", "least", "less", "lest", "let", "lets", "like", "like",
"ll", "look", "looking", "looks", "ltd", "made", "mainly", "make", "makes", "many",
"meantime", "meanwhile", "merely", "mg", "might", "million", "miss", "ml", "moreov",
"mug", "must", "n", "na", "name", "namely", "nay", "nd", "near", "nearly", "necessar",
"neither", "never", "nevertheless", "new", "next", "nine", "ninety", "nobody", "no",
"normally", "nos", "noted", "nothing", "nowhere", "obtain", "obtained", "obviously",
"omitted", "one", "ones", "onto", "ord", "others", "otherwise", "outside", "overall",
"particular", "particularly", "past", "per", "perhaps", "placed", "please", "plus",
"potentially", "pp", "predominantly", "present", "previously", "primarily", "prol",
"provides", "put", "q", "que", "quickly", "quite", "qv", "r", "ran", "rather", "rd",
"recently", "ref", "refs", "regarding", "regardless", "regards", "related", "rela",
"resulted", "resulting", "results", "right", "run", "said", "saw", "say", "saying",
"seeing", "seem", "seemed", "seeming", "seems", "seen", "self", "selves", "sent", "s",
"shes", "show", "showed", "shown", "shows", "significant", "significant",
"six", "slightly", "somebody", "somehow", "someone", "somethan", "something", "so",
"somewhere", "soon", "sorry", "specifically", "specified", "specify", "specifying",
"sub", "substantially", "successfully", "sufficiently", "suggest", "sup", "sure",
"tends", "th", "thank", "thanks", "thanx", "thats", "that've", "thence", "thereaft",
"therein", "there'll", "thereof", "therere", "theres", "thereto", "thereupon", "tl",
"thou", "though", "thoughh", "thousand", "throug", "throughout", "thru", "thus", "t",
"toward", "towards", "tried", "tries", "truly", "try", "trying", "ts", "twice", "tw",
"unless", "unlike", "unlikely", "unto", "upon", "ups", "us", "use", "used", "useful",
"using", "usually", "v", "value", "various", "'ve", "via", "viz", "vol", "vols", "vs",
"wed", "welcome", "went", "werent", "whatever", "what'll", "whats", "whence", "when",
"whereby", "wherein", "wheres", "whereupon", "wherever", "whether", "whim", "whit",
"who'll", "whomever", "whos", "whose", "widely", "willing", "wish", "within", "wit",
"wouldnt", "www", "x", "yes", "yet", "you'd", "you're", "z", "zero", "a's", "ain't", "a
```

```
"appreciate", "appropriate", "associated", "best", "better", "c'mon", "c's", "can",  
"consequently", "consider", "considering", "corresponding", "course", "currently",  
"entirely", "exactly", "example", "going", "greetings", "hello", "help", "hopeful",  
"indicated", "indicates", "inner", "insofar", "it'd", "keep", "keeps", "novel", "p",  
"secondly", "sensible", "serious", "seriously", "sure", "t's", "third", "thorough",  
"wonder"]
```

In [26]:

```
%%time  
# Combining all the above students  
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('\\\"', ' ')  
    sent = sent.replace('\\n', ' ')  
    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.append(sent.lower().strip())
```

Wall time: 0 ns

```
100%|████████████████████████████████████████| 109248/109248 [05:55<00:00, 307.31  
it/s]
```

In [27]:

```
# after preprocessing  
preprocessed_essays[20020]
```

Out[27]:

```
'year teacher pleasure teaching kindergarten class filled 21 sweet south c  
arolina students job kindergarten teacher students safe fun environment le  
arn good outlook school privilege change lives children class students 5 6  
years learning hardest things learn reading writing sharing love school fu  
n environment learn students walks life student chance succeed kindergarte  
n teacher public charter school open year month students year inspired cre  
ate project everyday sit stand playing fun students love play football yea  
r played imagination football big playground equipment move play basketbal  
l goal balls needing replaced bowling ball set hula hoops great play conta  
ct game mobile cart lifesaver moving equipment playground day students sta  
nd recess play wonderful items'
```

1.2.2 Text Preprocessing of project_title

In [28]:

```
print(project_data['project_title'].tail(1))
```

```
78306    News for Kids  
Name: project_title, dtype: object
```

In [29]:

```
# printing some random title texts
print(project_data['project_title'].values[20])
print('--'*19)
print(project_data['project_title'].values[120])
print('--'*19)
print(project_data['project_title'].values[1920])
print('--'*19)
print(project_data['project_title'].values[99920])
print('--'*19)
```

```
Pre-K Classroom Materials
-----
Empowering Kindergarteners
-----
A Classroom Fit for all Learners
-----
Turn That TV Game Off and Turn on the Learning!
-----
```

In [30]:

```
# 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 [31]:

```
sent = decontracted(project_data['project_title'].values[99920])
print(sent)
print("="*120)
```

```
Turn That TV Game Off and Turn on the Learning!
=====
=====
```

In [32]:

```
# \r \n \t remove from string python: http://texthandler.com/info/remove-line-breaks-py
sent = sent.replace('\r', ' ')
sent = sent.replace('\n', ' ')
sent = sent.replace('\t', ' ')
sent = sent.replace('!', ' ')
print(sent)
```

Turn That TV Game Off and Turn on the Learning

In [33]:

