DonorsChoose

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

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

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

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

About the DonorsChoose Data Set

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

Feature	Description
project_id	A unique identifier for the proposed project. Example: p036502
	Title of the project. Examples:
project_title	Art Will Make You Happy!
	• First Grade Fun
	Grade level of students for which the project is targeted. One of the following enumerated values:
project grade category	• Grades PreK-2
project_grade_category	• Grades 3-5
	• Grades 6-8
	• Grades 9-12
	One or more (comma-separated) subject categories for the project from the following enumerated list of values:
	Applied Learning
	• Care & Hunger
	• Health & Sports
	History & Civics
	• Literacy & Language
project_subject_categories	• Math & Science
	• Music & The Arts
	• Special Needs
	• Warmth
	Examples:
	• Music & The Arts
	• Literacy & Language, Math & Science
school_state	State where school is located (<u>Two-letter U.S. postal code</u>). Example
	One or more (comma-separated) subject subcategories for the project
project_subject_subcategories	Examples:
	• Literacy

Feature	• Literature & Writing, Social Sciences Description				
project_resource_summary	An explanation of the resources needed for the project. Example: • My students need hands on literacy materials to manage sensory needs!				
project_essay_1	First application essay [*]				
project_essay_2	Second application essay*				
project_essay_3	Third application essay*				
project_essay_4	Fourth application essay*				
project_submitted_datetime	Datetime when project application was submitted. Example: 2016–04–28 12:43:56.245				
teacher_id	A unique identifier for the teacher of the proposed project. Example: bdf8baa8fedef6bfeec7ae4ff1c15c56				
teacher_prefix	Teacher's title. One of the following enumerated values: • nan • Dr. • Mr. • Mrs. • Ms. • Teacher.				
teacher_number_of_previously_posted_projects	Number of project applications previously submitted by the same teacher. Example: 2				

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

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

Feature	Description					
id	project_id value from the train.csv file. Example: p036502					
description Desciption of the resource. Example: Tenor Saxophone Reeds, Box of						
quantity Quantity of the resource required. Example: 3						
price	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
project is approved	A binary flag indicating whether DonorsChoose approved the project. A value of 0 indicates the project
project_is_approved	was not approved, and a value of 1 indicates the project was approved.

Notes on the Essay Data

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

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

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

• __project_essay_1:__ "Describe your students: What makes your students special? Specific details about their background, your neighborhood, and your school are all helpful."

your neignbornoou, and your sonoor are an neighb.

 __project_essay_2:__ "About your project: How will these materials make a difference in your students' learning and improve their school lives?"

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

In [1]:

```
!pip install chart studio
Collecting chart studio
 Downloading
https://files.pythonhosted.org/packages/ca/ce/330794a6b6ca4b9182c38fc69dd2a9cbff60fd49421cb8648ee5f
2dc/chart studio-1.1.0-py3-none-any.whl (64kB)
                                     | 71kB 2.1MB/s
Requirement already satisfied: plotly in /usr/local/lib/python3.6/dist-packages (from
chart studio) (4.4.1)
Requirement already satisfied: six in /usr/local/lib/python3.6/dist-packages (from chart studio)
Requirement already satisfied: requests in /usr/local/lib/python3.6/dist-packages (from
chart studio) (2.21.0)
Requirement already satisfied: retrying>=1.3.3 in /usr/local/lib/python3.6/dist-packages (from
chart_studio) (1.3.3)
Requirement already satisfied: certifi>=2017.4.17 in /usr/local/lib/python3.6/dist-packages (from
requests->chart_studio) (2020.4.5.1)
Requirement already satisfied: idna<2.9,>=2.5 in /usr/local/lib/python3.6/dist-packages (from
requests->chart studio) (2.8)
Requirement already satisfied: chardet<3.1.0,>=3.0.2 in /usr/local/lib/python3.6/dist-packages
(from requests->chart studio) (3.0.4)
Requirement already satisfied: urllib3<1.25,>=1.21.1 in /usr/local/lib/python3.6/dist-packages
(from requests->chart studio) (1.24.3)
Installing collected packages: chart-studio
Successfully installed chart-studio-1.1.0
4
```

In [0]:

```
import chart_studio.plotly as py
import plotly.graph_objs as go
```

In [3]:

```
%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
import plotly.offline as offline
import plotly.graph_objs as go
offline.init notebook mode()
from collections import Counter
1.1 Reading Data
In [5]:
from google.colab import drive
drive.mount('/content/drive')
Go to this URL in a browser: https://accounts.google.com/o/oauth2/auth?client id=947318989803-6bn6
qk8qdgf4n4g3pfee6491hc0brc4i.apps.googleusercontent.com&redirect uri=urn%3aietf%3awg%3aoauth%3a2.0%
\verb|b&response_type=code&scope=email*20https*3a*2f*2fwww.googleapis.com*2fauth*2fdocs.test*20https*3a*2f*2fwww.googleapis.com*2fauth*2fdocs.test*20https*3a*2f*2fwww.googleapis.com*2fauth*2fdocs.test*20https*3a*2f*2fwww.googleapis.com*2fauth*2fdocs.test*20https*3a*2f*2fwww.googleapis.com*2fauth*2fdocs.test*20https*3a*2f*2fwww.googleapis.com*2fauth*2fdocs.test*20https*3a*2f*2fwww.googleapis.com*2fauth*2fdocs.test*20https*3a*2f*2fwww.googleapis.com*2fauth*2fdocs.test*20https*3a*2f*2fwww.googleapis.com*2fauth*2fdocs.test*20https*3a*2f*2fwww.googleapis.com*2fauth*2fdocs.test*20https*3a*2f*2fwww.googleapis.com*2fauth*2fdocs.test*20https*3a*2f*2fwww.googleapis.com*2fauth*2fdocs.test*20https*3a*2f*2fwww.googleapis.com*2fauth*2fdocs.test*2fauth*2fauth*2fdocs.test*2fauth*2fauth*2fdocs.test*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2fauth*2f
www.googleapis.com%2fauth%2fdrive%20https%3a%2f%2fwww.googleapis.com%2fauth%2fdrive.photos.readonly
ttps%3a%2f%2fwww.googleapis.com%2fauth%2fpeopleapi.readonly
Enter your authorization code:
Mounted at /content/drive
4
In [0]:
project data = pd.read csv('/content/drive/My Drive/Assignments DonorsChoose 2018/train data.csv')
resource data = pd.read csv('/content/drive/My Drive/Assignments DonorsChoose 2018/resources.csv')
In [7]:
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 [8]:
# 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]
project_data.head(2)
```

Out[8]:

	Unnamed: 0	id	teacher_id	teacher_prefix	school_state	Date	project_grade_cate

55660	Unnamed: 8393 0	p2054 [†] 9	2bf07ba08945e5d8b2a3f269b2b3cfe5	teacher_prefix Mrs.	school_state		project grade cate Grades Flek-2
						00:27:36	
76127	37728	p043609	3f60494c61921b3b43ab61bdde2904df	Ms.	UT	2016- 04-27 00:31:25	Grades 3-5
4							F

In [9]:

```
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[9]:

	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

1.2 preprocessing of project subject categories

In [0]:

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

1.3 preprocessing of project subject subcategories

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

Preprocessing of project grade category

In [0]:

1.3 Text preprocessing

```
In [0]:
```

```
project data.head(2)
```

Out[14]:

	Unnamed:	id	teacher_id	teacher_prefix	school_state	Date	project_title	projec
55660	8393	p205479	2bf07ba08945e5d8b2a3f269b2b3cfe5	Mrs.	CA	2016- 04-27 00:27:36	Engineering STEAM into the Primary Classroom	I have fortuna to use
76127	37728	p043609	3f60494c61921b3b43ab61bdde2904df	Ms.	UT	2016- 04-27 00:31:25	1 ools for	Imagin 9 year: You're th

In [15]:

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

I have been fortunate enough to use the Fairy Tale STEM kits in my classroom as well as the STEM j ournals, which my students really enjoyed. I would love to implement more of the Lakeshore STEM k its in my classroom for the next school year as they provide excellent and engaging STEM lessons.My students come from a variety of backgrounds, including language and socioeconomic statu s. Many of them don't have a lot of experience in science and engineering and these kits give me the materials to provide these exciting opportunities for my students. Each month I try to do several science or STEM/STEAM projects. I would use the kits and robot to help guide my science i nstruction in engaging and meaningful ways. I can adapt the kits to my current language arts paci ng guide where we already teach some of the material in the kits like tall tales (Paul Bunyan) or Johnny Appleseed. The following units will be taught in the next school year where I will implement these kits: magnets, motion, sink vs. float, robots. I often get to these units and don additional ideas, strategies, and lessons to prepare my students in science. It is challenging to d evelop high quality science activities. These kits give me the materials I need to provide my students with science activities that will go along with the curriculum in my classroom. Although I have some things (like magnets) in my classroom, I don't know how to use them effectively. The kits will provide me with the right amount of materials and show me how to use them in an appropriate way.

I teach high school English to students with learning and behavioral disabilities. My students all vary in their ability level. However, the ultimate goal is to increase all students literacy level s. This includes their reading, writing, and communication levels.I teach a really dynamic group o f students. However, my students face a lot of challenges. My students all live in poverty and in a dangerous neighborhood. Despite these challenges, I have students who have the the desire to def eat these challenges. My students all have learning disabilities and currently all are performing below grade level. My students are visual learners and will benefit from a classroom that fulfills their preferred learning style. The materials I am requesting will allow my students to be prepared for the classroom with the necessary supplies. Too often I am challenged with students who come t o school unprepared for class due to economic challenges. I want my students to be able to focus on learning and not how they will be able to get school supplies. The supplies will last all year . Students will be able to complete written assignments and maintain a classroom journal. The ch art paper will be used to make learning more visual in class and to create posters to aid students in their learning. The students have access to a classroom printer. The toner will be used to pr int student work that is completed on the classroom Chromebooks. I want to try and remove all barri ers for the students learning and create opportunities for learning. One of the biggest barriers i

s the students not having the resources to get pens, paper, and folders. My students will be able to increase their literacy skills because of this project.

\"Life moves pretty fast. If you don't stop and look around once in awhile, you could miss it.\" from the movie, Ferris Bueller's Day Off. Think back...what do you remember about your grandparents? How amazing would it be to be able to flip through a book to see a day in their lives?My second graders are voracious readers! They love to read both fiction and nonfiction books . Their favorite characters include Pete the Cat, Fly Guy, Piggie and Elephant, and Mercy Watson. They also love to read about insects, space and plants. My students are hungry bookworms! My stude nts are eager to learn and read about the world around them. My kids love to be at school and are like little sponges absorbing everything around them. Their parents work long hours and usually do not see their children. My students are usually cared for by their grandparents or a family friend. Most of my students do not have someone who speaks English at home. Thus it is difficult f or my students to acquire language. Now think forward... wouldn't it mean a lot to your kids, nieces or nephews or grandchildren, to be able to see a day in your life today 30 years from now? Memories are so precious to us and being able to share these memories with future generations will be a rewarding experience. As part of our social studies curriculum, students will be learning ab out changes over time. Students will be studying photos to learn about how their community has ch anged over time. In particular, we will look at photos to study how the land, buildings, clothing, and schools have changed over time. As a culminating activity, my students will capture a slice of their history and preserve it through scrap booking. Key important events in their young lives will be documented with the date, location, and names. Students will be using photos from home and from school to create their second grade memories. Their scrap books will preserve their unique stories for future generations to enjoy. Your donation to this project will provide my second graders with an opportunity to learn about social studies in a fun and creative manner. Th rough their scrapbooks, children will share their story with others and have a historical document for the rest of their lives.

\"A person's a person, no matter how small.\" (Dr.Seuss) I teach the smallest students with the bi ggest enthusiasm for learning. My students learn in many different ways using all of our senses an d multiple intelligences. I use a wide range of techniques to help all my students succeed. \r\nSt udents in my class come from a variety of different backgrounds which makes for wonderful sharing of experiences and cultures, including Native Americans.\r\nOur school is a caring community of su ccessful learners which can be seen through collaborative student project based learning in and ou t of the classroom. Kindergarteners in my class love to work with hands-on materials and have many different opportunities to practice a skill before it is mastered. Having the social skills to wor k cooperatively with friends is a crucial aspect of the kindergarten curriculum. Montana is the perfect place to learn about agriculture and nutrition. My students love to role play in our pretend kitchen in the early childhood classroom. I have had several kids ask me, \"Can we try coo king with REAL food?\" I will take their idea and create \"Common Core Cooking Lessons\" where we learn important math and writing concepts while cooking delicious healthy food for snack time. My students will have a grounded appreciation for the work that went into making the food and knowled ge of where the ingredients came from as well as how it's healthy for their bodies. This project w ould expand our learning of nutrition and agricultural cooking recipes by having us peel our own a pples to make homemade applesauce, make our own bread, and mix up healthy plants from our classroo m garden in the spring. We will also create our own cookbooks to be printed and shared with famili es. \r\nStudents will gain math and literature skills as well as a life long enjoyment for healthy

My classroom consists of twenty-two amazing sixth graders from different cultures and backgrounds. They are a social bunch who enjoy working in partners and working with groups. They are hard-worki $\ensuremath{\text{ng}}$ and eager to head to middle school next year. My job is to get them ready to make this transition and make it as smooth as possible. In order to do this, my students need to come to school every day and feel safe and ready to learn. Because they are getting ready to head to middle school, I give them lots of choice- choice on where to sit and work, the order to complete assignments, choice of projects, etc. Part of the students feeling safe is the ability for them to come into a welcoming, encouraging environment. My room is colorful and the atmosphere is casual. I want them to take ownership of the classroom because we ALL share it together. Because my time w ith them is limited, I want to ensure they get the most of this time and enjoy it to the best of t heir abilities. Currently, we have twenty-two desks of differing sizes, yet the desks are similar t o the ones the students will use in middle school. We also have a kidney table with crates for sea ting. I allow my students to choose their own spots while they are working independently or in groups. More often than not, most of them move out of their desks and onto the crates. Believe it or not, this has proven to be more successful than making them stay at their desks! It is because of this that I am looking toward the "Flexible Seating" option for my classroom.\r\n The students look forward to their work time so they can move around the room. I would like to get rid of the c onstricting desks and move toward more "fun" seating options. I am requesting various seating so m y students have more options to sit. Currently, I have a stool and a papasan chair I inherited fro m the previous sixth-grade teacher as well as five milk crate seats I made, but I would like to gi ve them more options and reduce the competition for the "good seats". I am also requesting two rug s as not only more seating options but to make the classroom more welcoming and appealing. In orde r for my students to be able to write and complete work without desks, I am requesting a class set of clipboards. Finally, due to curriculum that requires groups to work together, I am requesting t ables that we can fold up when we are not using them to leave more room for our flexible seating o ptions.\r\nI know that with more seating options, they will be that much more excited about coming to school! Thank you for your support in making my classroom one students will remember forever!nannan

In [0]:

