### **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	Descri	
project_id	A unique identifier for the proposed project. <b>Example:</b> p03	
	Title of the project. <b>Exam</b>	
project_title	<ul> <li>Art Will Make You Ha</li> <li>First Grade</li> </ul>	
	TITSE GLAGE	
	Grade level of students for which the project is targeted. One of the following enumerated va	
<pre>project_grade_category</pre>	• Grades Pr	
p. 0,000_8. uu=_encoge. y	<ul><li>Grades</li><li>Grades</li></ul>	
	• Grades	
	One or more (comma-separated) subject categories for the project fro following enumerated list of va	
	Applied Lear	
	• Care & Hu • Health & Sp	
	<ul><li>History &amp; Ci</li><li>Literacy &amp; Lang</li></ul>	
<pre>project_subject_categories</pre>	• Math & Sci	
pi oject_subject_categoi ies	<ul><li>Music &amp; The</li><li>Special N</li></ul>	
	• Wa	
	Exam	
	• Music & The	
	<ul> <li>Music &amp; The</li> <li>Literacy &amp; Language, Math &amp; Sci</li> </ul>	
school_state		
school_state	• Literacy & Language, Math & Sci  State where school is located (Two-letter U.S. postal (https://en.wikipedia.org/wiki/List of U.S. state abbreviations#Postal co  Example  One or more (comma-separated) subject subcategories for the pr	
school_state project_subject_subcategories	State where school is located (Two-letter U.S. postal (https://en.wikipedia.org/wiki/List_of_U.S. state_abbreviations#Postal_co  Example  One or more (comma-separated) subject subcategories for the pr Exam	
	• Literacy & Language, Math & Sci  State where school is located (Two-letter U.S. postal (https://en.wikipedia.org/wiki/List of U.S. state abbreviations#Postal co  Example  One or more (comma-separated) subject subcategories for the pr	
	State where school is located (Two-letter U.S. postal (https://en.wikipedia.org/wiki/List_of_U.S. state_abbreviations#Postal_co  Example  One or more (comma-separated) subject subcategories for the present.  Exam  Lite	
	State where school is located (Two-letter U.S. postal (https://en.wikipedia.org/wiki/List of U.S. state abbreviations#Postal co  Example  One or more (comma-separated) subject subcategories for the pr Exam  Lite Literature & Writing, Social Scie  An explanation of the resources needed for the project. Exam  My students need hands on literacy materials to mar	
project_subject_subcategories	State where school is located (Two-letter U.S. postal (https://en.wikipedia.org/wiki/List of U.S. state abbreviations#Postal co  Example  One or more (comma-separated) subject subcategories for the pr  Exam  Lite Literature & Writing, Social Scie	
project_subject_subcategories	State where school is located (Two-letter U.S. postal (https://en.wikipedia.org/wiki/List of U.S. state abbreviations#Postal co  Example  One or more (comma-separated) subject subcategories for the pr Exam  Lite Literature & Writing, Social Scie  An explanation of the resources needed for the project. Exam  My students need hands on literacy materials to mar	
<pre>project_subject_subcategories  project_resource_summary</pre>	State where school is located (Two-letter U.S. postal (https://en.wikipedia.org/wiki/List of U.S. state abbreviations#Postal co  Example  One or more (comma-separated) subject subcategories for the pr Exam  Lite Literature & Writing, Social Scie  An explanation of the resources needed for the project. Exan  My students need hands on literacy materials to mar sensory ne	
<pre>project_subject_subcategories  project_resource_summary  project_essay_1</pre>	State where school is located (Two-letter U.S. postal (https://en.wikipedia.org/wiki/List of U.S. state abbreviations#Postal co Example  One or more (comma-separated) subject subcategories for the pr Exam  Lite  Literature & Writing, Social Scie  An explanation of the resources needed for the project. Exan  My students need hands on literacy materials to mar sensory ne	

Descri	Feature
Datetime when project application was submitted. <b>Example:</b> 2016-04 12:43:56	project_submitted_datetime
A unique identifier for the teacher of the proposed project. <b>Exa</b> l bdf8baa8fedef6bfeec7ae4ff1c1	teacher_id
Teacher's title. One of the following enumerated va	
•	
•	toochon mostiv
•	teacher_prefix
•	
•	
• Teac	

teacher\_number\_of\_previously\_posted\_projects

Number of project applications previously submitted by the same tea

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

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

Project\_is\_approved

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

 $\triangleleft$ 

<sup>\*</sup> See the section **Notes on the Essay Data** for more details about these features.