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

Turn That TV Game Off and Turn on the Learning

In [34]:

```
# https://gist.github.com/sebleier/554280
# we are removing the words from the stop words list: 'no', 'nor', 'not'
stopwords= ["a", "about", "above", "after", "again", "against", "ain", "all", "am", "an", "and", "a",
"as", "at", "be", "because", "been", "before", "being", "below", "between", "both",
"d", "did", "didn", "didn't", "do", "does", "doesn", "doesn't", "doing", "don", "don
"for", "from", "further", "had", "hadn", "hadn't", "has", "hasn", "hasn't", "have",
"here", "hers", "herself", "him", "himself", "his", "how", "i", "if", "in", "into", "
"itself", "just", "ll", "m", "ma", "me", "mightn", "mightn't", "more", "most", "must
"needn't", "no", "nor", "not", "now", "o", "of", "off", "on", "once", "only", "or", "o
"out", "over", "own", "re", "s", "same", "shan", "shan't", "she", "she's", "should",
"so", "some", "such", "t", "than", "that", "that'll", "the", "their", "theirs", "the
"these", "they", "this", "those", "through", "to", "too", "under", "until", "up", "v
"we", "were", "weren", "weren't", "what", "when", "where", "which", "while", "who",
"won't", "wouldn", "wouldn't", "y", "you", "you'd", "you'll", "you're", "you've", "
"yourselves", "could", "he'd", "he'll", "he's", "here's", "how's", "i'd", "i'll", "
"she'd", "she'll", "that's", "there's", "they'd", "they'll", "they're", "they've"
"what's", "when's", "where's", "who's", "why's", "would", "able", "abst", "accord
"across", "act", "actually", "added", "adj", "affected", "affecting", "affects", "
"along", "already", "also", "although", "always", "among", "amongst", "announce",
"anymore", "anyone", "anything", "anyway", "anyways", "anywhere", "apparently", "
"around", "aside", "ask", "asking", "auth", "available", "away", "awfully", "b", "b
"becoming", "beforehand", "begin", "beginning", "beginnings", "begins", "behind"
"beyond", "biol", "brief", "briefly", "c", "ca", "came", "cannot", "can't", "cause"
"co", "com", "come", "comes", "contain", "containing", "contains", "couldnt", "dat
"due", "e", "ed", "edu", "effect", "eg", "eight", "eighty", "either", "else", "elsew
"especially", "et", "etc", "even", "ever", "every", "everybody", "everyone", "every
"f", "far", "ff", "fifth", "first", "five", "fix", "followed", "following", "follow
"found", "four", "furthermore", "g", "gave", "get", "gets", "getting", "give", "give
"gone", "got", "gotten", "h", "happens", "hardly", "hed", "hence", "hereafter", "he
"hes", "hi", "hid", "hither", "home", "howbeit", "however", "hundred", "id", "ie", "
"importance", "important", "inc", "indeed", "index", "information", "instead", "i
"it'll", "j", "k", "keep", "keeps", "kept", "kg", "km", "know", "known", "knows", "l"
"later", "latter", "latterly", "least", "less", "lest", "let", "lets", "like", "like
"ll", "look", "looking", "looks", "ltd", "made", "mainly", "make", "makes", "many"
"meantime", "meanwhile", "merely", "mg", "might", "million", "miss", "ml", "moreov
"mug", "must", "n", "na", "name", "namely", "nay", "nd", "near", "nearly", "necessar
"neither", "never", "nevertheless", "new", "next", "nine", "ninety", "nobody", "no
"normally", "nos", "noted", "nothing", "nowhere", "obtain", "obtained", "obviously
"omitted", "one", "ones", "onto", "ord", "others", "otherwise", "outside", "overal
"particular", "particularly", "past", "per", "perhaps", "placed", "please", "plus
"potentially", "pp", "predominantly", "present", "previously", "primarily", "prol
"provides", "put", "q", "que", "quickly", "quite", "qv", "r", "ran", "rather", "rd",
"recently", "ref", "refs", "regarding", "regardless", "regards", "related", "rela
"resulted", "resulting", "results", "right", "run", "said", "saw", "say", "saying"
"seeing", "seem", "seemed", "seeming", "seems", "seen", "self", "selves", "sent", "
"shes", "show", "showed", "shown", "shows", "significant", "significant
"six", "slightly", "somebody", "somehow", "someone", "somethan", "something", "so
"somewhere", "soon", "sorry", "specifically", "specified", "specify", "specifying
"sub", "substantially", "successfully", "sufficiently", "suggest", "sup", "sure"
"tends", "th", "thank", "thanks", "thanx", "thats", "that've", "thence", "thereaft
"therein", "there'll", "thereof", "therere", "theres", "thereto", "thereupon", "tl
"thou", "though", "thoughh", "thousand", "throug", "throughout", "thru", "thus", "
"toward", "towards", "tried", "tries", "truly", "try", "trying", "ts", "twice", "tw
"unless", "unlike", "unlikely", "unto", "upon", "ups", "us", "use", "used", "useful
"using", "usually", "v", "value", "various", "'ve", "via", "viz", "vol", "vols", "vs
"wed", "welcome", "went", "werent", "whatever", "what'll", "whats", "whence", "whe
"whereby", "wherein", "wheres", "whereupon", "wherever", "whether", "whim", "whitl
"who'll", "whomever", "whos", "whose", "widely", "willing", "wish", "within", "witl
"wouldnt", "www", "x", "yes", "yet", "you'd", "you're", "z", "zero", "a's", "ain't", "a
```

```
"appreciate", "appropriate", "associated", "best", "better", "c'mon", "c's", "can",  
"consequently", "consider", "considering", "corresponding", "course", "currentl",  
"entirely", "exactly", "example", "going", "greetings", "hello", "help", "hopeful",  
"indicated", "indicates", "inner", "insofar", "it'd", "keep", "keeps", "novel", "p",  
"secondly", "sensible", "serious", "seriously", "sure", "t's", "third", "thorough",  
"wonder"]
```

In [35]:

```
%%time  
# Combining all the above stundents  
from tqdm import tqdm  
preprocessed_project_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('\\\"', ' ')  
    sent = sent.replace('\\n', ' ')  
    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_project_title.append(sent.lower().strip())
```

```
100%|████████████████████████████████████████| 109248/109248 [00:12<00:00, 8822.72  
it/s]
```

Wall time: 12.4 s

In [36]:

```
preprocessed_project_title[99999]
```

Out[36]:

```
'turning flexible seating sixth grade class journey freedom'
```

1.3. Numerical normalization

1.3.1 normalization_price

In [38]:

```
# merge dataframes  
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')  
project_data.shape
```

Out[38]:

```
(109248, 20)
```

In [39]:

```
# Later merged dataframes
project_data.head(1)
```

Out[39]:

Unnamed: 0	id	teacher_id	teacher_prefix	school_state	Date
0	8393 p205479 2bf07ba08945e5d8b2a3f269b2b3cfe5		Mrs.	CA	2016-04-20 00:27:30

In [40]:

```
# Link: https://scikit-learn.org/stable/modules/generated/sklearn.preprocessing.Normalizer.html
from sklearn.preprocessing import Normalizer
# Reshaping price data using array.reshape(1,-1)
price_normalize = Normalizer()
price_normalizer = price_normalize.fit_transform(project_data['price'].values.reshape(1,-1))
price_normalizer = price_normalizer.T
print(price_normalizer)
print("-----")
print("shape of price_normalizer:", price_normalizer.shape)
```

```
[[4.63560392e-03]
 [1.36200635e-03]
 [2.10346002e-03]
 ...
 [2.55100471e-03]
 [1.83960046e-03]
 [3.51642253e-05]]
```

shape of price_normalizer: (109248, 1)

1.3.2 teacher number of previously posted projects normalization

In [41]:

```
project_data['teacher_number_of_previously_posted_projects'].values
```

Out[41]:

```
array([53,  4, 10, ...,  0,  1,  2], dtype=int64)
```

In [42]:

```
# Link: https://scikit-learn.org/stable/modules/generated/sklearn.preprocessing.Normalizer
from sklearn.preprocessing import Normalizer
teacher_number_of_previously_posted_projects_normalize = Normalizer()
teacher_number_of_previously_posted_projects_normalizer = teacher_number_of_previously_
teacher_number_of_previously_posted_projects_normalizer = teacher_number_of_previously_
print(teacher_number_of_previously_posted_projects_normalizer)
print("="*25)
print("Shape of teacher_number_of_previously_posted_projects_normalizer :", teacher_numl
```

```
[[0.00535705]
 [0.00040431]
 [0.00101076]
 ...
 [0.         ]
 [0.00010108]
 [0.00020215]]
```

=====

```
Shape of teacher_number_of_previously_posted_projects_normalizer : (10924
8, 1)
```

1.3.3. quantity Normalization

In [44]:

```
# Link: https://scikit-learn.org/stable/modules/generated/sklearn.preprocessing.Normalizer
normalizer = Normalizer()
quantity_normalizer = normalizer.fit_transform(project_data['quantity'].values.reshape(
print(quantity_normalizer)
print("Shape of quantity nomalizer :", quantity_normalizer.shape)
```

```
[[3.87894657e-04]
 [7.75789314e-04]
 [9.69736642e-05]
 ...
 [4.84868321e-04]
 [3.87894657e-04]
 [2.42434160e-03]]
```

```
Shape of quantity nomalizer : (109248, 1)
```

In [45]:

```
project_data['preprocessed_essays'] = preprocessed_essays
project_data['preprocessed_project_title'] = preprocessed_project_title
```

1.3.4 spilt the data into train ,CV and test

In [46]:

```
label = project_data['project_is_approved']
project_data.drop(['project_is_approved'], axis=1, inplace=True)
# splitting the data into train , test CV
# Reference Link :https://scikit-learn.org/stable/modules/generated/sklearn.model_selection
from sklearn.model_selection import train_test_split

X_train, X_test, y_train, y_test = train_test_split(project_data, label, test_size=0.33)
X_train, X_cv, y_train, y_cv = train_test_split(X_train, y_train, test_size=0.33) #

print("Shape of X_train and y_train :", X_train.shape, y_train.shape)
print("Shape of X_test and y_test :", X_test.shape, y_test.shape)
print("Shape of X_cv and y_cv :", X_cv.shape, y_cv.shape)
```

```
Shape of X_train and y_train : (49041, 21) (49041,)
Shape of X_test and y_test : (36052, 21) (36052,)
Shape of X_cv and y_cv : (24155, 21) (24155,)
```

1.4. Vectorizing Categorical data

1.4.1 Vectorization of clean_categories

<https://www.appliedaia.com/course/applied-ai-course-online/lessons/handling-categorical-and-numerical-features/> (<https://www.appliedaia.com/course/applied-ai-course-online/lessons/handling-categorical-and-numerical-features/>)