```
# https://stackoverflow.com/a/47091490/4084039
import re
def decontracted(phrase):
   # specific
   phrase = re.sub(r"won't", "will not", phrase)
   phrase = re.sub(r"can\'t", "can not", phrase)
   # general
   phrase = re.sub(r"n\'t", " not", phrase)
   phrase = re.sub(r"\'re", " are", phrase)
   phrase = re.sub(r"\'s", " is", phrase)
   phrase = re.sub(r"\'d", " would", phrase)
   phrase = re.sub(r"\'ll", " will", phrase)
   phrase = re.sub(r"\'t", " not", phrase)
   phrase = re.sub(r"\'ve", " have", phrase)
   phrase = re.sub(r"\'m", " am", phrase)
   return phrase
```

In [17]:

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

\"A person is a person, no matter how small.\" (Dr.Seuss) I teach the smallest students with the b iggest enthusiasm for learning. My students learn in many different ways using all of our senses a nd multiple intelligences. I use a wide range of techniques to help all my students succeed. \r\nS tudents in my class come from a variety of different backgrounds which makes for wonderful sharing of experiences and cultures, including Native Americans.\r\nOur school is a caring community of su ccessful learners which can be seen through collaborative student project based learning in and ou t of the classroom. Kindergarteners in my class love to work with hands-on materials and have many different opportunities to practice a skill before it is mastered. Having the social skills to wor k cooperatively with friends is a crucial aspect of the kindergarten curriculum. Montana is the perfect place to learn about agriculture and nutrition. My students love to role play in our pretend kitchen in the early childhood classroom. I have had several kids ask me, \"Can we try coo king with REAL food?\" I will take their idea and create \"Common Core Cooking Lessons\" where we learn important math and writing concepts while cooking delicious healthy food for snack time. My students will have a grounded appreciation for the work that went into making the food and knowled ge of where the ingredients came from as well as how it is healthy for their bodies. This project would expand our learning of nutrition and agricultural cooking recipes by having us peel our own apples to make homemade applesauce, make our own bread, and mix up healthy plants from our classro om garden in the spring. We will also create our own cookbooks to be printed and shared with famil ies. \r\nStudents will gain math and literature skills as well as a life long enjoyment for health v cooking.nannan

In [18]:

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

A person is a person, no matter how small. (Dr.Seuss) I teach the smallest students with the big gest enthusiasm for learning. My students learn in many different ways using all of our senses and multiple intelligences. I use a wide range of techniques to help all my students succeed. Students in my class come from a variety of different backgrounds which makes for wonderful sharing of experiences and cultures, including Native Americans. Our school is a caring community of successful learners which can be seen through collaborative student project based learning in a nd out of the classroom. Kindergarteners in my class love to work with hands-on materials and have many different opportunities to practice a skill before it is mastered. Having the social skills to work cooperatively with friends is a crucial aspect of the kindergarten curriculum. Montana is the perfect place to learn about agriculture and nutrition. My students love to role play in our p retend kitchen in the early childhood classroom. I have had several kids ask me, Can we try cooking with REAL food? I will take their idea and create Common Core Cooking Lessons where we learn important math and writing concepts while cooking delicious healthy food for snack time. My students will have a grounded appreciation for the work that went into making the food and knowled

ge of where the ingredients came from as well as how it is healthy for their bodies. This project would expand our learning of nutrition and agricultural cooking recipes by having us peel our own apples to make homemade applesauce, make our own bread, and mix up healthy plants from our classro om garden in the spring. We will also create our own cookbooks to be printed and shared with famil ies. Students will gain math and literature skills as well as a life long enjoyment for healthy cooking.nannan

In [19]:

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

A person is a person no matter how small Dr Seuss I teach the smallest students with the biggest enthusiasm for learning My students learn in many different ways using all of our senses and multi ple intelligences I use a wide range of techniques to help all my students succeed Students in my class come from a variety of different backgrounds which makes for wonderful sharing of experiences and cultures including Native Americans Our school is a caring community of successful learners which can be seen through collaborative student project based learning in and out of the classroom Kindergarteners in my class love to work with hands on materials and have many different opportunities to practice a skill before it is mastered Having the social skills to work cooperatively with friends is a crucial aspect of the kindergarten curriculum Montana is the perfect place to learn about agriculture and nutrition My students love to role play in our pretend kitchen in the early childhood classroom I have had several kids ask me Can we try cooking with REAL food I will take their idea and create Common Core Cooking Lessons where we learn important math and writing concepts while cooking delicious healthy food for snack time My students will have a grounded appreciation for the work that went into making the food and knowled ge of where the ingredients came from as well as how it is healthy for their bodies This project w ould expand our learning of nutrition and agricultural cooking recipes by having us peel our own a pples to make homemade applesauce make our own bread and mix up healthy plants from our classroom garden in the spring We will also create our own cookbooks to be printed and shared with families Students will gain math and literature skills as well as a life long enjoyment for healthy cooking nannan

In [0]:

```
# https://gist.github.com/sebleier/554280
# we are removing the words from the stop words list: 'no', 'nor', 'not'
stopwords= ['i', 'me', 'my', 'myself', 'we', 'our', 'ours', 'ourselves', 'you', "you're", "you've",
                           "you'll", "you'd", 'your', 'yourself', 'yourselves', 'he', 'him', 'his',
'himself', \
                           'she', "she's", 'her', 'hers', 'herself', 'it', "it's", 'its', 'itself', 'they', 'them',
'their',\
                           'theirs', 'themselves', 'what', 'which', 'who', 'whom', 'this', 'that', "that'll",
'these', 'those', \
                           'am', 'is', 'are', 'was', 'were', 'be', 'been', 'being', 'have', 'has', 'had', 'having',
'do', 'does', \
                           'did', 'doing', 'a', 'an', 'the', 'and', 'but', 'if', 'or', 'because', 'as', 'until', '
while', 'of', \
                           'at', 'by', 'for', 'with', 'about', 'against', 'between', 'into', 'through', 'during',
'before', 'after',\
                           'above', 'below', 'to', 'from', 'up', 'down', 'in', 'out', 'on', 'off', 'over', 'under'
, 'again', 'further',\
                           'then', 'once', 'here', 'there', 'when', 'where', 'why', 'how', 'all', 'any', 'both', '\( \)
ach', 'few', 'more',\
                           'most', 'other', 'some', 'such', 'only', 'own', 'same', 'so', 'than', 'too', 'very', \
                           's', 't', 'can', 'will', 'just', 'don', "don't", 'should', "should've", 'now', 'd', 'll'
, 'm', 'o', 're', \
                           've', 'y', 'ain', 'aren', "aren't", 'couldn', "couldn't", 'didn', "didn't", 'doesn', "doesn', "doesn',
esn't", 'hadn',\
                           "hadn't", 'hasn', "hasn't", 'haven', "haven't", 'isn', "isn't", 'ma', 'mightn',
"mightn't", 'mustn',\
                           "mustn't", 'needn', "needn't", 'shan', "shan't", 'shouldn', "shouldn't", 'wasn',
"wasn't", 'weren', "weren't", \
                           'won', "won't", 'wouldn', "wouldn't"]
4
```

In [21]:

```
# Combining all the above stundents
from tqdm import tqdm
preprocessed_essays = []
# tqdm is for printing the status bar
for sentance in tqdm(project data[lessay]) values):
```

```
sent = decontracted(sentance)
sent = sent.replace('\\r', ' ')
sent = sent.replace('\\"', ' ')
sent = sent.replace('\\"', ' ')
sent = sent.replace('\\", ' ')
sent = re.sub('[^A-Za-z0-9]+', ' ', sent)
# https://gist.github.com/sebleier/554280
sent = ' '.join(e for e in sent.split() if e.lower() not in stopwords)
preprocessed_essays.append(sent.lower().strip())
100%| 100%| 109248/109248 [01:02<00:00, 1742.51it/s]
```

```
In [22]:
```

```
project_data['preprocessed_essays'] = preprocessed_essays
project_data.head(2)
```

Out[22]:

	Unnamed: 0	id	teacher_id	teacher_prefix	school_state	Date	project_title	projec
55660	8393	p205479	2bf07ba08945e5d8b2a3f269b2b3cfe5	Mrs.	CA	2016- 04-27 00:27:36	Engineering STEAM into the Primary Classroom	I have fortuna to use
76127	37728	p043609	3f60494c61921b3b43ab61bdde2904df	Ms.	UT	2016- 04-27 00:31:25	Sensory Tools for Focus	Imagin 9 year You're th

1.4 Preprocessing of project_title

In [0]:

```
# https://stackoverflow.com/a/47091490/4084039
import re

def decontracted(phrase):
    # specific
    phrase = re.sub(r"won't", "will not", phrase)
    phrase = re.sub(r"can\'t", "can not", phrase)

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

In [24]:

```
sent = decontracted(project_data['project_title'].values[2000])
print(sent)
print("="*50)
```

Empowering Students through Art in the Makerspace

```
In [25]:
```

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

Empowering Students through Art in the Makerspace

In [26]:

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

Empowering Students through Art in the Makerspace

In [0]:

```
# https://gist.github.com/sebleier/554280
# we are removing the words from the stop words list: 'no', 'nor', 'not'
stopwords= ['i', 'me', 'my', 'myself', 'we', 'our', 'ours', 'ourselves', 'you', "you're", "you've",
            "you'll", "you'd", 'your', 'yours', 'yourself', 'yourselves', 'he', 'him', 'his',
'himself', \
            'she', "she's", 'her', 'hers', 'herself', 'it', "it's", 'its', 'itself', 'they', 'them',
'their',\
            'theirs', 'themselves', 'what', 'which', 'who', 'whom', 'this', 'that', "that'll",
'these', 'those', \
            'am', 'is', 'are', 'was', 'were', 'be', 'been', 'being', 'have', 'has', 'had', 'having',
'do', 'does', \
            'did', 'doing', 'a', 'an', 'the', 'and', 'but', 'if', 'or', 'because', 'as', 'until', '
while', 'of', \
            'at', 'by', 'for', 'with', 'about', 'against', 'between', 'into', 'through', 'during',
'before', 'after',\
            'above', 'below', 'to', 'from', 'up', 'down', 'in', 'out', 'on', 'off', 'over', 'under'
, 'again', 'further',\
            'then', 'once', 'here', 'there', 'when', 'why', 'how', 'all', 'any', 'both', '&
ach', 'few', 'more',\
            'most', 'other', 'some', 'such', 'only', 'own', 'same', 'so', 'than', 'too', 'very', \
            's', 't', 'can', 'will', 'just', 'don', "don't", 'should', "should've", 'now', 'd', 'll'
, 'm', 'o', 're', \
           've', 'y', 'ain', 'aren', "aren't", 'couldn', "couldn't", 'didn', "didn't", 'doesn', "do
esn't", 'hadn',\
            "hadn't", 'hasn', "hasn't", 'haven', "haven't", 'isn', "isn't", 'ma', 'mightn',
"mightn't", 'mustn',\
            "mustn't", 'needn', "needn't", 'shan', "shan't", 'shouldn', "shouldn't", 'wasn',
"wasn't", 'weren', "weren't", \
            'won', "won't", 'wouldn', "wouldn't"]
4
```

In [28]:

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

```
In [29]:

project_data['preprocessed_titles'] = preprocessed_titles

project_data.head(2)
```

Out[29]:

	Unnamed: 0	id	teacher_id	teacher_prefix	school_state	Date	project_title	projec
55660	8393	p205479	2bf07ba08945e5d8b2a3f269b2b3cfe5	Mrs.	CA	2016- 04-27 00:27:36	Engineering STEAM into the Primary Classroom	I have fortuna to use
76127	37728	p043609	3f60494c61921b3b43ab61bdde2904df	Ms.	UT	2016- 04-27 00:31:25	Sensory Tools for Focus	Imagin 9 year: You're th

Join train & Resource dataset

```
In [30]:
```

```
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.head(2)
```

Out[30]:

	Unnamed: 0	id	teacher_id	teacher_prefix	school_state	Date	project_title	project_es
0	8393	p205479	2bf07ba08945e5d8b2a3f269b2b3cfe5	Mrs.	CA	2016- 04-27 00:27:36	Engineering STEAM into the Primary Classroom	I have been fortunate er to use the F
1	37728	p043609	3f60494c61921b3b43ab61bdde2904df	Ms.	UT	2016- 04-27 00:31:25	Sensory Tools for Focus	Imagine bei 9 years old. You're in yc th