#### **Notes on the Essay Data**

learning and improve their school lives?"

Prior to May 17, 2016, the prompts for the essays were as follows:
\_\_project\_essay\_1:\_\_ "Introduce us to your classroom"
\_\_project\_essay\_2:\_\_ "Tell us more about your students"
\_\_project\_essay\_3:\_\_ "Describe how your students will use the materials you're requesting"
\_\_project\_essay\_3:\_\_ "Close by sharing why your project will make a difference"
Starting on May 17, 2016, the number of essays was reduced from 4 to 2, and the prompts for the first 2 essays were changed to the following:
\_\_project\_essay\_1:\_\_ "Describe your students: What makes your students special? Specific details about their background, your neighborhood, and your school are all helpful."

project essay 2: "About your project: How will these materials make a difference in your students'

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 [0]: | %matplotlib inline
        import warnings
        warnings.filterwarnings("ignore")
        import sqlite3
        import pandas as pd
        import numpy as np
        import nltk
        import string
        import matplotlib.pyplot as plt
        import seaborn as sns
        from sklearn.feature_extraction.text import TfidfTransformer
        from sklearn.feature_extraction.text import TfidfVectorizer
        from sklearn.feature extraction.text import CountVectorizer
        from sklearn.metrics import confusion matrix
        from sklearn import metrics
        from sklearn.metrics import roc_curve, auc
        from nltk.stem.porter import PorterStemmer
        import re
        # Tutorial about Python regular expressions: https://pymotw.com/2/re/
        import string
        from nltk.corpus import stopwords
        from nltk.stem import PorterStemmer
        from nltk.stem.wordnet import WordNetLemmatizer
        from gensim.models import Word2Vec
        from gensim.models import KeyedVectors
        import pickle
        from tqdm import tqdm
        import os
        from plotly import plotly
        import plotly.offline as offline
        import plotly.graph objs as go
        offline.init notebook mode()
        from collections import Counter
```

```
D:\installed\Anaconda3\lib\site-packages\gensim\utils.py:1197: UserWarning: d etected Windows; aliasing chunkize to chunkize_serial warnings.warn("detected Windows; aliasing chunkize to chunkize_serial")
```

# 1.1 Reading Data

```
In [0]: project_data = pd.read_csv('train_data.csv')
    resource_data = pd.read_csv('resources.csv')
```

```
In [0]: | 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' 'sc
         hool 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 [0]: | 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[0]:
                 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)
                                                                      14.95
```

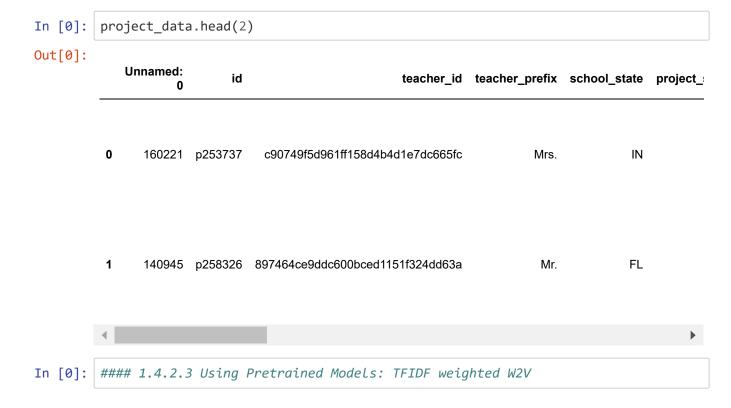
# 1.2 preprocessing of project\_subject\_categories