In [47]:

```
# we use count vectorizer to convert the values into one
from sklearn.feature_extraction.text import CountVectorizer
vectorizer_cat = CountVectorizer(vocabulary=list(sorted_cat_dict.keys()), lowercase=False)

clean_categories_one_hot_train = vectorizer_cat.fit_transform(X_train['clean_categories'])
clean_categories_one_hot_test = vectorizer_cat.transform(X_test['clean_categories'].values)
clean_categories_one_hot_cv = vectorizer_cat.transform(X_cv['clean_categories'].values)

print("vectorizer of project_subject_categories feature names :", vectorizer_cat.get_feature_names())
print("-----")
print("Shape of matrix after one hot encoding train :", clean_categories_one_hot_train.shape)
print("Shape of matrix after one hot encoding test :", clean_categories_one_hot_test.shape)
print("Shape of matrix after one hot encoding cv :", clean_categories_one_hot_cv.shape)
```

```
vectorizer of project_subject_categories feature names : ['Warmth', 'Care_
Hunger', 'History_Civics', 'Music_Arts', 'AppliedLearning', 'SpecialNeed
s', 'Health_Sports', 'Math_Science', 'Literacy_Language']
-----
```

```
Shape of matrix after one hot encoding train : (49041, 9)
Shape of matrix after one hot encoding test : (36052, 9)
Shape of matrix after one hot encoding cv : (24155, 9)
```

1.4.2 Vectorization of clean_sub_categories

In [48]:

```
# we use count vectorizer to convert the values into one
from sklearn.feature_extraction.text import CountVectorizer
vectorizer_subcat = CountVectorizer(vocabulary=list(sorted_sub_cat_dict.keys()), lowercase=True)

clean_subcategories_one_hot_train = vectorizer_subcat.fit_transform(X_train['clean_subcategories'])
clean_subcategories_one_hot_test = vectorizer_subcat.transform(X_test['clean_subcategories'])
clean_subcategories_one_hot_cv = vectorizer_subcat.transform(X_cv['clean_subcategories'])

print("vectorizer of project_subject_subcategories feature names :", vectorizer_subcat.get_feature_names())
print("-----")
print("Shape of matrix after one hot encoding train : ", clean_subcategories_one_hot_train.shape)
print("Shape of matrix after one hot encoding test : ", clean_subcategories_one_hot_test.shape)
print("Shape of matrix after one hot encoding cv : ", clean_subcategories_one_hot_cv.shape)
```

```
vectorizer of project_subject_subcategories feature names : ['Economics',
'CommunityService', 'FinancialLiteracy', 'ParentInvolvement', 'Extracurricular',
'Civics_Government', 'ForeignLanguages', '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']
-----
```

```
Shape of matrix after one hot encoding train : (49041, 30)
Shape of matrix after one hot encoding test : (36052, 30)
Shape of matrix after one hot encoding cv : (24155, 30)
```

1.4.3 Vectorization of school_state

In [49]:

```
# we use count vectorizer to convert the values into one
vectorizer_ss = CountVectorizer(vocabulary=list(sorted_school_state_dict.keys()), lowercase=True)

school_state_one_hot_train = vectorizer_ss.fit_transform(X_train['school_state'].values)
school_state_one_hot_test = vectorizer_ss.transform(X_test['school_state'].values)
school_state_one_hot_cv = vectorizer_ss.transform(X_cv['school_state'].values)

print("vectorizer of school_state feature names :", vectorizer_ss.get_feature_names())
print("-----")
print("Shape of matrix after one hot encoding train : ", school_state_one_hot_train.shape)
print("Shape of matrix after one hot encoding test : ", school_state_one_hot_test.shape)
print("Shape of matrix after one hot encoding cv : ", school_state_one_hot_cv.shape)
```

```
vectorizer of school_state feature names : ['VT', 'WY', 'ND', 'MT', 'RI',
'SD', 'NE', 'DE', 'AK', 'NH', 'WV', 'ME', 'HI', 'DC', 'NM', 'KS', 'IA', 'ID',
'AR', 'CO', 'MN', 'OR', 'KY', 'MS', 'NV', 'MD', 'CT', 'TN', 'UT', 'AL',
'WI', 'VA', 'AZ', 'NJ', 'OK', 'WA', 'MA', 'LA', 'OH', 'MO', 'IN', 'PA',
'MI', 'SC', 'GA', 'IL', 'NC', 'FL', 'NY', 'TX', 'CA']
-----
```

```
Shape of matrix after one hot encoding train : (49041, 51)
Shape of matrix after one hot encoding test : (36052, 51)
Shape of matrix after one hot encoding cv : (24155, 51)
```

1.4.4 Vectorization of teacher_prefix

In [155]:

```
# we use count vectorizer to convert the values into one hot encode
# we use count vectorizer to convert the values into one hot encoded features
vectorizer_tp = CountVectorizer(vocabulary=list(sorted_teacher_prefix_dict.keys()), low

teacher_prefix_one_hot_train = vectorizer_tp.fit_transform(X_train['teacher_prefix'].va
teacher_prefix_one_hot_test  = vectorizer_tp.transform(X_test['teacher_prefix'].values.
teacher_prefix_one_hot_cv     = vectorizer_tp.transform(X_cv['teacher_prefix'].values.as

print("Vectorizer of teacher_prefix :",vectorizer_tp.get_feature_names())
print("-----")
print("Shape of matrix after one hot encoding train : ",teacher_prefix_one_hot_train.sha
print("Shape of matrix after one hot encoding test  : ",teacher_prefix_one_hot_test.sha
print("Shape of matrix after one hot encoding cv     : ",teacher_prefix_one_hot_cv.shape)
```

```
Vectorizer of teacher_prefix : ['Dr', 'Teacher', 'Mr', 'Ms', 'Mrs']
```

```
-----
Shape of matrix after one hot encoding train : (49041, 5)
Shape of matrix after one hot encoding test  : (36052, 5)
Shape of matrix after one hot encoding cv     : (24155, 5)
```

In [52]:

```
vectorizer_cat.get_feature_names
```

Out[52]:

```
<bound method CountVectorizer.get_feature_names of CountVectorizer(analyze
r='word', binary=True, decode_error='strict',
                        dtype=<class 'numpy.int64'>, encoding='utf-8', input='cont
ent',
                        lowercase=False, max_df=1.0, max_features=None, min_df=1,
                        ngram_range=(1, 1), preprocessor=None, stop_words=None,
                        strip_accents=None, token_pattern='(?u)\\b\\w\\w+\\b',
                        tokenizer=None,
                        vocabulary=['Warmth', 'Care_Hunger', 'History_Civics',
                                'Music_Arts', 'AppliedLearning', 'SpecialNeed
s',
                                'Health_Sports', 'Math_Science',
                                'Literacy_Language'])>
```

1.4.5 Vectorization of project_grade_categorie

In [156]:

```
# we use count vectorizer to convert the values into one hot encoded features
vectorizer_pgc = CountVectorizer(vocabulary=list(sorted_project_grade_category_dict.keys()))

project_grade_category_one_hot_train = vectorizer_pgc.fit_transform(X_train['project_grade_category'])
project_grade_category_one_hot_test = vectorizer_pgc.transform(X_test['project_grade_category'])
project_grade_category_one_hot_cv = vectorizer_pgc.transform(X_cv['project_grade_category'])

print("Vectorizer of project_grade_categories :",vectorizer_pgc.get_feature_names())
print("-----")
print("Shape of matrix after one hot encoding train : ",project_grade_category_one_hot_train.shape)
print("Shape of matrix after one hot encoding test : ",project_grade_category_one_hot_test.shape)
print("Shape of matrix after one hot encoding cv : ",project_grade_category_one_hot_cv.shape)
```

```
Vectorizer of project_grade_categories : ['Grades9_12', 'Grades6_8', 'Grades3_5', 'GradesPreK_2']
-----
```

```
Shape of matrix after one hot encoding train : (49041, 4)
Shape of matrix after one hot encoding test : (36052, 4)
Shape of matrix after one hot encoding cv : (24155, 4)
```