Train Test split

```
In [31]:
```

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

Out[31]:

```
Hnnamed:
```

```
id
                                                teacher id teacher prefix school state
                                                                                             Date project title project ess
  Unnamed
                    id
                                                teacher_id teacher_prefix
                                                                           school_state
                                                                                             Date project_title project_ess
           n
                                                                                                  Engineering
                                                                                                               I have been
                                                                                         2016-
                                                                                                  STEAM into
                                                                                                               fortunate en
0 8393
             p205479 | 2bf07ba08945e5d8b2a3f269b2b3cfe5 | Mrs.
                                                                           CA
                                                                                         04-27
                                                                                                  the Primary
                                                                                                               to use the F
                                                                                         00:27:36
                                                                                                   Classroom
```

In [0]:

```
# train test split
from sklearn.model_selection import train_test_split
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.33, stratify=y)
```

F

In [0]:

In [34]:

```
# Combining all the above stundents
from tqdm import tqdm
train_preprocessed_essays = []
# tqdm is for printing the status bar
for sentance in tqdm(X_train['essay'].values):
    sent = decontracted(sentance)
    sent = sent.replace('\\r', '')
    sent = sent.replace('\\"', '')
    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)
    train_preprocessed_essays.append(sent.lower().strip())
```

In [35]:

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

In [36]:

```
# Combining all the above stundents
from tqdm import tqdm
cv_preprocessed_essays = []
# tqdm is for printing the status bar
for sentance in tqdm(X_cv['essay'].values):
    sent = decontracted(sentance)
    sent = sent.replace('\\r', ' ')
    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)
cv_preprocessed_essays.append(sent.lower().strip())

100%| 24155/24155 [00:13<00:00, 1749.53it/s]</pre>
```

In [37]:

```
# after preprocesing
test_preprocessed_essays[20000]
```

Out[37]:

'90 students come low income high risk backgrounds often unable receive supplies technology needed engage 21st century learning excited learn math digitally want provide better opportunity many not access basic technology home available classroom supplement missing homethese low cost highly efficient tablets offer small groups students 46 total opportunity engage technology much would like without battling campus computer lab campus one computer lab shared teachers sometimes becomes challenge use technology small group intervention students donation help continue make math funking teach not using pencil paper also digitally using date state art technology grow love learning names.

In [38]:

```
# Combining all the above stundents
from tqdm import tqdm
train_preprocessed_titles = []
# tqdm is for printing the status bar
for sentance in tqdm(X_train['project_title'].values):
    sent = decontracted(sentance)
    sent = sent.replace('\\r', ' ')
    sent = sent.replace('\\"', ' ')
    sent = sent.replace('\\"', ' ')
    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)
    train_preprocessed_titles.append(sent.lower().strip())
```

In [39]:

```
# Combining all the above stundents
from tqdm import tqdm
test_preprocessed_titles = []
# tqdm is for printing the status bar
for sentance in tqdm(X_test['project_title'].values):
    sent = decontracted(sentance)
    sent = sent.replace('\\r', ' ')
    sent = sent.replace('\\"', ' ')
    sent = sent.replace('\\"', ' ')
    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)
    test_preprocessed_titles.append(sent.lower().strip())
```

In [40]:

```
# Combining all the above stundents
from tqdm import tqdm
cv_preprocessed_titles = []
# tqdm is for printing the status bar
for sentance in tqdm(X_cv['project_title'].values):
    sent = decontracted(sentance)
    sent = sent.replace('\\r', '')
    sent = sent.replace('\\r', '')
    sent = sent.replace('\\r', '')
    sent = re.sub('[^A-Za-z0-9]+', '', sent)
# https://gist.github.com/sebleier/554280
    sent = '' isin'(a for a in cent colity); if a lever() not in stepwards)
```

```
sent = \cdot . Join(e for e in sent.spiit() if e.iower() not in stopwords)
    cv preprocessed titles.append(sent.lower().strip())
100%| 24155/24155 [00:00<00:00, 41055.74it/s]
In [41]:
preprocessed titles[2000]
Out[41]:
'empowering students art makerspace'
1.5 Preparing data for models
In [0]:
project_data.columns
Out[0]:
Index(['Unnamed: 0', 'id', 'teacher_id', 'teacher_prefix', 'school_state',
       'Date', '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',
       'clean categories', 'clean subcategories', 'grade cat list', 'essay',
       'preprocessed essays', 'preprocessed titles', 'price', 'quantity'],
      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
      - project title : text data
      - text : text data
      - project_resource_summary: text data (optinal)
      - quantity : numerical (optinal)
      - teacher_number_of_previously_posted_projects : numerical
      - price : numerical
1.5.1 Vectorizing Categorical data

    https://www.appliedaicourse.com/course/applied-ai-course-online/lessons/handling-categorical-and-numerical-features/

In [42]:
# we use count vectorizer to convert the values into one
from sklearn.feature_extraction.text import CountVectorizer
vectorizer = CountVectorizer(vocabulary=list(sorted cat dict.keys()), lowercase=False, binary=True
```

```
# we use count vectorizer to convert the values into one
from sklearn.feature_extraction.text import CountVectorizer
vectorizer = CountVectorizer(vocabulary=list(sorted_cat_dict.keys()), lowercase=False, binary=True
)
categories_one_hot = vectorizer.fit_transform(project_data['clean_categories'].values)
print(vectorizer.get_feature_names())
print("Shape of matrix after one hot encodig ",categories_one_hot.shape)

['Warmth', 'Care_Hunger', 'History_Civics', 'Music_Arts', 'AppliedLearning', 'SpecialNeeds',
'Health_Sports', 'Math_Science', 'Literacy_Language']
```

```
In [43]:
```

Shape of matrix after one hot encodig (109248, 9)

```
# we use count vectorizer to convert the values into one
vectorizer = CountVectorizer(vocabulary=list(sorted_sub_cat_dict.keys()), lowercase=False, binary=
True)
sub categories one hot = vectorizer.fit transform(project data['clean subcategories'].values)
print(vectorizer.get feature names())
print("Shape of matrix after one hot encodig ", sub categories one hot.shape)
['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 encodig (109248, 30)
In [0]:
\# you can do the similar thing with state, teacher_prefix and project_grade_category also
In [0]:
# count of all the words in corpus python: https://stackoverflow.com/a/22898595/4084039
my counter = Counter()
for word in project data['school state'].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]))
In [45]:
# we use count vectorizer to convert the values into one
from sklearn.feature extraction.text import CountVectorizer
vectorizer = CountVectorizer(vocabulary=list(sorted sub cat dict.keys()), lowercase=False, binary=
True)
state one hot = vectorizer.fit transform(project data['school state'].values)
print(vectorizer.get feature names())
print ("Shape of matrix after one hot encodig ", state one hot.shape)
['VT', 'WY', 'ND', 'MT', 'RI', 'SD', 'NE', 'DE', 'AK', 'NH', 'WV', 'ME', 'HI', 'DC', 'NM', 'KS', 'I
A', '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 encodig (109248, 51)
In [0]:
# count of all the words in corpus python: https://stackoverflow.com/a/22898595/4084039
project data['teacher prefix'] = project data['teacher prefix'].fillna(" ")
my_counter = Counter()
for word in project_data['teacher_prefix'].values.astype('str'):
#https://stackoverflow.com/questions/39116088/typeerror-in-countvectorizer-scikit-learn-expected-s
tring-or-buffer
   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]))
In [47]:
# we use count vectorizer to convert the values into one
from sklearn.feature extraction.text import CountVectorizer
vectorizer = CountVectorizer(vocabulary=list(sorted sub cat dict.keys()), lowercase=False, binary=
prefix_one_hot = vectorizer.fit_transform(project_data['teacher_prefix'].values.astype('str'))
#https://stackoverflow.com/questions/39116088/typeerror-in-countvectorizer-scikit-learn-expected-s
tring-or-buffer
print(vectorizer.get_feature_names())
print("Shape of matrix after one hot encodig ",prefix one hot.shape)
            1 1 126 1 126 1 126
```

```
['Dr.', 'Teacher', 'Mr.', 'Ms.', 'Mrs.']
Shape of matrix after one hot encodig (109248, 5)
In [0]:
# count of all the words in corpus python: https://stackoverflow.com/a/22898595/4084039
my counter = Counter()
for word in project data['grade cat list'].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]))
In [49]:
# we use count vectorizer to convert the values into one
from sklearn.feature extraction.text import CountVectorizer
vectorizer = CountVectorizer(vocabulary=list(sorted_sub_cat_dict.keys()), lowercase=False, binary=
True)
grade one hot = vectorizer.fit transform(project data['grade cat list'].values)
print(vectorizer.get_feature_names())
print("Shape of matrix after one hot encodig ",grade_one_hot.shape)
['9-12', '6-8', '3-5', 'PreK-2']
Shape of matrix after one hot encodig (109248, 4)
1.5.2 Vectorizing Text data
1.5.2.1 Bag of words
In [50]:
# We are considering only the words which appeared in at least 10 documents(rows or projects).
vectorizer = CountVectorizer(min df=10)
text bow = vectorizer.fit transform(preprocessed essays)
print("Shape of matrix after one hot encodig ",text bow.shape)
Shape of matrix after one hot encodig (109248, 16512)
In [51]:
# We are considering only the words which appeared in at least 10 documents (rows or projects).
vectorizer = CountVectorizer(min df=10)
text_bow = vectorizer.fit_transform(preprocessed_titles)
print("Shape of matrix after one hot encodig ",text bow.shape)
Shape of matrix after one hot encodig (109248, 3222)
1.5.2.2 TFIDF vectorizer
In [52]:
from sklearn.feature extraction.text import TfidfVectorizer
vectorizer = TfidfVectorizer(min df=10)
text_tfidf = vectorizer.fit_transform(preprocessed_essays)
print("Shape of matrix after one hot encodig ",text tfidf.shape)
Shape of matrix after one hot encodig (109248, 16512)
In [531:
from sklearn.feature extraction.text import TfidfVectorizer
vectorizer = TfidfVectorizer(min df=10)
text_tfidf = vectorizer.fit_transform(preprocessed_titles)
```

print("Shape of matrix after one hot encodig ",text_tfidf.shape)

1.5.2.3 Using Pretrained Models: Avg W2V

```
In [0]:
```

```
. . .
# Reading glove vectors in python: https://stackoverflow.com/a/38230349/4084039
def loadGloveModel(gloveFile):
   print ("Loading Glove Model")
   f = open(gloveFile,'r', encoding="utf8")
   model = {}
   for line in tqdm(f):
       splitLine = line.split()
       word = splitLine[0]
       embedding = np.array([float(val) for val in splitLine[1:]])
       model[word] = embedding
   print ("Done.",len(model)," words loaded!")
   return model
model = loadGloveModel('glove.42B.300d.txt')
Output:
Loading Glove Model
1917495it [06:32, 4879.69it/s]
Done. 1917495 words loaded!
# -----
words = []
for i in preproced texts:
   words.extend(i.split(' '))
for i in preproced titles:
   words.extend(i.split(' '))
print("all the words in the coupus", len(words))
words = set(words)
print("the unique words in the coupus", len(words))
inter words = set(model.keys()).intersection(words)
print("The number of words that are present in both glove vectors and our coupus", \
     len(inter_words),"(",np.round(len(inter_words)/len(words)*100,3),"%)")
words courpus = {}
words glove = set(model.keys())
for i in words:
   if i in words_glove:
       words courpus[i] = model[i]
print("word 2 vec length", len(words courpus))
# stronging variables into pickle files python: http://www.jessicayung.com/how-to-use-pickle-to-sa
ve-and-load-variables-in-python/
import pickle
with open('glove vectors', 'wb') as f:
   pickle.dump(words courpus, f)
```

Out[0]:

```
'\n# Reading glove vectors in python: https://stackoverflow.com/a/38230349/4084039\ndef
encoding="utf8")\n model = {}\n for line in tqdm(f):\n
                                                        splitLine = line.split() \n
word = splitLine[0]\n
                       embedding = np.array([float(val) for val in splitLine[1:]])\n
                       print ("Done.",len(model)," words loaded!")\n
odel[word] = embedding\n
                                                               return model\nmodel =
loadGloveModel(\'glove.42B.300d.txt\')\n\n# =============\nOutput:\n
                                                                         \nLoading G
love Model\n1917495it [06:32, 4879.69it/s]\nDone. 1917495 words loaded!\n\n#
======\n\nwords = []\nfor i in preproced_texts:\n
                                                               words.extend(i.split(\'
                               words.extend(i.split(\' \'))\nprint("all the words in the
\'))\n\nfor i in preproced titles:\n
countie" lan (words) \\nwords = set (words) \\nnrint ("the unique words in the countie"
```

```
coupus , Ten(words), \nwords = Set(words, \nprint( the unique words in the coupus ,
len(words))\n\ninter_words = Set(model.keys()).intersection(words)\nprint("The number of words tha
t are present in both glove vectors and our coupus", len(inter_words),"
(",np.round(len(inter_words)/len(words)*100,3),"%)")\n\nwords_courpus = {}\nwords_glove =
set(model.keys())\nfor i in words:\n if i in words_glove:\n words_courpus[i] = model[i]\r
print("word 2 vec length", len(words_courpus))\n\n\# stronging variables into pickle files python
: http://www.jessicayung.com/how-to-use-pickle-to-save-and-load-variables-in-python/\n\nimport pic
kle\nwith open(\'glove_vectors\', \'wb\') as f:\n pickle.dump(words_courpus, f)\n\n\"
```