```
In [0]: | catogories = list(project data['project subject categories'].values)
        # remove special characters from list of strings python: https://stackoverflo
        w.com/a/47301924/4084039
        # https://www.geeksforgeeks.org/removing-stop-words-nltk-python/
        # https://stackoverflow.com/questions/23669024/how-to-strip-a-specific-word-fr
        om-a-string
        # https://stackoverflow.com/questions/8270092/remove-all-whitespace-in-a-strin
        g-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 & Scienc"]
        e", "Warmth", "Care & Hunger"]
                 if 'The' in j.split(): # this will split each of the catogory based on
        space "Math & Science"=> "Math", "&", "Science"
                    j=j.replace('The','') # if we have the words "The" we are going to
        replace it with ''(i.e removing 'The')
                 j = j.replace(' ','') # we are placeing all the ' '(space) with ''(emp
        ty) ex: "Math & Science" => "Math&Science"
                temp+=j.strip()+" " #" abc ".strip() will return "abc", remove the tra
        iling spaces
                temp = temp.replace('&','_') # we are replacing the & value into
            cat_list.append(temp.strip())
        project data['clean categories'] = cat list
        project_data.drop(['project_subject_categories'], axis=1, inplace=True)
        from collections import Counter
        my counter = Counter()
        for word in project_data['clean_categories'].values:
            my counter.update(word.split())
        cat_dict = dict(my_counter)
        sorted_cat_dict = dict(sorted(cat_dict.items(), key=lambda kv: kv[1]))
```

# 1.3 preprocessing of project\_subject\_subcategories

```
In [0]: | sub catogories = list(project data['project subject subcategories'].values)
        # remove special characters from list of strings python: https://stackoverflo
        w.com/a/47301924/4084039
        # https://www.geeksforgeeks.org/removing-stop-words-nltk-python/
        # https://stackoverflow.com/questions/23669024/how-to-strip-a-specific-word-fr
        om-a-string
        # https://stackoverflow.com/questions/8270092/remove-all-whitespace-in-a-strin
        g-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 & Scienc"]
        e", "Warmth", "Care & Hunger"]
                 if 'The' in j.split(): # this will split each of the catogory based on
        space "Math & Science"=> "Math", "&", "Science"
                    j=j.replace('The','') # if we have the words "The" we are going to
        replace it with ''(i.e removing 'The')
                 j = j.replace(' ','') # we are placeing all the ' '(space) with ''(emp
        ty) ex:"Math & Science"=>"Math&Science"
                temp +=j.strip()+" "#" abc ".strip() will return "abc", remove the tra
        iling 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/2289859
        5/4084039
        my counter = Counter()
        for word in project_data['clean_subcategories'].values:
            my_counter.update(word.split())
        sub cat dict = dict(my counter)
        sorted_sub_cat_dict = dict(sorted(sub_cat_dict.items(), key=lambda kv: kv[1]))
```

# 1.3 Text preprocessing