1.5. Vectorizing Text and Numerical

1.5.1 Vectorization of essays_tfidf:: Train , Test and CV

In [54]:

```
# we are considering only the words which appeared in at least 10 documents (rows or projects)
from sklearn.feature_extraction.text import TfidfVectorizer
vectorizer_essays_tfidf = TfidfVectorizer(preprocessed_essays, min_df=10, max_features=5000)

essays_tfidf_train = vectorizer_essays_tfidf.fit_transform(X_train['preprocessed_essays'])
essays_tfidf_test = vectorizer_essays_tfidf.transform(X_test['preprocessed_essays'].values)
essays_tfidf_cv = vectorizer_essays_tfidf.transform(X_cv['preprocessed_essays'].values)

print("Shape of matrix after one hot encoding of train : ",essays_tfidf_train.shape)
print("Shape of matrix after one hot encoding test : ",essays_tfidf_test.shape)
print("Shape of matrix after one hot encoding cv : ",essays_tfidf_cv.shape)
```

```
Shape of matrix after one hot encoding of train : (49041, 5000)
Shape of matrix after one hot encoding test : (36052, 5000)
Shape of matrix after one hot encoding cv : (24155, 5000)
```

1.5.2 Vectorization of project_title tfidf ::Train, Test and Cv

In [55]:

```
# we are considering only the words which appeared in at least 10 documents (rows or pr
from sklearn.feature_extraction.text import TfidfVectorizer
vectorizer_project_title_tfidf = TfidfVectorizer(preprocessed_project_title, min_df=10,

project_title_tfidf_train = vectorizer_project_title_tfidf.fit_transform(X_train['proje
project_title_tfidf_test  = vectorizer_project_title_tfidf.transform(X_test['project_ti
project_title_tfidf_cv    = vectorizer_project_title_tfidf.transform(X_cv['project_titl

print("Shape of matrix after one hot encodig of train : ",project_title_tfidf_train.sha
print("Shape of matrix after one hot encodig test      : ",project_title_tfidf_test.sha
print("Shape of matrix after one hot encodig cv        : ",project_title_tfidf_cv.shape)
```

Shape of matrix after one hot encodig of train : (49041, 2111)

Shape of matrix after one hot encodig test : (36052, 2111)

Shape of matrix after one hot encodig cv : (24155, 2111)

1.5.3 Vectorization of price :: train, test and cv

In [103]:

```
# Link: https://scikit-learn.org/stable/modules/generated/sklearn.preprocessing.Normalizer.html
# https://docs.scipy.org/doc/numpy/reference/generated/numpy.reshape.html
# Reshaping price data using array.reshape(-1, 1)
price_normalizer = Normalizer()
price_normalizer_train = price_normalizer.fit_transform(X_train['price'].values.reshape(1,-1))
price_normalizer_test = price_normalizer.transform(X_test['price'].values.reshape(1,-1))
price_normalizer_cv = price_normalizer.transform(X_cv['price'].values.reshape(1,-1))

print("shape of price_normalizer_train:", price_normalizer_train.shape)
print("-----")
print(price_normalizer_train)

print("shape of price_normalizer_test :", price_normalizer_test.shape)
print("-----")
print(price_normalizer_test)

print("shape of price_normalizer_cv :", price_normalizer_cv.shape)
print("-----")
print(price_normalizer_cv)
```

shape of price_normalizer_train: (49041, 1)

```
[[0.0017128 ]
 [0.00052738]
 [0.00133911]
 ...
 [0.0006195 ]
 [0.00622391]
 [0.00297778]]
```

shape of price_normalizer_test : (36052, 1)

```
[[0.00173636]
 [0.00245599]
 [0.00017022]
 ...
 [0.00519457]
 [0.00457257]
 [0.00179421]]
```

shape of price_normalizer_cv : (24155, 1)

```
[[0.00398983]
 [0.00302489]
 [0.00184272]
 ...
 [0.00373467]
 [0.01030215]
 [0.00143378]]
```

1.5.4 Vectorization of teacher number of previously posted projects :: train , test , cv

In [104]:

```
# Link: https://scikit-learn.org/stable/modules/generated/sklearn.preprocessing.Normalizer
# https://docs.scipy.org/doc/numpy/reference/generated/numpy.reshape.html
# Reshaping price data using array.reshape(-1, 1)
tnppp_normalizer = Normalizer()
tnppp_normalizer_train = tnppp_normalizer.fit_transform(X_train['teacher_number_of_previously_posted_projects'])
tnppp_normalizer_test = tnppp_normalizer.transform(X_test['teacher_number_of_previously_posted_projects'])
tnppp_normalizer_cv = tnppp_normalizer.transform(X_cv['teacher_number_of_previously_posted_projects'])
print("shape of teacher_number_of_previously_posted_projects_normalizer_train:", tnppp_normalizer_train)
print("-----")
print(tnppp_normalizer_train)

print("shape of teacher_number_of_previously_posted_projects_normalizer_test :", tnppp_normalizer_test)
print("-----")
print(tnppp_normalizer_test)

print("shape of teacher_number_of_previously_posted_projects_normalizer_cv      :", tnppp_normalizer_cv)
print("-----")
print(tnppp_normalizer_cv)
```

shape of teacher_number_of_previously_posted_projects_normalizer_train: (49041, 1)

```
[[0.          ]
 [0.00343163]
 [0.00268562]
 ...
 [0.00074601]
 [0.0005968 ]
 [0.          ]]
```

shape of teacher_number_of_previously_posted_projects_normalizer_test : (36052, 1)

```
[[0.00160903]
 [0.          ]
 [0.          ]
 ...
 [0.00071512]
 [0.          ]
 [0.00035756]]
```

shape of teacher_number_of_previously_posted_projects_normalizer_cv : (24155, 1)

```
[[0.0004296 ]
 [0.00365163]
 [0.          ]
 ...
 [0.0006444 ]
 [0.0006444 ]
 [0.          ]]
```

1.5.5 Vectorization of quantity : train, test, cv

In [111]:

```
# Link: https://scikit-learn.org/stable/modules/generated/sklearn.preprocessing.Normalizer
normalizer = Normalizer()
quantity_normalizer_train = normalizer.fit_transform(X_train['quantity'].values.reshape(
quantity_normalizer_test = normalizer.transform(X_test['quantity'].values.reshape(1,-1)
quantity_normalizer_cv = normalizer.transform(X_cv['quantity'].values.reshape(1,-1)).

print("Shape of quantity nomalizer train :", quantity_normalizer_train.shape)
print(quantity_normalizer_train)
print("-----")
print(quantity_normalizer_test)
print("Shape of quantity nomalizer test :", quantity_normalizer_test.shape)
print("-----")
print(quantity_normalizer_cv)
print("Shape of quantity nomalizer cv :", quantity_normalizer_cv.shape)
```

Shape of quantity nomalizer train : (49041, 1)

```
[[0.00028972]
 [0.00028972]
 [0.00115888]
 ...
 [0.0028972 ]
 [0.00478037]
 [0.00130374]]
```

```
-----
[[0.00283631]
 [0.0020021 ]
 [0.03003147]
 ...
 [0.00650682]
 [0.00083421]
 [0.00316999]]
```

Shape of quantity nomalizer test : (36052, 1)

```
-----
[[0.00901331]
 [0.00083845]
 [0.00230573]
 ...
 [0.00838447]
 [0.00020961]
 [0.00167689]]
```

Shape of quantity nomalizer cv : (24155, 1)

1.6 concatenating essay + project_title

In [56]:

```
# After preprocessing concatenating essay + project_title
concat_essays_titles = (list(X_train['preprocessed_essays'])+list(X_train['preprocessed
len(concat_essays_titles)
```

Out[56]:

98082

1.6.1 Calculating the idf values for every word in combined text

In [59]:

```
tfidf_vec = TfidfVectorizer()  
tfidf_vec.fit_transform(concat_essays_titles)
```