In [0]:

```
# stronging variables into pickle files python: http://www.jessicayung.com/how-to-use-pickle-to-sa
ve-and-load-variables-in-python/
# make sure you have the glove_vectors file
with open('/content/drive/My Drive/Assignments_DonorsChoose_2018/glove_vectors', 'rb') as f:
    model = pickle.load(f)
    glove_words = set(model.keys())
```

In [55]:

```
# average Word2Vec
# compute average word2vec for each review.
avg w2v vectors = []; # the avg-w2v for each sentence/review is stored in this list
for sentence in tqdm(preprocessed essays): # for each review/sentence
   vector = np.zeros(300) # 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 vectors.append(vector)
print(len(avg_w2v_vectors))
print(len(avg w2v vectors[0]))
100%| 100%| 109248/109248 [00:35<00:00, 3115.74it/s]
```

109248 300

In [56]:

```
# average Word2Vec
# compute average word2vec for each review.
train avg w2v vectors = []; # the avg-w2v for each sentence/review is stored in this list
for sentence in tqdm(train preprocessed essays): # for each review/sentence
   vector = np.zeros(300) # 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
    train_avg_w2v_vectors.append(vector)
print(len(train avg w2v vectors))
print(len(train avg w2v vectors[0]))
100%| 49041/49041 [00:15<00:00, 3231.23it/s]
49041
```

In [57]:

300

```
# average Word2Vec
# compute average word2vec for each review.
```

In [58]:

300

```
# average Word2Vec
# compute average word2vec for each review.
cv avg w2v vectors = []; # the avg-w2v for each sentence/review is stored in this list
for sentence in tqdm(cv_preprocessed_essays): # for each review/sentence
   vector = np.zeros(300) # 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
    cv avg w2v vectors.append(vector)
print(len(cv avg w2v vectors))
print(len(cv avg w2v vectors[0]))
100%| 24155/24155 [00:07<00:00, 3300.53it/s]
```

24155 300

In [59]:

```
# average Word2Vec
# compute average word2vec for each review.
avg w2v vectors1 = []; # the avg-w2v for each sentence/review is stored in this list
for sentence in tqdm(preprocessed_titles): # for each review/sentence
   vector = np.zeros(300) # 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 vectors1.append(vector)
print(len(avg_w2v_vectors1))
print(len(avg w2v vectors1[0]))
100%| 100%| 109248/109248 [00:01<00:00, 64107.67it/s]
```

109248 300

```
# average Word2Vec
# compute average word2vec for each review.
train avg w2v vectors1 = []; # the avg-w2v for each sentence/review is stored in this list
for sentence in tqdm(train_preprocessed_titles): # for each review/sentence
    vector = np.zeros(300) # 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
    train avg w2v vectors1.append(vector)
print(len(train avg w2v vectors1))
print(len(train avg w2v vectors1[0]))
100%| 49041/49041 [00:00<00:00, 64956.11it/s]
```

49041 300

```
In [61]:
```

```
# average Word2Vec
# compute average word2vec for each review.
test_avg_w2v_vectors1 = []; # the avg-w2v for each sentence/review is stored in this list
for sentence in tqdm(test_preprocessed_titles): # for each review/sentence
    vector = np.zeros(300) # 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
    test_avg_w2v_vectors1.append(vector)
print(len(test_avg_w2v_vectors1))
print(len(test avg w2v vectors1[0]))
100%| 36052/36052 [00:00<00:00, 63787.04it/s]
36052
```

36052 300

In [62]:

```
# average Word2Vec
# compute average word2vec for each review.
cv avg w2v vectors1 = []; # the avg-w2v for each sentence/review is stored in this list
for sentence in tqdm(cv_preprocessed_titles): # for each review/sentence
   vector = np.zeros(300) # 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
    cv avg w2v vectors1.append(vector)
print(len(cv_avg_w2v_vectors1))
print(len(cv_avg_w2v_vectors1[0]))
100%| 24155/24155 [00:00<00:00, 63444.08it/s]
```

1.5.2.3 Using Pretrained Models: TFIDF weighted W2V

In [0]:

```
# S = ["abc def pqr", "def def def abc", "pqr pqr def"]
tfidf_model = TfidfVectorizer()
tfidf_model.fit(X_train['preprocessed_essays'])
# we are converting a dictionary with word as a key, and the idf as a value
dictionary = dict(zip(tfidf_model.get_feature_names(), list(tfidf_model.idf_)))
tfidf_words = set(tfidf_model.get_feature_names())
```

In [64]:

```
# average Word2Vec
# compute average word2vec for each review.
tfidf w2v vectors = []; # the avg-w2v for each sentence/review is stored in this list
for sentence in tqdm(preprocessed essays): # for each review/sentence
    vector = np.zeros(300) # as word vectors are of zero length
    tf_idf_weight =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) and (word in tfidf words):
            vec = model[word] # getting the vector for each word
            # here we are multiplying idf value(dictionary[word]) and the tf
value((sentence.count(word)/len(sentence.split())))
            tf_idf = dictionary[word]*(sentence.count(word)/len(sentence.split())) # getting the tf
idf value for each word
            vector += (vec * tf idf) # calculating tfidf weighted w2v
            tf idf weight += tf idf
    if tf idf weight != 0:
        vector /= tf idf weight
    tfidf w2v vectors.append(vector)
print(len(tfidf w2v vectors))
print(len(tfidf w2v vectors[0]))
100%| 100%| 1009248/109248 [03:22<00:00, 538.44it/s]
```

109248 300

In [65]:

```
# average Word2Vec
# compute average word2vec for each review.
train tfidf w2v vectors = []; # the avg-w2v for each sentence/review is stored in this list
\textbf{for} \ \texttt{sentence} \ \textbf{in} \ \texttt{tqdm} (\texttt{train\_preprocessed\_essays}) : \ \textit{\#} \ \textit{for each review/sentence}
    vector = np.zeros(300) # as word vectors are of zero length
    tf idf weight =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) and (word in tfidf words):
            vec = model[word] # getting the vector for each word
             # here we are multiplying idf value(dictionary[word]) and the tf
value((sentence.count(word)/len(sentence.split())))
            tf idf = dictionary[word]*(sentence.count(word)/len(sentence.split())) # getting the tf
idf value for each word
            vector += (vec * tf idf) # calculating tfidf weighted w2v
            tf idf weight += tf idf
    if tf idf weight != 0:
        vector /= tf idf weight
    train_tfidf_w2v_vectors.append(vector)
print(len(train tfidf w2v vectors))
print(len(train tfidf w2v vectors[0]))
100%| 49041/49041 [01:30<00:00, 542.01it/s]
```

```
In [66]:
```

```
# average Word2Vec
# compute average word2vec for each review.
test tfidf w2v vectors = []; # the avg-w2v for each sentence/review is stored in this list
for sentence in tqdm(test preprocessed essays): # for each review/sentence
   vector = np.zeros(300) # as word vectors are of zero length
    tf idf weight =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) and (word in tfidf words):
           vec = model[word] # getting the vector for each word
            # here we are multiplying idf value(dictionary[word]) and the tf
value((sentence.count(word)/len(sentence.split())))
           tf_idf = dictionary[word]*(sentence.count(word)/len(sentence.split())) # getting the tf
idf value for each word
           vector += (vec * tf idf) # calculating tfidf weighted w2v
           tf_idf_weight += tf_idf
    if tf idf weight != 0:
       vector /= tf idf weight
    test tfidf w2v vectors.append(vector)
print(len(test tfidf w2v vectors))
print(len(test tfidf w2v vectors[0]))
100%| 36052/36052 [01:06<00:00, 543.67it/s]
36052
```

In [67]:

300

```
# average Word2Vec
# compute average word2vec for each review.
cv tfidf w2v vectors = []; # the avg-w2v for each sentence/review is stored in this list
for sentence in tqdm(cv_preprocessed_essays): # for each review/sentence
   vector = np.zeros(300) # as word vectors are of zero length
    tf idf weight =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) and (word in tfidf_words):
           vec = model[word] # getting the vector for each word
            # here we are multiplying idf value(dictionary[word]) and the tf
value((sentence.count(word)/len(sentence.split())))
           tf idf = dictionary[word]*(sentence.count(word)/len(sentence.split())) # getting the tf
idf value for each word
           vector += (vec * tf idf) # calculating tfidf weighted w2v
            tf idf weight += tf idf
    if tf idf weight != 0:
       vector /= tf idf weight
    cv tfidf w2v vectors.append(vector)
print(len(cv tfidf w2v vectors))
print(len(cv_tfidf_w2v_vectors[0]))
100%| 24155/24155 [00:44<00:00, 543.04it/s]
24155
```

In [0]

300

```
# Similarly you can vectorize for title also
```

In [0]:

```
# Similarly you can vectorize for title also
tfidf_model2 = TfidfVectorizer()
tfidf_model2.fit(X_train['preprocessed_titles'])
# we are converting a dictionary with word as a key, and the idf as a value
dictionary = dict(zip(tfidf_model2.get_feature_names(), list(tfidf_model2.idf_)))
```

```
triar_words = set(triar_model2.get_reature_names())
```

In [70]:

```
# average Word2Vec
# compute average word2vec for each review.
tfidf w2v vectors1 = []; # the avg-w2v for each sentence/review is stored in this list
for sentence in tqdm(preprocessed titles): # for each review/sentence
   vector = np.zeros(300) # as word vectors are of zero length
    tf idf weight =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) and (word in tfidf words):
            vec = model[word] # getting the vector for each word
            # here we are multiplying idf value(dictionary[word]) and the tf
value((sentence.count(word)/len(sentence.split())))
            tf idf = dictionary[word]*(sentence.count(word)/len(sentence.split())) # getting the tf
idf value for each word
            vector += (vec * tf idf) # calculating tfidf weighted w2v
            tf idf weight += tf idf
    if tf_idf_weight != 0:
       vector /= tf idf weight
    tfidf_w2v_vectors1.append(vector)
print(len(tfidf w2v vectors1))
print(len(tfidf w2v vectors1[0]))
100%| 100%| 109248/109248 [00:04<00:00, 22435.28it/s]
109248
```

In [71]:

300

```
# average Word2Vec
# compute average word2vec for each review.
train tfidf w2v vectors1 = []; # the avg-w2v for each sentence/review is stored in this list
for sentence in tqdm(train preprocessed titles): # for each review/sentence
    vector = np.zeros(300) # as word vectors are of zero length
    tf idf weight =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) and (word in tfidf words):
           vec = model[word] # getting the vector for each word
            # here we are multiplying idf value(dictionary[word]) and the tf
value((sentence.count(word)/len(sentence.split())))
           tf idf = dictionary[word]*(sentence.count(word)/len(sentence.split())) # getting the tf
idf value for each word
           vector += (vec * tf idf) # calculating tfidf weighted w2v
           tf idf weight += tf idf
    if tf idf weight != 0:
        vector /= tf idf weight
    train tfidf w2v vectors1.append(vector)
print(len(train tfidf w2v vectors1))
print(len(train tfidf w2v vectors1[0]))
100%| 49041/49041 [00:02<00:00, 22579.90it/s]
```

49041 300

In [72]:

```
# average Word2Vec
# compute average word2vec for each review.
test_tfidf_w2v_vectors1 = []; # the avg-w2v for each sentence/review is stored in this list
for sentence in tqdm(test_preprocessed_titles): # for each review/sentence
    vector = np.zeros(300) # as word vectors are of zero length
    tf_idf_weight =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) and (word in tfidf_words):
            vec = model[word] # getting the vector for each word
```

36052 300

```
In [73]:
```

```
# average Word2Vec
# compute average word2vec for each review.
cv_tfidf_w2v_vectors1 = []; # the avg-w2v for each sentence/review is stored in this list
for sentence in tqdm(cv preprocessed titles): # for each review/sentence
   vector = np.zeros(300) # as word vectors are of zero length
    tf_idf_weight =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) and (word in tfidf words):
           vec = model[word] # getting the vector for each word
            # here we are multiplying idf value(dictionary[word]) and the tf
value((sentence.count(word)/len(sentence.split())))
            tf idf = dictionary[word]*(sentence.count(word)/len(sentence.split())) # getting the tf
idf value for each word
            vector += (vec * tf idf) # calculating tfidf weighted w2v
            tf idf weight += tf idf
    if tf idf weight != 0:
       vector /= tf_idf_weight
    cv tfidf w2v vectors1.append(vector)
print(len(cv tfidf w2v vectors1))
print(len(cv tfidf w2v vectors1[0]))
        24155/24155 [00:00<00:00, 24197.29it/s]
100%|
24155
300
```