```
In [0]: # 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("="*50)
    print(project_data['essay'].values[20000])
    print("="*50)
    print(project_data['essay'].values[99999])
    print(project_data['essay'].values[99999])
    print("="*50)
```

My students are English learners that are working on English as their second or third languages. We are a melting pot of refugees, immigrants, and nativeborn Americans bringing the gift of language to our school. \r\n\r\n We have over 24 languages represented in our English Learner program with students at every level of mastery. We also have over 40 countries represented with the families within our school. Each student brings a wealth of knowledge and ex periences to us that open our eyes to new cultures, beliefs, and respect.\"Th e limits of your language are the limits of your world.\"-Ludwig Wittgenstein Our English learner's have a strong support system at home that begs for more resources. Many times our parents are learning to read and speak English alo ng side of their children. Sometimes this creates barriers for parents to be able to help their child learn phonetics, letter recognition, and other readi ng skills.\r\n\r\nBy providing these dvd's and players, students are able to continue their mastery of the English language even if no one at home is able to assist. All families with students within the Level 1 proficiency status, will be a offered to be a part of this program. These educational videos wil 1 be specially chosen by the English Learner Teacher and will be sent home re gularly to watch. The videos are to help the child develop early reading ski lls.\r\n\r\nParents that do not have access to a dvd player will have the opp ortunity to check out a dvd player to use for the year. The plan is to use t hese videos and educational dvd's for the years to come for other EL student s.\r\nnannan

The 51 fifth grade students that will cycle through my classroom this year al 1 love learning, at least most of the time. At our school, 97.3% of the stude nts receive free or reduced price lunch. Of the 560 students, 97.3% are minor ity students. \r\nThe school has a vibrant community that loves to get togeth er and celebrate. Around Halloween there is a whole school parade to show off the beautiful costumes that students wear. On Cinco de Mayo we put on a big f estival with crafts made by the students, dances, and games. At the end of th e year the school hosts a carnival to celebrate the hard work put in during t he school year, with a dunk tank being the most popular activity. My students will use these five brightly colored Hokki stools in place of regular, statio nary, 4-legged chairs. As I will only have a total of ten in the classroom an d not enough for each student to have an individual one, they will be used in a variety of ways. During independent reading time they will be used as speci al chairs students will each use on occasion. I will utilize them in place of chairs at my small group tables during math and reading times. The rest of th e day they will be used by the students who need the highest amount of moveme nt in their life in order to stay focused on school.\r\n\r\nWhenever asked wh at the classroom is missing, my students always say more Hokki Stools. They c an't get their fill of the 5 stools we already have. When the students are si tting in group with me on the Hokki Stools, they are always moving, but at th e same time doing their work. Anytime the students get to pick where they can sit, the Hokki Stools are the first to be taken. There are always students wh o head over to the kidney table to get one of the stools who are disappointed as there are not enough of them. \r\n\r\nWe ask a lot of students to sit for 7 hours a day. The Hokki stools will be a compromise that allow my students t o do desk work and move at the same time. These stools will help students to meet their 60 minutes a day of movement by allowing them to activate their co re muscles for balance while they sit. For many of my students, these chairs will take away the barrier that exists in schools for a child who can't sit s till.nannan

\_\_\_\_\_

How do you remember your days of school? Was it in a sterile environment with plain walls, rows of desks, and a teacher in front of the room? A typical day in our room is nothing like that. I work hard to create a warm inviting theme

d room for my students look forward to coming to each day.\r\n\r\nMy class is made up of 28 wonderfully unique boys and girls of mixed races in Arkansas.\r \nThey attend a Title I school, which means there is a high enough percentage of free and reduced-price lunch to qualify. Our school is an \"open classroom \" concept, which is very unique as there are no walls separating the classro oms. These 9 and 10 year-old students are very eager learners; they are like sponges, absorbing all the information and experiences and keep on wanting mo re.With these resources such as the comfy red throw pillows and the whimsical nautical hanging decor and the blue fish nets, I will be able to help create the mood in our classroom setting to be one of a themed nautical environment. Creating a classroom environment is very important in the success in each and every child's education. The nautical photo props will be used with each chil d as they step foot into our classroom for the first time on Meet the Teacher evening. I'll take pictures of each child with them, have them developed, and then hung in our classroom ready for their first day of 4th grade. This kind gesture will set the tone before even the first day of school! The nautical t hank you cards will be used throughout the year by the students as they creat e thank you cards to their team groups.\r\n\r\nYour generous donations will h elp me to help make our classroom a fun, inviting, learning environment from day one.\r\n\r\nIt costs lost of money out of my own pocket on resources to g et our classroom ready. Please consider helping with this project to make our new school year a very successful one. Thank you!nannan

\_\_\_\_\_

My kindergarten students have varied disabilities ranging from speech and lan guage delays, cognitive delays, gross/fine motor delays, to autism. They are eager beavers and always strive to work their hardest working past their limi tations. \r\n\r\nThe materials we have are the ones I seek out for my student s. I teach in a Title I school where most of the students receive free or red uced price lunch. Despite their disabilities and limitations, my students lo ve coming to school and come eager to learn and explore. Have you ever felt li ke you had ants in your pants and you needed to groove and move as you were i n a meeting? This is how my kids feel all the time. The want to be able to mo ve as they learn or so they say. Wobble chairs are the answer and I love then because they develop their core, which enhances gross motor and in Turn fine motor skills. \r\nThey also want to learn through games, my kids don't want t o sit and do worksheets. They want to learn to count by jumping and playing. Physical engagement is the key to our success. The number toss and color and shape mats can make that happen. My students will forget they are doing work and just have the fun a 6 year old deserves.nannan

\_\_\_\_\_

The mediocre teacher tells. The good teacher explains. The superior teacher d emonstrates. The great teacher inspires. -William A. Ward\r\n\r\nMy school ha s 803 students which is makeup is 97.6% African-American, making up the large st segment of the student body. A typical school in Dallas is made up of 23. 2% African-American students. Most of the students are on free or reduced lun ch. We aren't receiving doctors, lawyers, or engineers children from rich bac kgrounds or neighborhoods. As an educator I am inspiring minds of young child ren and we focus not only on academics but one smart, effective, efficient, a nd disciplined students with good character. In our classroom we can utilize t he Bluetooth for swift transitions during class. I use a speaker which does n't amplify the sound enough to receive the message. Due to the volume of my speaker my students can't hear videos or books clearly and it isn't making th e lessons as meaningful. But with the bluetooth speaker my students will be a ble to hear and I can stop, pause and replay it at any time.\r\nThe cart will allow me to have more room for storage of things that are needed for the day and has an extra part to it I can use. The table top chart has all of the le tter, words and pictures for students to learn about different letters and it

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