Out[59]:

```
<98082x42148 sparse matrix of type '<class 'numpy.float64'>'  
  with 4011283 stored elements in Compressed Sparse Row format>
```

In [208]:

```
idf_score = tfidf_vec.idf_  
features_names = tfidf_vec.get_feature_names()
```

1.6.2 sorting vocab based on idf value in descending order

1.6.3 Selecting top 2000 words

In [198]:

```
idf_score_features = []  
for i in range(len(idf_score)):  
    idf_score_features.append([idf_score[i], features_names[i]])  
idf_sort = idf_score_features.sort(reverse=True)  
idf_score_features=idf_score_features[:2000]
```

In []:

```
n = 2000  
topFeat = [feat[i] for i in ind[:n]]  
topIDF = [tfidfvect.idf_[i] for i in ind[:n]]  
featIDF = list(zip(topFeat,topIDF))  
print(featIDF[:20])
```

Preparing data for models

In [149]:

```
project_data.columns
```

Out[149]:

```
Index(['Unnamed: 0', 'id', 'teacher_id', 'teacher_prefix', 'school_state',  
      'Date', 'project_grade_category', 'project_title', 'project_essay_  
1',  
      'project_essay_2', 'project_essay_3', 'project_essay_4',  
      'project_resource_summary',  
      'teacher_number_of_previously_posted_projects', 'clean_categories',  
      'clean_subcategories', 'essay', 'price', 'quantity',  
      'preprocessed_essays', 'preprocessed_project_title'],  
      dtype='object')
```

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
- 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
- essays : Text
- project_titiel : text

Assignment 11: TruncatedSVD

- **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 `idf_` (https://scikit-learn.org/stable/modules/generated/sklearn.feature_extraction.text.TfidfVectorizer.html) values
- **step 2** Compute the co-occurence matrix with these 2k words, with window size=5 ([ref](https://www.analyticsvidhya.com/blog/2017/06/word-embeddings-count-word2veec/) (<https://www.analyticsvidhya.com/blog/2017/06/word-embeddings-count-word2veec/>))

```
the cat sat on the wall
window=1
```

	the	cat	sat	on	wall
the	1	1	0	0	1
cat	1	1	1	0	0
sat	0	1	1	1	0
on	1	0	1	1	0
wall	1	0	0	0	1

- **step 3** Use [TruncatedSVD](http://scikit-learn.org/stable/modules/generated/sklearn.decomposition.TruncatedSVD.html) (<http://scikit-learn.org/stable/modules/generated/sklearn.decomposition.TruncatedSVD.html>) on calculated co-occurence matrix and reduce its dimensions, choose the number of components (`n_components`) using [elbow method](https://www.appliedaicourse.com/course/applied-ai-course-online/lessons/pca-code-example-using-non-visualization/) (<https://www.appliedaicourse.com/course/applied-ai-course-online/lessons/pca-code-example-using-non-visualization/>).

- The shape of the matrix after TruncatedSVD will be $2000 \times n$, 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 (<https://www.kdnuggets.com/2017/03/simple-xgboost-tutorial-iris-dataset.html>)**
- **step 6**:Hyper parameter tuning (Consider any two hyper parameters)
 - Find the best hyper parameter which will give the maximum [AUC](https://www.appliedaicourse.com/course/applied-ai-course-online/lessons/receiver-operating-characteristic-curve-roc-curve-and-auc-1/) 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

In [131]:

```
pip install xgboost
```

Collecting xgboost

Downloading https://files.pythonhosted.org/packages/5e/49/b95c037b717b4ceadc76b6e164603471225c27052d1611d5a2e832757945/xgboost-0.90-py2.py3-none-win_amd64.whl (18.3MB)

Requirement already satisfied: scipy in c:\programdata\anaconda3\lib\site-packages (from xgboost) (1.3.1)

Requirement already satisfied: numpy in c:\programdata\anaconda3\lib\site-packages (from xgboost) (1.17.1)

Installing collected packages: xgboost

Successfully installed xgboost-0.90

Note: you may need to restart the kernel to use updated packages.

In [150]:

```
import sys
import math

import numpy as np
from sklearn.model_selection 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)

    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
```

2. TruncatedSVD

Testing Co-occurrence matrix code before applying it on top 2000 words which are occurred rarely in given corpus

In [242]:

```
import numpy as np
corpus = ["ABC DEF IJK PQR", "PQR KLM OPQ", "LMN PQR XYZ ABC DEF PQR ABC"]
top_words = ["ABC", "PQR", "DEF" ]
window_size=2

df = pd.DataFrame()
df['concat_essays_titles'] = corpus
df.head()
```

Out[242]:

	concat_essays_titles
0	ABC DEF IJK PQR
1	PQR KLM OPQ
2	LMN PQR XYZ ABC DEF PQR ABC

In [231]:

```
#
coo_matrix_3x3=(np.zeros((3,3)))
window=2
```

In [232]:

```
for sentence in concat_essays_titles:
    word_sen=sentence.split()
    for idss,word in enumerate(word_sen):
        if word in top_words:
            for i in range(max(0,idss-window),min(idss+window,len(word_sen))):
                if word_sen[i] in top_words:
                    coo_matrix_3x3[top_words.index(word_sen[i]),top_words.index(word)]+=1
```

In [239]:

```
print(coo_matrix_3x3.shape)
```

(3, 3)

2.1 Computing Co-occurrence matrix

In [182]:

```
coo_matrix=(np.zeros((2000,2000)))
window=5
```

In [183]:

```
final_2000_features=[]
for i in range(2000):
    final_2000_features.append(idf_score_features[i][1])
```

In [184]:

```
for sentence in concat_essays_titles:
    word_sen=sentence.split()
    for idss,word in enumerate(word_sen):
        if word in final_2000_features:
            for i in range(max(0,idss-window),min(idss+window,len(word_sen))):
                if word_sen[i] in final_2000_features:
                    coo_matrix[final_2000_features.index(word_sen[i]),final_2000_features.index(word)] += 1
```

In [69]:

```
coo_matrix = np.array(coo_matrix)
```

In [235]:

```
coo_matrix
```

Out[235]:

```
array([[1., 0., 0., ..., 0., 0., 0.],
       [0., 1., 0., ..., 0., 0., 0.],
       [0., 0., 1., ..., 0., 0., 0.],
       ...,
       [0., 0., 0., ..., 1., 0., 0.],
       [0., 0., 0., ..., 0., 1., 0.],
       [0., 0., 0., ..., 0., 0., 1.]])
```

2.2 Applying TruncatedSVD on co-occurrence matrix for dimension reduction

In [71]:

```
# finding optimal value of n_components(n) using truncated svd
from sklearn.decomposition import TruncatedSVD
n_components=[10,20,50,60,100,200,400,500,1000,1200,1500,1700,1800,1900,1999]
explained_variance=[]
for n in n_components:
    svd=TruncatedSVD(n_components=n,random_state=42)
    svd.fit(coo_matrix)
    exvar=svd.explained_variance_ratio_.sum()
    explained_variance.append(exvar)

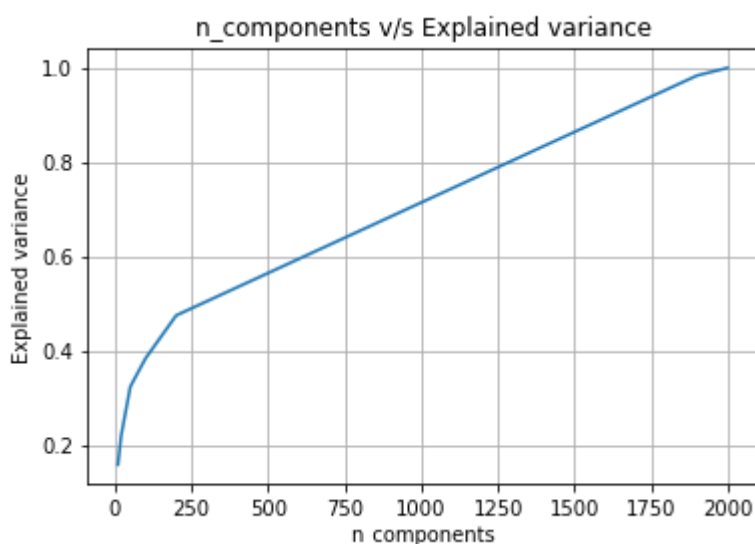
    print('n_components=',n,'variance=',exvar)
```

```
n_components= 10 variance= 0.15780057219930133
n_components= 20 variance= 0.21804522783821934
n_components= 50 variance= 0.32280942985278505
n_components= 60 variance= 0.3353576384645209
n_components= 100 variance= 0.3835141175728213
n_components= 200 variance= 0.4746408973562684
n_components= 400 variance= 0.5346212639665506
n_components= 500 variance= 0.5645759445391754
n_components= 1000 variance= 0.714353292558117
n_components= 1200 variance= 0.77425139377073
n_components= 1500 variance= 0.8641206516318087
n_components= 1700 variance= 0.9240131577340902
n_components= 1800 variance= 0.9539683514807485
n_components= 1900 variance= 0.9839189220531424
n_components= 1999 variance= 0.99999999999999817
```