1.5.3 Vectorizing Numerical features

In [74]:

```
# check this one: https://www.youtube.com/watch?v=0HOqOcln3Z4&t=530s
# standardization sklearn: https://scikit-
learn.org/stable/modules/generated/sklearn.preprocessing.StandardScaler.html
from sklearn.preprocessing import StandardScaler
# price_standardized = standardScalar.fit(project_data['price'].values)
# this will rise the error
# ValueError: Expected 2D array, got 1D array instead: array=[725.05 213.03 329. ... 399.
                                                                                              287.
73 5.5 ].
# Reshape your data either using array.reshape(-1, 1)
price scalar = StandardScaler()
price scalar.fit(project data['price'].values.reshape(-1,1)) # finding the mean and standard
deviation of this data
print(f"Mean : {price_scalar.mean_[0]}, Standard deviation : {np.sqrt(price scalar.var [0])}")
# Now standardize the data with above maen and variance.
price standardized = price scalar.transform(project data['price'].values.reshape(-1, 1))
tr price standardized = price_scalar.transform(X_train['price'].values.reshape(-1, 1))
```

```
te price standardized = price scalar.transform(X test['price'].values.reshape(-1, 1))
cv_price_standardized = price_scalar.transform(X_cv['price'].values.reshape(-1, 1))
Mean : 298.1193425966608, Standard deviation : 367.49634838483496
In [75]:
price standardized
Out[75]:
array([[ 1.16172762],
       [-0.23153793],
       [ 0.08402983],
       [ 0.27450792],
       [-0.0282706],
       [-0.79625102]]
In [76]:
# check this one: https://www.youtube.com/watch?v=0HOqOcln3Z4&t=530s
# standardization sklearn: https://scikit-
learn.org/stable/modules/generated/sklearn.preprocessing.StandardScaler.html
from sklearn.preprocessing import StandardScaler
# price standardized = standardScalar.fit(project data['price'].values)
# this will rise the error
# ValueError: Expected 2D array, got 1D array instead: array=[725.05 213.03 329. ... 399.
73 5.5 ].
# Reshape your data either using array.reshape(-1, 1)
quantity scalar = StandardScaler()
quantity_scalar.fit(project_data['quantity'].values.reshape(-1,1)) # finding the mean and standard
deviation of this data
print(f"Mean : {quantity_scalar.mean_[0]}, Standard deviation :
{np.sqrt(quantity_scalar.var_[0])}")
# Now standardize the data with above maen and variance.
quantity standardized = quantity scalar.transform(project data['quantity'].values.reshape(-1, 1))
tr_quantity_standardized = quantity_scalar.transform(X_train['quantity'].values.reshape(-1, 1))
te_quantity_standardized = quantity_scalar.transform(X_test['quantity'].values.reshape(-1, 1))
cv quantity standardized = quantity scalar.transform(X cv['quantity'].values.reshape(-1, 1))
Mean: 16.965610354422964, Standard deviation: 26.18282191909318
In [77]:
quantity standardized
Out[77]:
array([[-0.4951953],
       [-0.34242338],
       [-0.60977424],
       [-0.45700232],
       [-0.4951953],
       [ 0.3068572811)
```

Assignment 3: Apply KNN

- 1. [Task-1] Apply KNN(brute force version) on these feature sets
 - Set 1: categorical, numerical features + project_title(BOW) + preprocessed_essay (BOW)
 - Set 2: categorical, numerical features + project_title(TFIDF)+ preprocessed_essay (TFIDF)
 - Set 3: categorical, numerical features + project_title(AVG W2V)+ preprocessed_essay (AVG W2V)
 - Set 4: categorical, numerical features + project title(TFIDF W2V)+ preprocessed essay (TFIDF W2V)

2. Hyper paramter tuning to find best K

- Find the best hyper parameter which results in the maximum AUC value
- Find the best hyper paramter using k-fold cross validation (or) simple cross validation data
- Use gridsearch-cv or randomsearch-cv or write your own for loops to do this task

3. Representation of results

- You need to plot the performance of model both on train data and cross validation data for each hyper parameter, as shown in the figure
- Once you find the best hyper parameter, you need to train your model-M using the best hyper-param. Now, find the AUC on test data and plot the ROC curve on both train and test using model-M.
- Along with plotting ROC curve, you need to print the <u>confusion matrix</u> with predicted and original labels of test data points

4. [Task-2]

Select top 2000 features from feature Set 2 using <u>SelectKBest</u> and then apply KNN on top of these features

```
from sklearn.datasets import load_digits
from sklearn.feature_selection import SelectKBest, chi2
X, y = load_digits(return_X_y=True)
X.shape
X_new = SelectKBest(chi2, k=20).fit_transform(X, y)
X_new.shape
=======
output:
(1797, 64)
(1797, 20)
```

• Repeat the steps 2 and 3 on the data matrix after feature selection

5. Conclusion

• You need to summarize the results at the end of the notebook, summarize it in the table format. To print out a table please refer to this prettytable library link

Note: Data Leakage

- 1. There will be an issue of data-leakage if you vectorize the entire data and then split it into train/cv/test.
- 2. To avoid the issue of data-leakag, make sure to split your data first and then vectorize it.
- 3. While vectorizing your data, apply the method fit_transform() on you train data, and apply the method transform() on cv/test
- 4. For more details please go through this link.

2.2 Make Data Model Ready: encoding numerical, categorical features

Encoding - Categorical

```
In [78]:
```

```
print(X_train.shape, y_train.shape)
print(X_test.shape, y_test.shape)
print(X_cv.shape, y_cv.shape)

print("="*100)

vectorizer = CountVectorizer()
vectorizer.fit(X_train['clean_categories'].values) # fit has to happen only on train data

# we use the fitted CountVectorizer to convert the text to vector
X_train_cl_categories_ohe = vectorizer.transform(X_train['clean_categories'].values)
Y_test_cl_categories_ohe = vectorizer_transform(Y_test['clean_categories'].values)
```

```
A LEST CI CALEGOLIES ONE - VECTOLIZET. CLANSTOLM (A LEST CITEAU CALEGOLIES J. VALUES)
X_cv_cl_categories_ohe = vectorizer.transform(X_cv['clean_categories'].values)
print("After vectorizations")
print(X train cl categories ohe.shape, y train.shape)
print(X_test_cl_categories_ohe.shape, y_test.shape)
print (X cv cl categories ohe.shape, y cv.shape)
print("="*100)
# print("YOU SHOULD NOT DO SOMETHING LIKE THIS")
# vectorizer = CountVectorizer()
# x train bow = vectorizer.fit transform(X train['essay'].values)
# x cv bow = vectorizer.fit transform(X cv['essay'].values)
# x test bow = vectorizer.fit transform(X test['essay'].values)
# print(x_train_bow.shape, y_train.shape)
# print(x cv bow.shape, y cv.shape)
# print(x_test_bow.shape, y_test.shape)
print("NOTE: THE NUMBER OF COLUMNS IN EACH OF THE VECTOR WONT BE SAME")
(49041, 21) (49041,)
(36052, 21) (36052,)
(24155, 21) (24155,)
______
After vectorizations
(49041, 9) (49041,)
(36052, 9) (36052,)
(24155, 9) (24155,)
NOTE: THE NUMBER OF COLUMNS IN EACH OF THE VECTOR WONT BE SAME
In [79]:
print(X train.shape, y_train.shape)
print(X_test.shape, y_test.shape)
print(X_cv.shape, y_cv.shape)
print("="*100)
vectorizer = CountVectorizer()
vectorizer.fit(X train['clean subcategories'].values) # fit has to happen only on train data
# we use the fitted CountVectorizer to convert the text to vector
X train cl subcategories ohe = vectorizer.transform(X train['clean subcategories'].values)
X test cl subcategories ohe = vectorizer.transform(X test['clean subcategories'].values)
X_cv_cl_subcategories_ohe = vectorizer.transform(X_cv['clean_subcategories'].values)
print("After vectorizations")
print(X_train_cl_subcategories_ohe.shape, y_train.shape)
print(X test cl subcategories ohe.shape, y test.shape)
print(X cv cl subcategories ohe.shape, y cv.shape)
print("="*100)
# print("YOU SHOULD NOT DO SOMETHING LIKE THIS")
# vectorizer = CountVectorizer()
# x_train_bow = vectorizer.fit_transform(X_train['essay'].values)
# x cv bow = vectorizer.fit transform(X cv['essay'].values)
# x test bow = vectorizer.fit transform(X test['essay'].values)
# print(x train bow.shape, y train.shape)
# print(x cv bow.shape, y cv.shape)
# print(x_test_bow.shape, y_test.shape)
print("NOTE: THE NUMBER OF COLUMNS IN EACH OF THE VECTOR WONT BE SAME")
(49041, 21) (49041,)
(36052, 21) (36052,)
(24155, 21) (24155,)
______
```

```
After vectorizations
(49041, 30) (49041,)
(36052, 30) (36052,)
(24155, 30) (24155,)
NOTE: THE NUMBER OF COLUMNS IN EACH OF THE VECTOR WONT BE SAME
In [80]:
print(X_train.shape, y_train.shape)
print(X test.shape, y test.shape)
print(X cv.shape, y cv.shape)
print("="*100)
vectorizer = CountVectorizer()
vectorizer.fit(X_train['school_state'].values) # fit has to happen only on train data
\slash\hspace{-0.4em}\# we use the fitted CountVectorizer to convert the text to vector
X train school state ohe = vectorizer.transform(X train['school state'].values)
X test school state ohe = vectorizer.transform(X test['school state'].values)
X_cv_school_state_ohe = vectorizer.transform(X_cv['school_state'].values)
print("After vectorizations")
print(X_train_school_state_ohe.shape, y_train.shape)
print(X test school state ohe.shape, y test.shape)
print(X cv school state ohe.shape, y cv.shape)
print("="*100)
# print("YOU SHOULD NOT DO SOMETHING LIKE THIS")
# vectorizer = CountVectorizer()
# x train bow = vectorizer.fit transform(X train['essay'].values)
# x cv bow = vectorizer.fit transform(X cv['essay'].values)
# x_test_bow = vectorizer.fit_transform(X_test['essay'].values)
# print(x train bow.shape, y train.shape)
# print(x cv bow.shape, y cv.shape)
# print(x test bow.shape, y test.shape)
print ("NOTE: THE NUMBER OF COLUMNS IN EACH OF THE VECTOR WONT BE SAME")
(49041, 21) (49041,)
(36052, 21) (36052,)
(24155, 21) (24155,)
After vectorizations
(49041, 51) (49041,)
(36052, 51) (36052,)
(24155, 51) (24155,)
NOTE: THE NUMBER OF COLUMNS IN EACH OF THE VECTOR WONT BE SAME
4
In [81]:
print(X train.shape, y train.shape)
print(X test.shape, y test.shape)
print(X cv.shape, y cv.shape)
print("="*100)
vectorizer = CountVectorizer()
vectorizer.fit(X train['teacher prefix'].values.astype('U')) # fit has to happen only on train
#https://stackoverflow.com/questions/39303912/tfidfvectorizer-in-scikit-learn-valueerror-np-nan-is
-an-invalid-document
# we use the fitted CountVectorizer to convert the text to vector
X train teacher prefix ohe = vectorizer.transform(X train['teacher prefix'].values.astype('U'))
X test teacher prefix ohe = vectorizer.transform(X test['teacher prefix'].values.astype('U'))
```

```
X cv teacher prefix ohe = vectorizer.transform(X cv['teacher prefix'].values.astype('U'))
print("After vectorizations")
print(X train teacher prefix ohe.shape, y train.shape)
print(X test teacher prefix ohe.shape, y test.shape)
print(X cv teacher prefix ohe.shape, y cv.shape)
print("="*100)
# print("YOU SHOULD NOT DO SOMETHING LIKE THIS")
 # vectorizer = CountVectorizer()
# x_train_bow = vectorizer.fit_transform(X_train['essay'].values)
# x_cv_bow = vectorizer.fit_transform(X_cv['essay'].values)
# x test bow = vectorizer.fit transform(X test['essay'].values)
# print(x_train_bow.shape, y_train.shape)
# print(x cv bow.shape, y cv.shape)
# print(x_test_bow.shape, y_test.shape)
print ("NOTE: THE NUMBER OF COLUMNS IN EACH OF THE VECTOR WONT BE SAME")
(49041, 21) (49041,)
(36052, 21) (36052,)
(24155, 21) (24155,)
After vectorizations
(49041, 6) (49041,)
(36052, 6) (36052,)
(24155, 6) (24155,)
NOTE: THE NUMBER OF COLUMNS IN EACH OF THE VECTOR WONT BE SAME
In [82]:
print(X train.shape, y train.shape)
print(X_test.shape, y_test.shape)
print(X_cv.shape, y_cv.shape)
print("="*100)
vectorizer = CountVectorizer(vocabulary=list(sorted sub cat dict.keys()), lowercase=False, binary=
vectorizer.fit(X train['grade cat list'].values) # fit has to happen only on train data
# we use the fitted CountVectorizer to convert the text to vector
X train grade ohe = vectorizer.transform(X train['grade cat list'].values)
X_test_grade_ohe = vectorizer.transform(X_test['grade_cat_list'].values)
X cv grade ohe = vectorizer.transform(X cv['grade cat list'].values)
print("After vectorizations")
print(X train grade ohe.shape, y train.shape)
print(X_test_grade_ohe.shape, y_test.shape)
print(X_cv_grade_ohe.shape, y_cv.shape)
print("="*100)
# print("YOU SHOULD NOT DO SOMETHING LIKE THIS")
# vectorizer = CountVectorizer()
# x train bow = vectorizer.fit transform(X train['essay'].values)
# x cv bow = vectorizer.fit transform(X cv['essay'].values)
# x_test_bow = vectorizer.fit_transform(X_test['essay'].values)
# print(x_train_bow.shape, y_train.shape)
# print(x cv bow.shape, y_cv.shape)
# print(x test bow.shape, y test.shape)
print ("NOTE: THE NUMBER OF COLUMNS IN EACH OF THE VECTOR WONT BE SAME")
(49041, 21) (49041,)
(36052, 21) (36052,)
(24155, 21) (24155,)
                 ______
```

```
After vectorizations
(49041, 4) (49041,)
(36052, 4) (36052,)
(24155, 4) (24155,)

NOTE: THE NUMBER OF COLUMNS IN EACH OF THE VECTOR WONT BE SAME
```