```
In [0]: sent = decontracted(project_data['essay'].values[20000])
    print(sent)
    print("="*50)
```

My kindergarten students have varied disabilities ranging from speech and lan guage delays, cognitive delays, gross/fine motor delays, to autism. They are eager beavers and always strive to work their hardest working past their limi tations. \r\n\r\nThe materials we have are the ones I seek out for my student s. I teach in a Title I school where most of the students receive free or red uced price lunch. Despite their disabilities and limitations, my students lo ve coming to school and come eager to learn and explore. Have you ever felt li ke you had ants in your pants and you needed to groove and move as you were i n a meeting? This is how my kids feel all the time. The want to be able to mo ve as they learn or so they say. Wobble chairs are the answer and I love then because they develop their core, which enhances gross motor and in Turn fine motor skills. \r\nThey also want to learn through games, my kids do not want to sit and do worksheets. They want to learn to count by jumping and playing. Physical engagement is the key to our success. The number toss and color and shape mats can make that happen. My students will forget they are doing work and just have the fun a 6 year old deserves.nannan

```
In [0]: # \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)
```

My kindergarten students have varied disabilities ranging from speech and lan guage delays, cognitive delays, gross/fine motor delays, to autism. They are eager beavers and always strive to work their hardest working past their limi tations. The materials we have are the ones I seek out for my students. I teach in a Title I school where most of the students receive free or reduced price lunch. Despite their disabilities and limitations, my students love co ming to school and come eager to learn and explore. Have you ever felt like yo u had ants in your pants and you needed to groove and move as you were in a m eeting? This is how my kids feel all the time. The want to be able to move as they learn or so they say. Wobble chairs are the answer and I love then becaus e they develop their core, which enhances gross motor and in Turn fine motor They also want to learn through games, my kids do not want to sit a nd do worksheets. They want to learn to count by jumping and playing. Physica l engagement is the key to our success. The number toss and color and shape m ats can make that happen. My students will forget they are doing work and jus t have the fun a 6 year old deserves.nannan

```
In [0]: #remove spacial character: https://stackoverflow.com/a/5843547/4084039
sent = re.sub('[^A-Za-z0-9]+', ' ', sent)
print(sent)
```

My kindergarten students have varied disabilities ranging from speech and lan guage delays cognitive delays gross fine motor delays to autism They are eage r beavers and always strive to work their hardest working past their limitati ons The materials we have are the ones I seek out for my students I teach in a Title I school where most of the students receive free or reduced price lun ch Despite their disabilities and limitations my students love coming to scho ol and come eager to learn and explore Have you ever felt like you had ants i n your pants and you needed to groove and move as you were in a meeting This is how my kids feel all the time The want to be able to move as they learn or so they say Wobble chairs are the answer and I love then because they develop their core which enhances gross motor and in Turn fine motor skills They also want to learn through games my kids do not want to sit and do worksheets They want to learn to count by jumping and playing Physical engagement is the key to our success The number toss and color and shape mats can make that happen My students will forget they are doing work and just have the fun a 6 year ol d deserves 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', 'yours', 'yourself', 'yourselves', 'he'
        , 'him', 'his', 'himself', \
                    'she', "she's", 'her', 'hers', 'herself', 'it', "it's", 'its', 'it
        self', 'they', 'them', 'their',\
                    'theirs', 'themselves', 'what', 'which', 'who', 'whom', 'this', 't
        hat', "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', 'becau
        se', 'as', 'until', 'while', 'of', \
                   'at', 'by', 'for', 'with', 'about', 'against', 'between', 'into',
        'off', 'over', 'under', 'again', 'further',\
                    'then', 'once', 'here', 'there', 'when', 'where', 'why', 'how', 'a
        11', 'any', 'both', 'each', 'few', 'more',\
                    'most', 'other', 'some', 'such', 'only', 'own', 'same', 'so', 'tha
        n', 'too', 'very', \
                    's', 't', 'can', 'will', 'just', 'don', "don't", 'should', "shoul
        d've", 'now', 'd', 'll', 'm', 'o', 're', \
                   've', 'y', 'ain', 'aren', "aren't", 'couldn', "couldn't", 'didn',
        "didn't", 'doesn', "doesn't", 'hadn',\
                   "hadn't", 'hasn', "hasn't", 'haven', "haven't", 'isn', "isn't", 'm
        a', 'mightn', "mightn't", 'mustn',\
                   "mustn't", 'needn', "needn't", 'shan', "shan't", 'shouldn', "shoul
        dn't", 'wasn', "wasn't", 'weren', "weren't", \
                    'won', "won't", 'wouldn', "wouldn't"]
```

```
In [0]: # Combining all the above stundents
    from tqdm import tqdm
    preprocessed_essays = []
    # tqdm is for printing the status bar
    for sentance in tqdm(project_data['essay'].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 = ''.join(e for e in sent.split() if e not in stopwords)
        preprocessed_essays.append(sent.lower().strip())
```

```
100%| 109248/109248 [00:53<00:00, 2055.17it/s]
```

```
In [0]: # after preprocesing
    preprocessed_essays[20000]
```