2.3 Plotting n_components v/s Explained Variance

In [72]:

```
# plotting curve between n_components and explained_variance
plt.plot(n_components, explained_variance)
plt.xlabel('n_components')
plt.ylabel("Explained variance")
plt.title("n_components v/s Explained variance")
plt.grid()
plt.show()
```



In [73]:

```
from sklearn.decomposition import TruncatedSVD
# got 95 % variance vs explained variance - 1800
tsvd=TruncatedSVD(n_components=1800,random_state=42)
final_coo_matrix=tsvd.fit_transform(coo_matrix)
```

In [74]:

```
print(final_coo_matrix.shape)
print(final_coo_matrix)
```

```
(2000, 1800)
[[ 8.34928680e-14 -3.76507005e-13 -3.49006635e-13 ...  6.85346944e-03
 -1.19551037e-02 -1.35526925e-02]
 [-4.12573837e-14  2.96351799e-12  2.20875301e-12 ...  3.91722404e-03
 -5.21262643e-03 -1.70517786e-02]
 [-6.30599387e-14 -1.76923458e-12 -1.58643119e-13 ... -2.49691566e-04
 -1.97192405e-02 -1.42059002e-02]
 ...
 [-5.96701889e-14 -1.15225741e-13 -5.63068311e-13 ...  1.05779710e-03
 -7.49475877e-03 -4.70852759e-03]
 [-5.78546607e-18  6.47476970e-16  6.62649109e-16 ... -2.74097268e-16
  2.89361278e-17  1.11030646e-15]
 [-7.59615871e-19 -2.55055037e-17  4.91507179e-17 ...  8.72979916e-16
 -3.49854234e-16 -8.94614803e-16]]
```

2.4 using avg w2v vectorizer method : essays and project_title

In [75]:

```
model = {}
for i in range(len(final_2000_features)):
    model[final_2000_features[i]] = final_coo_matrix[i]
```

In [76]:

```
# model = final_2000_features
glove_words = set(model.keys())
```

project_title - avg w2v - train

In [80]:

```
# average Word2Vec
# compute average word2vec for each review.
avg_w2v_titles_vectors_train = []; # the avg-w2v for each sentence/review is stored in
for sentence in tqdm(X_train['preprocessed_project_title']): # for each review/sentence
    vector = np.zeros(1800) # as word vectors are of zero length
    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[word]
            cnt_words += 1
    if cnt_words != 0:
        vector /= cnt_words
    avg_w2v_titles_vectors_train.append(vector)
```

100%|██| 49041/49041 [00:00<00:00, 54793.57
it/s]

project_title - avg w2v - test

In [81]:

```
avg_w2v_titles_vectors_test = []; # the avg-w2v for each sentence/review is stored in t
for sentence in tqdm(X_test['preprocessed_project_title']): # for each review/sentence
    vector = np.zeros(1800) # as word vectors are of zero length
    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[word]
            cnt_words += 1
    if cnt_words != 0:
        vector /= cnt_words
    avg_w2v_titles_vectors_test.append(vector)
```

100%|██| 36052/36052 [00:00<00:00, 46368.43
it/s]

essays - avg w2v - train

In [84]:

```
# average Word2Vec
# compute average word2vec for each review.
avg_w2v_essays_vectors_train = []; # the avg-w2v for each sentence/review is stored in
for sentence in tqdm(X_train['preprocessed_essays']): # for each review/sentence
    vector = np.zeros(1800) # as word vectors are of zero length
    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[word]
            cnt_words += 1
    if cnt_words != 0:
        vector /= cnt_words
    avg_w2v_essays_vectors_train.append(vector)
```

100%|██| 49041/49041 [00:04<00:00, 10501.12
it/s]

essays - avg w2v - test

In [83]:

```
avg_w2v_essays_vectors_test = []; # the avg-w2v for each sentence/review is stored in t
for sentence in tqdm(X_test['preprocessed_essays']): # for each review/sentence
    vector = np.zeros(1800) # as word vectors are of zero length
    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[word]
            cnt_words += 1
    if cnt_words != 0:
        vector /= cnt_words
    avg_w2v_essays_vectors_test.append(vector)
```

100%|██| 36052/36052 [00:02<00:00, 12829.70
it/s]

In [99]:

```
print("avg w2v essays vector train:", len(avg_w2v_essays_vectors_train))
print("avg w2v essays vector test:", len(avg_w2v_essays_vectors_test))
print("avg w2v titles vector train:", len(avg_w2v_titles_vectors_train))
print("avg w2v titles vector test:", len(avg_w2v_titles_vectors_test))
```

```
avg w2v essays vector train: 49041
avg w2v essays vector test: 36052
avg w2v titles vector train: 49041
avg w2v titles vector test: 36052
```

2.5 word count project_title

In [87]:

```
title_wordcount_train = []
title_train = list(X_train['preprocessed_project_title'])
for i in tqdm(title_train):
    b = len(str(i).split())
    title_wordcount_train.append(b)
title_wordcount_train = np.array(title_wordcount_train)

title_wordcount_test = []
title_test = list(X_test['preprocessed_project_title'])
for i in tqdm(title_test):
    b = len(str(i).split())
    title_wordcount_test.append(b)
title_wordcount_test = np.array(title_wordcount_test)

print(title_wordcount_train.shape)
print(title_wordcount_test.shape)
```

```
100%|████████████████████████████████████████| 49041/49041 [00:00<00:00, 405274.99
it/s]
100%|████████████████████████████████████████| 36052/36052 [00:00<00:00, 474338.81
it/s]
```

```
(49041,)
(36052,)
```

2.5.1 Standardizing Word counts(TITLES)

In [88]:

```
from sklearn.preprocessing import StandardScaler

title_wordcount_scalar = StandardScaler()
title_wordcount_scalar.fit(title_wordcount_train.reshape(-1,1))

title_wordcount_standardized_train = title_wordcount_scalar.transform(title_wordcount_train)
title_wordcount_standardized_test = title_wordcount_scalar.transform(title_wordcount_test)

print(title_wordcount_standardized_train.shape)
print(title_wordcount_standardized_test.shape)
```

```
(49041, 1)
(36052, 1)
```

2.6 word count essays

In [91]:

```
essay_wordcount_train = []
essay_train = list(X_train['preprocessed_essays'])
for i in tqdm(essay_train):
    b = len(str(i).split())
    essay_wordcount_train.append(b)
essay_wordcount_train = np.array(essay_wordcount_train)

essay_wordcount_test = []
essay_test = list(X_test['preprocessed_essays'])
for i in tqdm(essay_test):
    b = len(str(i).split())
    essay_wordcount_test.append(b)
essay_wordcount_test = np.array(essay_wordcount_test)

print(essay_wordcount_train.shape)
print(essay_wordcount_test.shape)
```

```
100%|████████████████████████████████████████| 49041/49041 [00:00<00:00, 67267.72
it/s]
100%|████████████████████████████████████████| 36052/36052 [00:00<00:00, 68666.53
it/s]

(49041,)
(36052,)
```