2.3 Make Data Model Ready: encoding eassay, and project_title

Text - BOW

```
In [83]:
print(X_train.shape, y_train.shape)
print(X_test.shape, y_test.shape)
print(X_cv.shape, y_cv.shape)
print("="*100)
vectorizer = CountVectorizer(ngram range=(2,2), min df=10, max features=5000)
vectorizer.fit(train preprocessed essays) # fit has to happen only on train data
# we use the fitted CountVectorizer to convert the text to vector
X train essay bow = vectorizer.transform(train preprocessed essays)
X test essay bow = vectorizer.transform(test preprocessed essays)
X cv essay bow = vectorizer.transform(cv preprocessed essays)
print("After vectorizations")
print(X_train_essay_bow.shape, y_train.shape)
print(X_test_essay_bow.shape, y_test.shape)
print(X_cv_essay_bow.shape, y_cv.shape)
print("="*100)
# print("YOU SHOULD NOT DO SOMETHING LIKE THIS")
# vectorizer = CountVectorizer()
# x train bow = vectorizer.fit transform(X train['essay'].values)
# x_cv_bow = vectorizer.fit_transform(X_cv['essay'].values)
# x test bow = vectorizer.fit transform(X test['essay'].values)
# print(x train bow.shape, y train.shape)
# print(x cv bow.shape, y cv.shape)
# print(x_test_bow.shape, y_test.shape)
print("NOTE: THE NUMBER OF COLUMNS IN EACH OF THE VECTOR WONT BE SAME")
(49041, 21) (49041,)
(36052, 21) (36052,)
(24155, 21) (24155,)
After vectorizations
(49041, 5000) (49041,)
(36052, 5000) (36052,)
(24155, 5000) (24155,)
NOTE: THE NUMBER OF COLUMNS IN EACH OF THE VECTOR WONT BE SAME
In [84]:
print(X train.shape, y train.shape)
print(X test.shape, y test.shape)
print(X_cv.shape, y_cv.shape)
print("="*100)
```

```
vectorizer = CountVectorizer(min df=10, max features=5000)
vectorizer.fit(train_preprocessed_titles) # fit has to happen only on train data
# we use the fitted CountVectorizer to convert the text to vector
X train title bow = vectorizer.transform(train preprocessed titles)
X test title bow = vectorizer.transform(test_preprocessed_titles)
X cv title bow = vectorizer.transform(cv preprocessed titles)
print("After vectorizations")
print(X train title bow.shape, y train.shape)
print(X test title bow.shape, y test.shape)
print(X cv title bow.shape, y cv.shape)
print("="*100)
# print("YOU SHOULD NOT DO SOMETHING LIKE THIS")
# vectorizer = CountVectorizer()
# x train bow = vectorizer.fit transform(X train['essay'].values)
# x_cv_bow = vectorizer.fit_transform(X_cv['essay'].values)
# x_test_bow = vectorizer.fit_transform(X_test['essay'].values)
# print(x_train_bow.shape, y_train.shape)
# print(x cv bow.shape, y cv.shape)
# print(x test bow.shape, y test.shape)
print("NOTE: THE NUMBER OF COLUMNS IN EACH OF THE VECTOR WONT BE SAME")
(49041, 21) (49041,)
(36052, 21) (36052,)
(24155, 21) (24155,)
After vectorizations
(49041, 2013) (49041,)
(36052, 2013) (36052,)
(24155, 2013) (24155,)
______
NOTE: THE NUMBER OF COLUMNS IN EACH OF THE VECTOR WONT BE SAME
```

Text - Tfldf

In [85]:

```
print(X train.shape, y_train.shape)
print(X test.shape, y test.shape)
print(X_cv.shape, y_cv.shape)
print("="*100)
vectorizer = TfidfVectorizer(ngram_range=(2,2), min_df=10, max_features=5000)
vectorizer.fit(train preprocessed essays) # fit has to happen only on train data
# we use the fitted CountVectorizer to convert the text to vector
X train essay tf = vectorizer.transform(train preprocessed essays)
X test essay tf = vectorizer.transform(test preprocessed essays)
X cv essay tf = vectorizer.transform(cv preprocessed essays)
print("After vectorizations")
print(X_train_essay_tf.shape, y_train.shape)
print(X test essay tf.shape, y test.shape)
print(X_cv_essay_tf.shape, y_cv.shape)
print("="*100)
# print("YOU SHOULD NOT DO SOMETHING LIKE THIS")
# vectorizer = CountVectorizer()
# x train bow = vectorizer.fit transform(X train['essay'].values)
# x_cv_bow = vectorizer.fit_transform(X_cv['essay'].values)
# x test bow = vectorizer.fit transform(X test['essay'].values)
# print(x train bow.shape, y train.shape)
```

```
# print(x cv bow.shape, y cv.shape)
# print(x_test_bow.shape, y_test.shape)
print("NOTE: THE NUMBER OF COLUMNS IN EACH OF THE VECTOR WONT BE SAME")
(49041, 21) (49041,)
(36052, 21) (36052,)
(24155, 21) (24155,)
After vectorizations
(49041, 5000) (49041,)
(36052, 5000) (36052,)
(24155, 5000) (24155,)
NOTE: THE NUMBER OF COLUMNS IN EACH OF THE VECTOR WONT BE SAME
In [86]:
print(X_train.shape, y_train.shape)
print(X test.shape, y test.shape)
print(X cv.shape, y cv.shape)
print("="*100)
vectorizer = TfidfVectorizer(min_df=10, max_features=5000)
vectorizer.fit(train_preprocessed_titles) # fit has to happen only on train data
# we use the fitted CountVectorizer to convert the text to vector
X train title tf = vectorizer.transform(train preprocessed titles)
X test title tf = vectorizer.transform(test preprocessed titles)
X cv title tf = vectorizer.transform(cv preprocessed titles)
print("After vectorizations")
print(X_train_title_tf.shape, y_train.shape)
print(X test title_tf.shape, y_test.shape)
print(X cv title tf.shape, y cv.shape)
print("="*100)
# print("YOU SHOULD NOT DO SOMETHING LIKE THIS")
# vectorizer = CountVectorizer()
# x_train_bow = vectorizer.fit_transform(X_train['essay'].values)
# x cv bow = vectorizer.fit transform(X cv['essay'].values)
# x_test_bow = vectorizer.fit_transform(X_test['essay'].values)
# print(x train bow.shape, y train.shape)
# print(x_cv_bow.shape, y_cv.shape)
# print(x_test_bow.shape, y_test.shape)
print("NOTE: THE NUMBER OF COLUMNS IN EACH OF THE VECTOR WONT BE SAME")
(49041, 21) (49041,)
(36052, 21) (36052,)
(24155, 21) (24155,)
After vectorizations
(49041, 2013) (49041,)
(36052, 2013) (36052,)
(24155, 2013) (24155,)
NOTE: THE NUMBER OF COLUMNS IN EACH OF THE VECTOR WONT BE SAME
```

Text - Avg Word 2 Vec

In [0]:

```
|#List to Numpy array
#for Essays
X train essay avgw2v = np.array(train avg w2v vectors)
X test_essay_avgw2v = np.array(test_avg_w2v_vectors)
X_cv_essay_avgw2v = np.array(cv_avg_w2v_vectors)
#similarly, we are doing it for titles
X_train_title_avgw2v = np.array(train_avg_w2v_vectors1)
X test title avgw2v = np.array(test_avg_w2v_vectors1)
X cv title avgw2v = np.array(cv avg w2v vectors1)
In [88]:
#For Essays - Avgw2v
print(X_train.shape, y_train.shape)
print(X_test.shape, y_test.shape)
print(X_cv.shape, y_cv.shape)
print("="*100)
print("After vectorizations")
print(X_train_essay_avgw2v.shape, y_train.shape)
print(X_test_essay_avgw2v.shape, y_test.shape)
print(X_cv_essay_avgw2v.shape, y_cv.shape)
print("="*100)
(49041, 21) (49041,)
(36052, 21) (36052,)
(24155, 21) (24155,)
After vectorizations
(49041, 300) (49041,)
(36052, 300) (36052,)
(24155, 300) (24155,)
In [89]:
#For Titles - Avgw2v
print(X_train.shape, y_train.shape)
print(X_test.shape, y_test.shape)
print(X_cv.shape, y_cv.shape)
print("="*100)
print("After vectorizations")
print(X train title avgw2v.shape, y train.shape)
print(X_test_title_avgw2v.shape, y_test.shape)
print(X cv title avgw2v.shape, y cv.shape)
print("="*100)
(49041, 21) (49041,)
(36052, 21) (36052,)
(24155, 21) (24155,)
After vectorizations
(49041, 300) (49041,)
(36052, 300) (36052,)
(24155, 300) (24155,)
```

Text - Tfldf Weighted W2vec

In [0]:

```
#nttps://stackoverilow.com/questions/21U13b/4/11st-object-nas-no-attribute-snape
#List to Numpy array
#for Essays
X train es tfidf w2v = np.array(train tfidf w2v vectors)
X_test_es_tfidf_w2v = np.array(test_tfidf_w2v_vectors)
X_cv_es_tfidf_w2v = np.array(cv_tfidf_w2v_vectors)
#similarly, we are doing it for titles
X train title tfidf w2v = np.array(train tfidf w2v vectors1)
X test title tfidf w2v = np.array(test tfidf w2v vectors1)
X cv title tfidf w2v = np.array(cv tfidf w2v vectors1)
In [91]:
#For Essays - TfIdf weighted W2vec
print(X_train.shape, y_train.shape)
print(X_test.shape, y_test.shape)
print(X_cv.shape, y_cv.shape)
print("="*100)
print("After vectorizations")
print(X_train_es_tfidf_w2v.shape, y_train.shape)
print(X test es tfidf w2v.shape, y test.shape)
print(X_cv_es_tfidf_w2v.shape, y_cv.shape)
print("="*100)
(49041, 21) (49041,)
(36052, 21) (36052,)
(24155, 21) (24155,)
After vectorizations
(49041, 300) (49041,)
(36052, 300) (36052,)
(24155, 300) (24155,)
______
                                                                                     ▶
In [92]:
#For Titles - TfIdf Weighted W2Vec
print(X_train.shape, y_train.shape)
print(X test.shape, y_test.shape)
print(X cv.shape, y cv.shape)
print("="*100)
print("After vectorizations")
print(X train title tfidf w2v.shape, y train.shape)
print(X test title tfidf w2v.shape, y test.shape)
print(X_cv_title_tfidf_w2v.shape, y_cv.shape)
print("="*100)
(49041, 21) (49041,)
(36052, 21) (36052,)
(24155, 21) (24155,)
After vectorizations
(49041, 300) (49041,)
(36052, 300) (36052,)
(24155, 300) (24155,)
```

Concatinating all the features

```
In [93]:
from scipy.sparse import hstack
# with the same hstack function we are concatinating a sparse matrix and a dense matirx
X Bow train = hstack((X train essay bow, X train title bow, tr price standardized,
tr_quantity_standardized, X_train_cl_categories_ohe,
                     X train cl subcategories ohe, X train school state ohe,
X train teacher prefix ohe, X train grade ohe)).tocsr()
print(X_Bow_train.shape, y_train.shape)
(49041, 7115) (49041,)
In [94]:
X Bow test = hstack((X test essay bow, X test title bow, te price standardized, te quantity standar
dized, X test cl categories ohe,
                     X_test_cl_subcategories_ohe, X_test_school_state ohe,
X test teacher prefix ohe, X test grade ohe)).tocsr()
print(X_Bow_test.shape, y_test.shape)
(36052, 7115) (36052,)
In [95]:
X Bow cv = hstack((X cv essay_bow, X_cv_title_bow, cv_price_standardized, cv_quantity_standardized,
X cv cl categories ohe,
                     X cv cl subcategories ohe, X cv school state ohe, X cv teacher prefix ohe, X c
v_grade_ohe)).tocsr()
print(X_Bow_cv.shape, y_cv.shape)
4
(24155, 7115) (24155,)
In [96]:
from scipy.sparse import hstack
# with the same hstack function we are concatinating a sparse matrix and a dense matirx
X_tf_train = hstack((X_train_essay_tf, X_train_title_tf, tr_price_standardized, tr_quantity_standar
dized, X_train_cl_categories_ohe,
                     X_train_cl_subcategories_ohe, X_train_school_state_ohe,
X_train_teacher_prefix_ohe, X_train_grade_ohe)).tocsr()
print(X tf train.shape, y train.shape)
(49041, 7115) (49041,)
In [97]:
X tf test = hstack((X test essay tf, X test title tf, te price standardized,
te quantity standardized, X test cl categories ohe,
                     X test cl subcategories ohe, X test school state ohe,
X test teacher prefix ohe, X test grade ohe)).tocsr()
print(X_tf_test.shape, y_test.shape)
(36052, 7115) (36052,)
In [98]:
X_tf_cv = hstack((X_cv_essay_tf, X_cv_title_tf, cv_price_standardized, cv_quantity_standardized, X_
cv_cl_categories_ohe,
                     X_cv_cl_subcategories_ohe, X_cv_school_state_ohe, X_cv_teacher_prefix_ohe, X_c
```



```
(24155, 7115) (24155,)
In [99]:
from scipy.sparse import hstack
# with the same hstack function we are concatinating a sparse matrix and a dense matirx
X_avg_w2v_train = hstack((X_train_essay_avgw2v, X_train_title_avgw2v, tr_price_standardized,
tr quantity standardized, X train cl categories ohe,
                    X train cl subcategories ohe, X train school state ohe,
X_train_teacher_prefix_ohe, X_train_grade_ohe)).tocsr()
print(X avg w2v train.shape, y train.shape)
(49041, 702) (49041,)
In [100]:
X_avg_w2v_test = hstack((X_test_essay_avgw2v, X_test_title_avgw2v, te_price_standardized, te_quanti
ty_standardized, X test cl categories ohe,
                     X test cl subcategories ohe, X test school state ohe,
X_test_teacher_prefix_ohe, X_test_grade_ohe)).tocsr()
print(X avg w2v test.shape, y test.shape)
(36052, 702) (36052,)
In [101]:
X avg w2v cv = hstack((X cv essay avgw2v, X cv title avgw2v, cv price standardized,
cv quantity standardized, X cv cl categories ohe,
                     X_cv_cl_subcategories_ohe, X_cv_school_state_ohe, X_cv_teacher_prefix_ohe, X_c
v grade ohe)).tocsr()
print(X avg w2v cv.shape, y cv.shape)
(24155, 702) (24155,)
In [102]:
from scipy.sparse import hstack
# with the same hstack function we are concatinating a sparse matrix and a dense matirx
X tf w2v train = hstack((X train es tfidf w2v, X train title tfidf w2v, tr price standardized,
tr_quantity_standardized, X_train_cl_categories_ohe,
                    X train cl subcategories ohe, X train school state ohe,
X_train_teacher_prefix_ohe, X_train_grade_ohe)).tocsr()
print(X tf w2v train.shape, y train.shape)
(49041, 702) (49041,)
In [103]:
X tf w2v test = hstack((X test es tfidf w2v, X test title tfidf w2v, te price standardized,
te quantity standardized, X test cl categories ohe,
                    X_test_cl_subcategories_ohe, X_test_school_state_ohe,
X_test_teacher_prefix_ohe, X_test grade ohe)).tocsr()
print(X_avg_w2v_test.shape, y_test.shape)
(36052, 702) (36052,)
In [104]:
X_tf_w2v_cv = hstack((X_cv_es_tfidf_w2v, X_cv_title_tfidf_w2v, cv_price_standardized,
cv_quantity_standardized, X_cv_cl_categories_ohe,
```

```
v_grade_ohe)).tocsr()

print(X_avg_w2v_cv.shape, y_cv.shape)

(24155, 702) (24155,)
```