Out[0]: 'my kindergarten students varied disabilities ranging speech language delays cognitive delays gross fine motor delays autism they eager beavers always str ive work hardest working past limitations the materials ones i seek students i teach title i school students receive free reduced price lunch despite disa bilities limitations students love coming school come eager learn explore hav e ever felt like ants pants needed groove move meeting this kids feel time th e want able move learn say wobble chairs answer i love develop core enhances gross motor turn fine motor skills they also want learn games kids not want s it worksheets they want learn count jumping playing physical engagement key s uccess the number toss color shape mats make happen my students forget work f un 6 year old deserves nannan'

# 1.4 Preprocessing of `project\_title`

```
In [0]: # similarly you can preprocess the titles also
```

# 1.5 Preparing data for models

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

<a href="https://www.appliedaicourse.com/course/applied-ai-course-online/lessons/handling-categorical-and-numerical-features/">https://www.appliedaicourse.com/course-online/lessons/handling-categorical-and-numerical-features/</a>)

```
In [0]: # 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()), lowercas
    e=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']
    Shape of matrix after one hot encodig (109248, 9)
```

```
In [0]: # we use count vectorizer to convert the values into one
    vectorizer = CountVectorizer(vocabulary=list(sorted_sub_cat_dict.keys()), lowe
    rcase=False, binary=True)
    sub_categories_one_hot = vectorizer.fit_transform(project_data['clean_subcateg
    ories'].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', 'NutritionEducati
    on', 'Warmth', 'Care_Hunger', 'SocialSciences', 'PerformingArts', 'CharacterE
    ducation', 'TeamSports', 'Other', 'College_CareerPrep', 'Music', 'History_Geo
    graphy', 'Health_LifeScience', 'EarlyDevelopment', 'ESL', 'Gym_Fitness', 'Env
    ironmentalScience', 'VisualArts', 'Health_Wellness', 'AppliedSciences', 'Spec
    ialNeeds', '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\_ca
tegory also

#### 1.5.2 Vectorizing Text data

#### 1.5.2.1 Bag of words

```
In [0]: # We are considering only the words which appeared in at least 10 documents(ro
    ws 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, 16623)

In [0]: # you can vectorize the title also
    # before you vectorize the title make sure you preprocess it
```

#### 1.5.2.2 TFIDF vectorizer

```
In [0]: 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, 16623)
```

#### 1.5.2.3 Using Pretrained Models: Avg W2V

```
In [0]:
        # Reading glove vectors in python: https://stackoverflow.com/a/38230349/408403
        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 coup
        us", \
              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-save-and-load-variables-in-python/
        import pickle
        with open('glove_vectors', 'wb') as f:
            pickle.dump(words courpus, f)
```

'\n# Reading glove vectors in python: https://stackoverflow.com/a/38230349/40 84039\ndef loadGloveModel(gloveFile):\n print ("Loading Glove Model")\n f = open(gloveFile,\'r\', encoding="utf8")\n  $model = {}\n$ for line in t word = splitLine[0]\n  $adm(f):\n$ splitLine = line.split()\n embedding = np.array([float(val) for val in splitLine[1:]])\n model[wo rd] = embedding\n print ("Done.",len(model)," words loaded!")\n return model\nmodel = loadGloveModel(\'glove.42B.300d.txt\')\n\n# =========== ======\nOutput:\n \nLoading Glove Model\n1917495it [06:32, 4879.69it/ s]\nDone. 1917495 words loaded!\n\n# ==========\n\nwords = words.extend(i.split(\' \'))\n\nfor i in p []\nfor i in preproced texts:\n words.extend(i.split(\' \'))\nprint("all the words in t reproced titles:\n he coupus", len(words))\nwords = set(words)\nprint("the unique words in the c oupus", len(words))\n\ninter words = set(model.keys()).intersection(words)\np rint("The number of words that are present in both glove vectors and our coup len(inter\_words),"(",np.round(len(inter\_words)/len(words)\*100, 3),"%)")\n\nwords courpus = {}\nwords glove = set(model.keys())\nfor i in wor ds:\n if i in words glove:\n words courpus[i] = model[i]\nprint("wo rd 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-v ariables-in-python/\n\nimport pickle\nwith open(\'glove vectors\', \'wb\') as pickle.dump(words courpus, f)\n\n\n'