2.6.1 Standardizing Word counts(ESSAYS)

In [248]:

```
from sklearn.preprocessing import StandardScaler

essay_wordcount_scalar = StandardScaler()
essay_wordcount_scalar.fit(essay_wordcount_train.reshape(-1,1))

essay_wordcount_standardized_train = essay_wordcount_scalar.transform(essay_wordcount_train)
essay_wordcount_standardized_test = essay_wordcount_scalar.transform(essay_wordcount_test)

print(essay_wordcount_standardized_train.shape)
print(essay_wordcount_standardized_test.shape)
```

```
(49041, 1)
(36052, 1)
```

2.7 Sentiment scores for each essay

In [245]:

```
import nltk
from nltk.sentiment.vader import SentimentIntensityAnalyzer

# Link :https://stackoverflow.com/questions/53680690/List-object-has-no-attribute-encode

vader = SentimentIntensityAnalyzer()
essay_train_str = X_train['preprocessed_essays'].str[0].str.join(" ")
sid = essay_train_str.apply(lambda x: vader.polarity_scores(x)['compound'])

sid_train = sid.values.reshape(1,-1).T
print(sid_train)

print("Shape of sid_train and y_train :", sid_train.shape, y_train.shape)
```

```
[[0.]
 [0.]
 [0.]
 ...
 [0.]
 [0.]
 [0.]]
Shape of sid_train and y_train : (49041, 1) (49041,)
```

In [246]:

```
# sid test

vader = SentimentIntensityAnalyzer()
essay_test_str = X_test['preprocessed_essays'].str[0].str.join(" ")
sid = essay_test_str.apply(lambda x: vader.polarity_scores(x)['compound'])

sid_test = sid.values.reshape(1,-1).T
print(sid_test)

print("Shape of sid_test and y_test :", sid_test.shape, y_test.shape)
```

```
[[0.]
 [0.]
 [0.]
 ...
 [0.]
 [0.]
 [0.]]
Shape of sid_test and y_test : (36052, 1) (36052,)
```

In [247]:

```
# sid cv

vader = SentimentIntensityAnalyzer()
essay_cv_str = X_cv['preprocessed_essays'].str[0].str.join(" ")
sid = essay_cv_str.apply(lambda x: vader.polarity_scores(x)['compound'])
sid_cv = sid.values.reshape(1,-1).T
print(sid_cv)

print("Shape of sid_cv and y_cv :", sid_cv.shape, y_cv.shape)
```

```
[[0.]
 [0.]
 [0.]
 ...
 [0.]
 [0.]
 [0.]]
Shape of sid_cv and y_cv : (24155, 1) (24155,)
```

In [243]:

```
print(clean_categories_one_hot_train.shape)
print(clean_subcategories_one_hot_train.shape)
print(school_state_one_hot_train.shape)
print(teacher_prefix_one_hot_train.shape)
print(project_grade_category_one_hot_train.shape)
print(price_normalizer_train.shape)
print(tnppp_normalizer_train.shape)
print(quantity_normalizer_train.shape)
print(essays_tfidf_train.shape)
print(project_title_tfidf_train.shape)
print(len(avg_w2v_essays_vectors_train))
print(len(avg_w2v_titles_vectors_train))
print(len(title_wordcount_standardized_train))
print(len(essay_wordcount_standardized_train))
print(sid_train.shape)
print(final_coo_matrix.shape)
```

```
(49041, 9)
(49041, 30)
(49041, 51)
(49041, 5)
(49041, 4)
(49041, 1)
(49041, 1)
(49041, 1)
(49041, 5000)
(49041, 2111)
49041
49041
49041
49041
(49041, 1)
(2000, 1800)
```

2.8 Merge matrices

In [122]:

```
# merge two sparse matrices: https://stackoverflow.com/a/19710648/4084039
from scipy.sparse import hstack
set_train = hstack((clean_categories_one_hot_train, clean_subcategories_one_hot_train, school_prefix_one_hot_train, teacher_prefix_one_hot_train, project_grade_category_one_hot_train, price_normalizer_train, tnppp_normalizer_train, quantity_normalizer_train, title_wordcount_standardized_train, essay_wordcount_standardized_train, avg_w2v_essays_vectors_train, avg_w2v_titles_vectors_train)).tocsr()

set_test = hstack((clean_categories_one_hot_test, clean_subcategories_one_hot_test, school_prefix_one_hot_test, teacher_prefix_one_hot_test, project_grade_category_one_hot_test, price_normalizer_test, tnppp_normalizer_test, quantity_normalizer_test, sid_test, title_wordcount_standardized_test, essay_wordcount_standardized_test, avg_w2v_essays_vectors_test, avg_w2v_titles_vectors_test)).tocsr()

print("Final merged datamatrix of set_train:", set_train.shape, y_train.shape)
print("Final merged datamatrix of set_test: ", set_test.shape, y_test.shape)
print("="*100)
```

```
Final merged datamatrix of set_train: (49041, 3705) (49041,)
Final merged datamatrix of set_test: (36052, 3705) (36052,)
=====
=====
```

2.9 Apply XGBoost on the Final Features from the above section

https://xgboost.readthedocs.io/en/latest/python/python_intro.html
(https://xgboost.readthedocs.io/en/latest/python/python_intro.html)

In [140]:

```
%%time
from sklearn.metrics import roc_auc_score
from sklearn.model_selection import cross_val_score
from sklearn.model_selection import GridSearchCV
import matplotlib.pyplot as plt
import xgboost as xgb
import time

start_time = time.time()
gbdt = xgb.XGBClassifier(n_jobs=-1, class_weight='balanced')
parameters = {'n_estimators': [10, 100, 300, 500], "learning_rate": [0.01, 0.1, 1]}
clf1 = GridSearchCV(gbdt, parameters, cv=3, scoring='roc_auc', return_train_score=True)
clf1.fit(set_train, y_train)
print("Execution time: " + str((time.time() - start_time)) + ' ms')
```

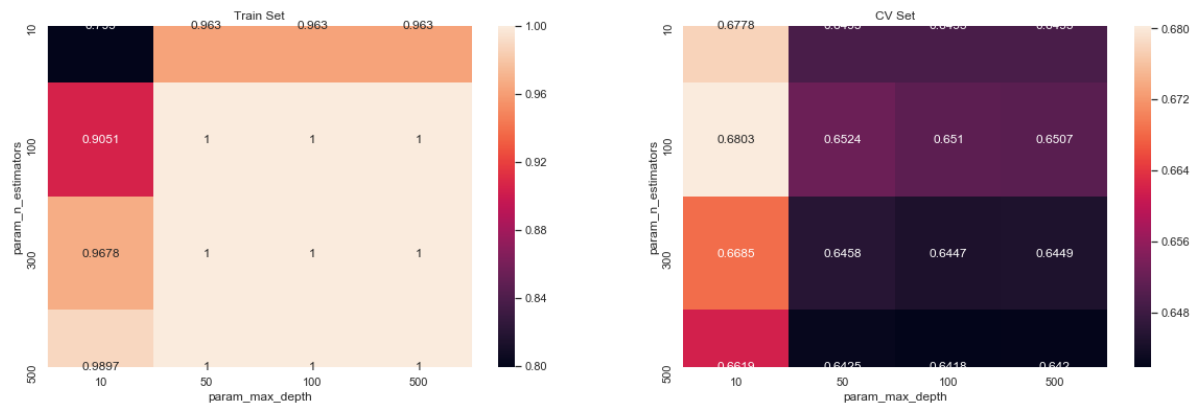
```
Execution time: 23555.665098667145 ms
Wall time: 6h 32min 35s
```