2.4 Appling KNN on different kind of featurization as mentioned in the instructions

Apply KNN on different kind of featurization as mentioned in the instructions For Every model that you work on make sure you do the step 2 and step 3 of instructions

2.4.1 Applying KNN brute force on BOW, SET 1

```
In [0]:
```

```
from sklearn.neighbors import KNeighborsClassifier
from sklearn.metrics import roc auc score
import matplotlib.pyplot as plt
train_auc = []
cv_auc = []
K = [5, 15, 21, 31, 41, 51]
for i in K:
    neigh = KNeighborsClassifier(n neighbors=i)
    neigh.fit(X_Bow_train,y_train)
    y train pred = neigh.predict proba(X Bow train)[:,1]
    y_cv_pred = neigh.predict_proba(X_Bow_cv)[:,1]
    # roc_auc_score(y_true, y_score) the 2nd parameter should be probability estimates of the posi
tive class
    # not the predicted outputs
    train auc.append(roc_auc_score(y_train,y_train_pred))
    cv_auc.append(roc_auc_score(y_cv, y_cv_pred))
    print("Completed for k = {}".format(i))
plt.plot(K, train auc, label='Train AUC')
plt.plot(K, cv auc, label='CV AUC')
plt.scatter(K, train_auc, label='Train AUC points')
plt.scatter(K, cv auc, label='CV AUC points')
plt.legend()
plt.xlabel("K: hyperparameter")
plt.ylabel("AUC")
plt.title("ERROR PLOTS")
plt.xticks(K)
plt.grid()
plt.show()
```

```
Completed for k=5
Completed for k=15
Completed for k=21
Completed for k=31
Completed for k=41
Completed for k=51
```



```
0.55

0.50

5 15 21 31 41 51

K: hyperparameter
```

In [0]:

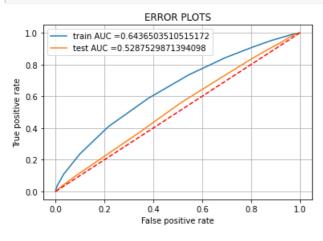
```
score_t_cv = [x for x in cv_auc]
opt_t_cv = K[score_t_cv.index(max(score_t_cv))]
print("Maximum AUC score of cv is:" + ' ' + str(max(score_t_cv)))
print("Corresponding k value of cv is:",opt_t_cv, '\n')
best_k=opt_t_cv
print(best_k)
```

Maximum AUC score of cv is: 0.5233624838859942 Corresponding k value of cv is: 51

51

In [0]:

```
# https://scikit-
learn.org/stable/modules/generated/sklearn.metrics.roc curve.html#sklearn.metrics.roc curve
from sklearn.metrics import roc curve, auc
from sklearn.neighbors import KNeighborsClassifier
from sklearn.metrics import roc auc score
import matplotlib.pyplot as plt
neigh = KNeighborsClassifier(n_neighbors=51)
neigh.fit(X_Bow_train, y_train)
# roc_auc_score(y_true, y_score) the 2nd parameter should be probability estimates of the positive
class
# not the predicted outputs
y_train_pred = neigh.predict_proba(X_Bow_train)[:,1]
y test pred = neigh.predict proba(X Bow test)[:,1]
train fpr, train tpr, tr thresholds = roc curve(y train, y train pred)
test fpr, test tpr, te thresholds = roc curve(y test, y test pred)
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.title("AUC plot")
plt.xlabel("False positive rate")
plt.ylabel("True positive rate")
plt.plot([0, 1], [0, 1], 'r--')
plt.title("ERROR PLOTS")
plt.grid()
plt.show()
```



```
y_train_pred_bow = neigh.predict(X_bow_train)

y_test_pred_Bow = neigh.predict(X_Bow_test)
```

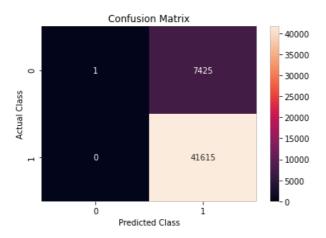
In [0]:

```
cm = confusion_matrix(y_train,y_train_pred_Bow)
print("Train confusion matrix")
# https://seaborn.pydata.org/generated/seaborn.heatmap.html
sns.heatmap(cm, annot=True, fmt="d")
plt.ylabel("Actual Class")
plt.xlabel("Predicted Class")
plt.title("Confusion Matrix")
```

Train confusion matrix

Out[0]:

Text(0.5, 1.0, 'Confusion Matrix')



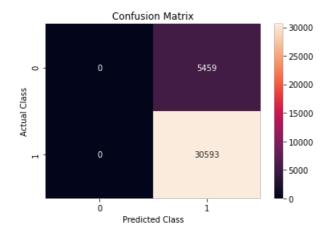
In [0]:

```
#y_test_pred = neigh.predict(set2_t)
cm = confusion_matrix(y_test,y_test_pred_Bow)
print("Test confusion matrix")
# https://seaborn.pydata.org/generated/seaborn.heatmap.html
sns.heatmap(cm, annot=True, fmt="d")
plt.ylabel("Actual Class")
plt.xlabel("Predicted Class")
plt.title("Confusion Matrix")
```

Test confusion matrix

Out[0]:

Text(0.5, 1.0, 'Confusion Matrix')

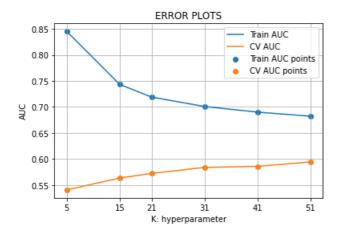


2.4.2 Applying KNN brute force on TFIDF, SET 2

```
In [0]:
```

```
from sklearn.neighbors import KNeighborsClassifier
from sklearn.metrics import roc auc score
import matplotlib.pyplot as plt
train_auc = []
cv_auc = []
K = [5, 15, 21, 31, 41, 51]
for i in K:
    neigh = KNeighborsClassifier(n neighbors=i)
   neigh.fit(X_tf_train,y_train)
    y train pred = neigh.predict proba(X tf train)[:,1]
    y cv pred = neigh.predict proba(X tf cv)[:,1]
    # roc_auc_score(y_true, y_score) the 2nd parameter should be probability estimates of the posi
tive class
    # not the predicted outputs
    train_auc.append(roc_auc_score(y_train,y_train_pred))
    cv auc.append(roc auc score(y cv, y cv pred))
    print("Completed for k = {}".format(i))
plt.plot(K, train_auc, label='Train AUC')
plt.plot(K, cv_auc, label='CV AUC')
plt.scatter(K, train_auc, label='Train AUC points')
plt.scatter(K, cv auc, label='CV AUC points')
plt.legend()
plt.xlabel("K: hyperparameter")
plt.ylabel("AUC")
plt.title("ERROR PLOTS")
plt.xticks(K)
plt.grid()
plt.show()
Completed for k = 5
Completed for k = 15
Completed for k = 21
```

Completed for k = 5Completed for k = 15Completed for k = 21Completed for k = 31Completed for k = 41Completed for k = 51



In [0]:

```
score_t_cv = [x for x in cv_auc]
opt_t_cv = K[score_t_cv.index(max(score_t_cv))]
print("Maximum AUC score of cv is:" + ' ' + str(max(score_t_cv)))
print("Corresponding k value of cv is:",opt_t_cv, '\n')
best_k=opt_t_cv
print(best_k)
```

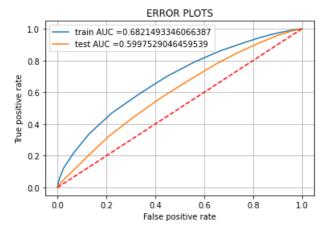
Maximum AUC score of cv is: 0.5942390305831075

```
Corresponding k value of cv is: 51
```

51

```
In [0]:
```

```
# https://scikit-
learn.org/stable/modules/generated/sklearn.metrics.roc curve.html#sklearn.metrics.roc curve
from sklearn.metrics import roc_curve, auc
neigh1 = KNeighborsClassifier(n neighbors=51)
neigh1.fit(X_tf_train, y_train)
# roc_auc_score(y_true, y_score) the 2nd parameter should be probability estimates of the positive
class
# not the predicted outputs
y train pred = neigh1.predict proba(X tf train)[:,1]
y test pred = neigh1.predict proba(X tf test)[:,1]
train_fpr, train_tpr, tr_thresholds = roc_curve(y_train, y_train_pred)
test fpr, test tpr, te thresholds = roc curve(y test, y test pred)
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.title("AUC plot")
plt.xlabel("False positive rate")
plt.ylabel("True positive rate")
plt.plot([0, 1], [0, 1], 'r--')
plt.title("ERROR PLOTS")
plt.grid()
plt.show()
```



Confusion Matrix

```
In [0]:
```

```
y_train_pred_tf = neigh1.predict(X_tf_train)
y_test_pred_tf = neigh1.predict(X_tf_test)
```

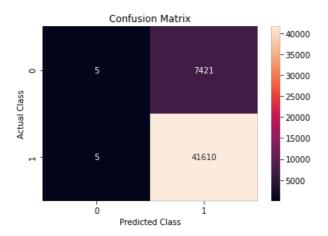
In [0]:

```
cm = confusion_matrix(y_train,y_train_pred_tf)
print("Train confusion matrix")
# https://seaborn.pydata.org/generated/seaborn.heatmap.html
sns.heatmap(cm, annot=True, fmt="d")
plt.ylabel("Actual Class")
plt.xlabel("Predicted Class")
plt.title("Confusion Matrix")
```

Train confusion matrix

Out[0]:

Text(0.5, 1.0, 'Confusion Matrix')



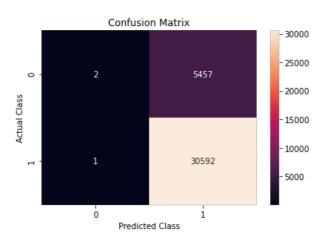
In [0]:

```
#y_test_pred = neigh.predict(set2_t)
cm = confusion_matrix(y_test,y_test_pred_tf)
print("Test_confusion matrix")
# https://seaborn.pydata.org/generated/seaborn.heatmap.html
sns.heatmap(cm, annot=True, fmt="d")
plt.ylabel("Actual Class")
plt.xlabel("Predicted Class")
plt.title("Confusion Matrix")
```

Test confusion matrix

Out[0]:

Text(0.5, 1.0, 'Confusion Matrix')



2.4.3 Applying KNN brute force on AVG W2V, SET 3

```
from sklearn.neighbors import KNeighborsClassifier
from sklearn.metrics import roc_auc_score
import matplotlib.pyplot as plt

train_auc = []
cv_auc = []
K = [5, 15, 21, 31, 41, 51]
for i in K:
    neigh = KNeighborsClassifier(n_neighbors=i)
    neigh.fit(X_avg_w2v_train,y_train)

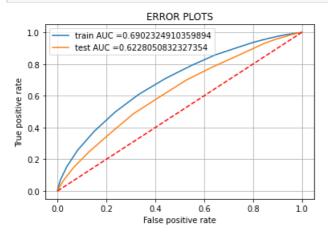
v train pred = neigh.predict proba(X avg w2v train)[:,1]
```

```
y_cv_pred = neigh.predict_proba(X_avg_w2v_cv)[:,1]

# roc_auc_score(y_true, y_score) the 2nd parameter should be probability estimates of the posi
tive class
# not the predicted outputs
train_auc.append(roc_auc_score(y_train,y_train_pred))
cv_auc.append(roc_auc_score(y_cv, y_cv_pred))
```