2.9.1 Hyper parameter Tuning to find best estimator :: GridSearchCV set

In [142]:

```
# plotting the performance of model both on train data and cross validation data for each
import seaborn as sns; sns.set()
max_scores1 = pd.DataFrame(clf1.cv_results_).groupby(['param_n_estimators', 'param_learning_rate'])

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()
```



In [144]:

```
# Best tune parameters with score
print(clf1.best_estimator_)
print(clf1.score(set_train,y_train))
print(clf1.score(set_test,y_test))
```

```
XGBClassifier(base_score=0.5, booster='gbtree', class_weight='balanced',
              colsample_bylevel=1, colsample_bynode=1, colsample_bytree=1,
              gamma=0, learning_rate=0.1, max_delta_step=0, max_depth=10,
              min_child_weight=1, missing=None, n_estimators=100, n_jobs=-
1,
              nthread=None, objective='binary:logistic', random_state=0,
              reg_alpha=0, reg_lambda=1, scale_pos_weight=1, seed=None,
              silent=None, subsample=1, verbosity=1)
0.8814326232614123
0.682222849518572
```

2.9.2 Train model using the best hyperparameter

In [145]:

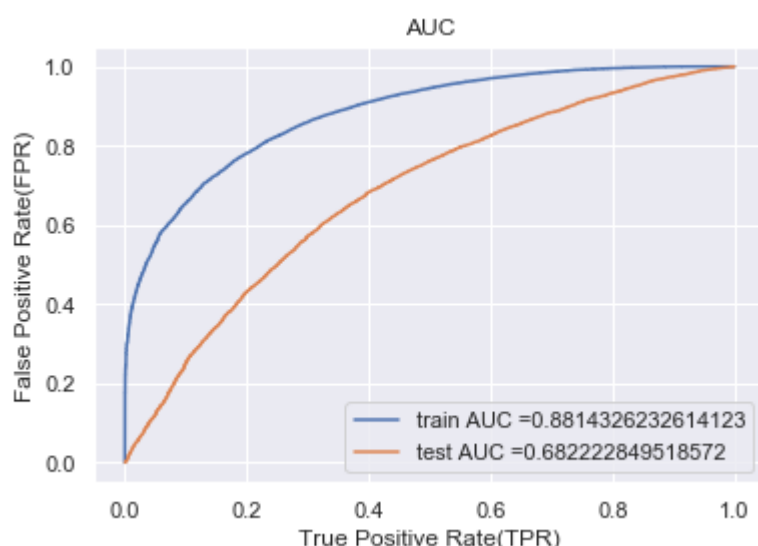
```
%%time
from sklearn.metrics import auc,roc_curve

clf_1v =xgb.XGBClassifier (class_weight = 'balanced',learning_rate=10,n_estimator=100)
clf_1v.fit(set_train,y_train)
# roc_auc_score(y_true, y_score) the 2nd parameter should be probability estimates of t
# not the predicted outputs

train_fpr, train_tpr, thresholds = roc_curve(y_train,clf_1v.predict_proba(set_train)[:,:1])
test_fpr, test_tpr, thresholds = roc_curve(y_test, clf_1v.predict_proba(set_test)[:,:1])

plt.plot(train_fpr, train_tpr, label="train AUC =" +str(auc(train_fpr, train_tpr)))
plt.plot(test_fpr, test_tpr, label="test AUC =" +str(auc(test_fpr, test_tpr)))

plt.legend()
plt.grid(True)
plt.xlabel("False Positive Rate(FPR)")
plt.ylabel("True Positive Rate(TPR)")
plt.title("AUC")
plt.show()
```



Wall time: 5min 3s

2.9.3 Confusion Matrix : set_train and set_test

In [146]:

```
def predict(proba, threshold, fpr, tpr):
    t = threshold[np.argmax(fpr*(1-tpr))]
    # (tpr*(1-fpr)) will be maximum if your fpr is very low and tpr is very high
    print("the maximum value of tpr*(1-fpr)", max(tpr*(1-fpr)), "for threshold", np.round(t, 2))
    predictions = []
    for i in proba:
        if i >= t:
            predictions.append(1)
        else:
            predictions.append(0)
    return predictions
```

In [147]:

```
y_train_pred_1 = clf_1v.predict_proba(set_train)[: ,1]
y_test_pred_1 = clf_1v.predict_proba(set_test)[: ,1]

#https://www.quantinsti.com/blog/creating-heatmap-using-python-seaborn
import seaborn as sns; sns.set()
conf_matr_df_train_1 = confusion_matrix(y_train, predict(y_train_pred_1, thresholds, tpr, fpr))
conf_matr_df_test_1 = confusion_matrix(y_test, predict(y_test_pred_1, thresholds, tpr, fpr))

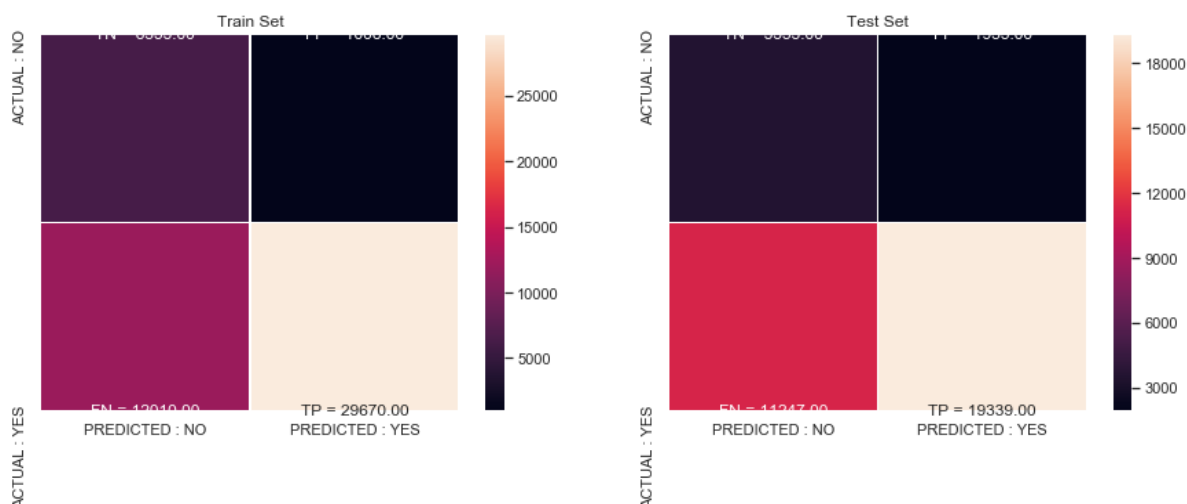
key = (np.asarray(['TN', 'FP'], ['FN', 'TP'])))
fig, ax = plt.subplots(1,2, figsize=(15,5))

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

sns.heatmap(conf_matr_df_train_1, linewidths=.5, xticklabels=['PREDICTED : NO', 'PREDICTED : YES'], yticklabels=['ACTUAL : NO', 'ACTUAL : YES'], annot = labels_train, fmt = '')
sns.heatmap(conf_matr_df_test_1, linewidths=.5, xticklabels=['PREDICTED : NO', 'PREDICTED : YES'], yticklabels=['ACTUAL : NO', 'ACTUAL : YES'], annot = labels_test, fmt = '')

ax[0].set_title('Train Set')
ax[1].set_title('Test Set')
plt.show()
```

the maximum value of tpr*(1-fpr) 0.6258925691530375 for threshold 0.851
the maximum value of tpr*(1-fpr) 0.41074889661678426 for threshold 0.844



3. Conclusion

In [148]:

```
from prettytable import PrettyTable
p = PrettyTable()
p.field_names = ["Vectorizer", "Model", "Hyper Parameter", "AUC"]
p.add_row(["AVG W2V", "XGBoost", "Max Depth:10 , n_estimators:100", 0.68])
print(p)
```

Vectorizer	Model	Hyper Parameter	AUC
AVG W2V	XGBoost	Max Depth:10 , n_estimators:100	0.68

Thank You.

[Sign Off Ramesh Battu \(https://www.linkedin.com/in/rameshbattuai/\)](https://www.linkedin.com/in/rameshbattuai/)