In [105]:

```
# https://scikit-
learn.org/stable/modules/generated/sklearn.metrics.roc\_curve.html \# sklearn.metrics.roc\_curve.html \# sklearn.metrics.html \# sklearn.html \# sklearn.metrics.html \# sklearn.html \# sklea
from sklearn.metrics import roc curve, auc
from sklearn.neighbors import KNeighborsClassifier
from sklearn.metrics import roc auc score
import matplotlib.pyplot as plt
neigh2 = KNeighborsClassifier(n neighbors=51)
neigh2.fit(X_avg_w2v_train, y_train)
 # roc_auc_score(y_true, y_score) the 2nd parameter should be probability estimates of the positive
class
 # not the predicted outputs
y train pred = neigh2.predict proba(X avg w2v train)[:,1]
y test pred = neigh2.predict proba(X avg w2v test)[:,1]
train fpr, train tpr, tr thresholds = roc curve (y train, y train pred)
test_fpr, test_tpr, te_thresholds = roc_curve(y_test, y_test_pred)
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.title("AUC plot")
plt.xlabel("False positive rate")
plt.ylabel("True positive rate")
plt.plot([0, 1], [0, 1], 'r--')
plt.title("ERROR PLOTS")
plt.grid()
plt.show()
```



In [0]:

#Confusion Matrix

In [0]:

```
y_train_pred_w2v = neigh2.predict(X_avg_w2v_train)
y_test_pred_w2v = neigh2.predict(X_avg_w2v_test)
```

In [107]:

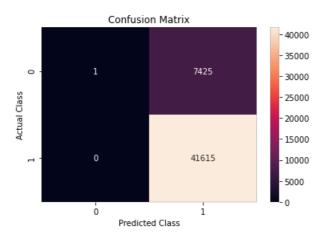
```
cm = confusion_matrix(y_train,y_train_pred_w2v)
print("Train confusion matrix")
# https://seaborn.pydata.org/generated/seaborn.heatmap.html
```

```
sns.heatmap(cm, annot=True, fmt="d")
plt.ylabel("Actual Class")
plt.xlabel("Predicted Class")
plt.title("Confusion Matrix")
```

Train confusion matrix

Out[107]:

Text(0.5, 1.0, 'Confusion Matrix')



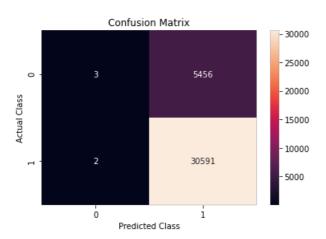
In [108]:

```
#y_test_pred = neigh.predict(set2_t)
cm = confusion_matrix(y_test,y_test_pred_w2v)
print("Test confusion matrix")
# https://seaborn.pydata.org/generated/seaborn.heatmap.html
sns.heatmap(cm, annot=True, fmt="d")
plt.ylabel("Actual Class")
plt.xlabel("Predicted Class")
plt.title("Confusion Matrix")
```

Test confusion matrix

Out[108]:

Text(0.5, 1.0, 'Confusion Matrix')



2.4.4 Applying KNN brute force on TFIDF W2V, SET 4

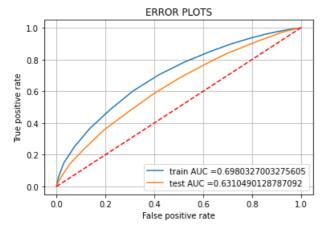
```
train_auc = []
cv_auc = []
K = [5, 15, 21, 31, 41, 51]
for i in K:
    neigh = KNeighborsClassifier(n_neighbors=i)
    neigh.fit(X_tf_w2v_train,y_train)

    y_train_pred = neigh.predict(X_tf_w2v_train)
    y_cv_pred = neigh.predict(X_tf_w2v_cv)

# roc_auc_score(y_true, y_score) the 2nd parameter should be probability estimates of the positive class
    # not the predicted outputs
    train_auc.append(roc_auc_score(y_train,y_train_pred))
    cv_auc.append(roc_auc_score(y_cv, y_cv_pred))
```

In [111]:

```
# https://scikit-
learn.org/stable/modules/generated/sklearn.metrics.roc\_curve.html \# sklearn.metrics.roc\_curve.html \# sklearn.metrics.html \# sklearn.html \# sklearn.metrics.html \# sklearn.html \# sklea
from sklearn.metrics import roc curve, auc
from sklearn.neighbors import KNeighborsClassifier
from sklearn.metrics import roc_auc_score
import matplotlib.pyplot as plt
neigh3 = KNeighborsClassifier(n neighbors=51)
neigh3.fit(X tf w2v train, y train)
# roc_auc_score(y_true, y_score) the 2nd parameter should be probability estimates of the positive
 # not the predicted outputs
y train pred = neigh3.predict proba(X tf w2v train)[:,1]
y_test_pred = neigh3.predict_proba(X_tf_w2v_test)[:,1]
train_fpr, train_tpr, tr_thresholds = roc_curve(y_train, y_train_pred)
test_fpr, test_tpr, te_thresholds = roc_curve(y_test, y_test_pred)
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.title("AUC plot")
plt.xlabel("False positive rate")
plt.ylabel("True positive rate")
plt.plot([0, 1], [0, 1], 'r--')
plt.title("ERROR PLOTS")
plt.grid()
plt.show()
```



In [0]:

```
#Confusion Matrix
```

```
y_train_pred_tf_w2v = neigh3.predict(X_tf_w2v_train)
```

```
y_test_pred_tf_w2v = neigh3.predict(X_tf_w2v_train)
```

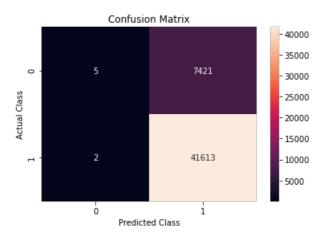
In [113]:

```
cm = confusion_matrix(y_train,y_train_pred_tf_w2v)
print("Train confusion matrix")
# https://seaborn.pydata.org/generated/seaborn.heatmap.html
sns.heatmap(cm, annot=True, fmt="d")
plt.ylabel("Actual Class")
plt.xlabel("Predicted Class")
plt.title("Confusion Matrix")
```

Train confusion matrix

Out[113]:

Text(0.5, 1.0, 'Confusion Matrix')



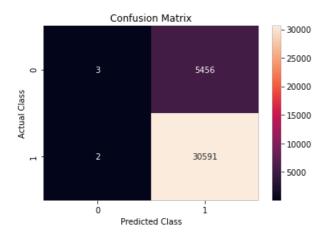
In [114]:

```
#y_test_pred = neigh.predict(set2_t)
cm = confusion_matrix(y_test,y_test_pred_w2v)
print("Test confusion matrix")
# https://seaborn.pydata.org/generated/seaborn.heatmap.html
sns.heatmap(cm, annot=True, fmt="d")
plt.ylabel("Actual Class")
plt.xlabel("Predicted Class")
plt.xlabel("Confusion Matrix")
```

Test confusion matrix

Out[114]:

Text(0.5, 1.0, 'Confusion Matrix')



2.5 Feature selection with `SelectKBest`

```
In [0]:
```

```
from sklearn.feature_selection import SelectKBest
sk = SelectKBest(k=2000).fit(X_tf_train, y_train)
X train new = sk.transform(X tf train)
X test new = sk.transform(X tf test)
X cv new = sk.transform(X tf cv)
```

In [0]:

```
print(X train new.shape)
print(X test new.shape)
print(X_cv_new.shape)
```

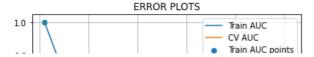
(49041, 2000) (36052, 2000) (24155, 2000)

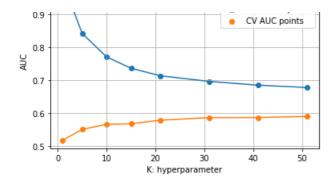
Hyper-Parameter Tunning

In [0]:

```
#train essay tfidf w2v vectors
 #test_essay_tfidf_w2v_vectors
train auc = []
cv auc = []
K = [1, 5, 10, 15, 21, 31, 41, 51]
for i in tqdm(K):
           neigh4 = KNeighborsClassifier(n neighbors=i)
          neigh4.fit(X train_new, y_train)
           \# roc_auc_score(y_true, y_score) the 2nd parameter should be probability estimates of the positive positive positive probability estimates of the positive probability estimates and the probability estimates of the positive probability estimates and the probability estimates are probabilities and the probabilities are probabilities are probabilities are probabilities and the probabilities are probabilities 
tive class
           # not the predicted outputs
          y_train_pred = neigh4.predict_proba(X_train_new)[:,1]#Return probability estimates for the
set3x , for the class label 1 or +ve.
           y cv pred = neigh4.predict proba(X cv new)[:,1] #Return probability estimates for the
set3cvx, for the class label 1 or +ve .
           # roc_auc_score(y_true, y_score) the 2nd parameter should be probability estimates of the posi
 tive class
           # not the predicted outputs
          train auc.append(roc auc score(y train, y train pred))
           cv_auc.append(roc_auc_score(y_cv, y_cv_pred))
plt.plot(K, train auc, label='Train AUC')
plt.plot(K, cv auc, label='CV AUC')
plt.scatter(K, train auc, label='Train AUC points')
plt.scatter(K, cv auc, label='CV AUC points')
plt.legend()
plt.xlabel("K: hyperparameter")
plt.ylabel("AUC")
plt.title("ERROR PLOTS")
plt.grid()
plt.show()
```

100%| 8/8 [22:40<00:00, 170.04s/it]





In [0]:

```
scl = [x for x in cv_auc]
opt_t_cv_4 = K[scl.index(max(scl))]
print("Maximum AUC score of cv is:" + ' ' + str(max(scl)))
print("Corresponding k value of cv is:",opt_t_cv_4, '\n')
best_k=opt_t_cv_4
print(best_k)
```

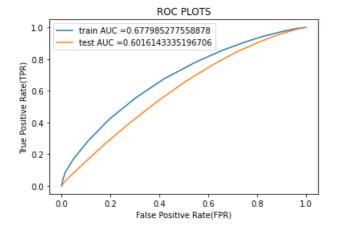
Maximum AUC score of cv is: 0.5901771751583547 Corresponding k value of cv is: 51

51

In [0]:

#ROC/AUC Curve

```
# https://scikit-
learn.org/stable/modules/generated/sklearn.metrics.roc curve.html#sklearn.metrics.roc curve
from sklearn.metrics import roc curve, auc
neigh4 = KNeighborsClassifier(n_neighbors=51)
neigh4.fit(X train new ,y train)
# roc auc score(y true, y score) the 2nd parameter should be probability estimates of the positive
class
# not the predicted outputs
train_fpr, train_tpr, thresholds = roc_curve(y_train, neigh4.predict_proba(X_train_new)[:,1])
test fpr, test tpr, thresholds = roc curve(y test, neigh4.predict proba(X test new)[:,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.ylabel("True Positive Rate(TPR)")
plt.xlabel("False Positive Rate(FPR)")
plt.title("ROC PLOTS")
plt.show()
```



Confusion Matrix

In [0]:

```
y_train_K = neigh4.predict(X_train_new)
y_test_K = neigh4.predict(X_test_new)
```

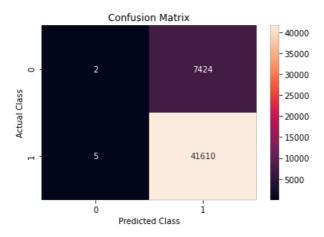
In [0]:

```
cm = confusion_matrix(y_train,y_train_K)
print("Train confusion matrix")
# https://seaborn.pydata.org/generated/seaborn.heatmap.html
sns.heatmap(cm, annot=True, fmt="d")
plt.ylabel("Actual Class")
plt.xlabel("Predicted Class")
plt.title("Confusion Matrix")
```

Train confusion matrix

Out[0]:

Text(0.5, 1.0, 'Confusion Matrix')



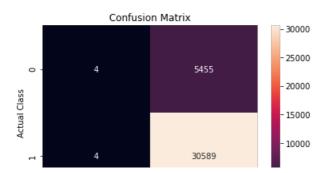
In [0]:

```
#y_test_pred = neigh.predict(set2_t)
cm = confusion_matrix(y_test,y_test_K)
print("Test confusion matrix")
# https://seaborn.pydata.org/generated/seaborn.heatmap.html
sns.heatmap(cm, annot=True, fmt="d")
plt.ylabel("Actual Class")
plt.xlabel("Predicted Class")
plt.title("Confusion Matrix")
```

Test confusion matrix

Out[0]:

Text(0.5, 1.0, 'Confusion Matrix')



3. Conclusions

In [0]:

```
# Please compare all your models using Prettytable library
```

In [115]:

```
# Please compare all your models using Prettytable library
#how to use pretty table http://zetcode.com/python/prettytable/
from prettytable import PrettyTable

tb = PrettyTable()
tb.field_names= ("Vectorizer", "Model","HyperParameter" ,"AUC")
tb.add_row(["BOW", "Auto", "51","0.528" ])
tb.add_row(["Tf-Idf", "Auto", "51", "0.599"])
tb.add_row(["Tf-Idf W2v", "Auto", "51", "0.622"])
tb.add_row(["Tf-Idf W2v", "Auto", "51", "0.631"])
tb.add_row(["Tf-Idf KBest", "Auto", "51", "0.601"])
print(tb.get_string(titles = "KNN - Observations"))
#print(tb)
```

+.		- + -		+-		. + -		- +
İ	Vectorizer	i	Model	İ	HyperParameter	İ	AUC	İ
+-		-+-			 51	•	0.528	
- 1	BOW		Auto		31		0.528	-
	Tf-Idf		Auto		51		0.599	
	AVG-W2v		Auto		51		0.622	
	Tf-Idf W2v		Auto		51		0.631	
	Tf-Idf KBest		Auto		51		0.601	
- 1		- 1						