```
In [0]: # 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%| 109248/109248 [00:27<00:00, 3968.04it/s]

109248
```

#### 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(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 [0]: # average Word2Vec
        # compute average word2vec for each review.
        tfidf w2v vectors = []; # the avq-w2v for each sentence/review is stored in th
        is 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/revie
        W
            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 v
        alue((sentence.count(word)/len(sentence.split())))
                    tf_idf = dictionary[word]*(sentence.count(word)/len(sentence.split
        ())) # getting the tfidf 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]))
        | 109248/109248 [03:22<00:00, 540.56it/s]
        109248
        300
In [0]: # Similarly you can vectorize for title also
```

#### 1.5.3 Vectorizing Numerical features

```
In [0]: # check this one: https://www.youtube.com/watch?v=0HOqOcln3Z4&t=530s
        # standardization sklearn: https://scikit-learn.org/stable/modules/generated/s
        klearn.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 32
             ... 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 mea
        n and standard deviation of this data
        print(f"Mean : {price scalar.mean [0]}, Standard deviation : {np.sqrt(price sc
        alar.var_[0])}")
        # Now standardize the data with above maen and variance.
        price standardized = price scalar.transform(project data['price'].values.resha
        pe(-1, 1)
In [0]: price_standardized
Out[0]: array([[0.00098843, 0.00191166, 0.00330448, ..., 0.00153418, 0.00046704,
```

#### 1.5.4 Merging all the above features

0.0007026511)

we need to merge all the numerical vectors i.e catogorical, text, numerical vectors

```
In [0]: print(categories one hot.shape)
        print(sub categories one hot.shape)
        print(text bow.shape)
        print(price standardized.shape)
        (109248, 9)
        (109248, 30)
        (109248, 16623)
        (109248, 1)
In [0]: # merge two sparse matrices: https://stackoverflow.com/a/19710648/4084039
        from scipy.sparse import hstack
        # with the same hstack function we are concatinating a sparse matrix and a den
        se matirx :)
        X = hstack((categories_one_hot, sub_categories_one_hot, text_bow, price_standa
        rdized))
        X.shape
Out[0]: (109248, 16663)
```

#### **Computing Sentiment Scores**

```
In [3]:
        import nltk
        from nltk.sentiment.vader import SentimentIntensityAnalyzer
        import nltk
        nltk.download('vader lexicon')
        sid = SentimentIntensityAnalyzer()
        for sentiment = '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 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 o
        ur school is a caring community of successful \
        learners which can be seen through collaborative student project based learnin
        g in and out of the classroom kindergarteners \
        in my class love to work with hands on materials and have many different oppor
        tunities to practice a skill before it is\
        mastered having the social skills to work cooperatively with friends is a cruc
        ial aspect of the kindergarten curriculum\
        montana is the perfect place to learn about agriculture and nutrition my stude
        nts 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 writi
        ng 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 knowledge \
        of where the ingredients came from as well as how it is healthy for their bodi
        es 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 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'
        ss = sid.polarity scores(for sentiment)
        for k in ss:
            print('{0}: {1}, '.format(k, ss[k]), end='')
        # we can use these 4 things as features/attributes (neg, neu, pos, compound)
        # neg: 0.0, neu: 0.753, pos: 0.247, compound: 0.93
        neg: 0.01, neu: 0.745, pos: 0.245, compound: 0.9975,
        [nltk data] Downloading package vader lexicon to
                        C:\Users\LENOVO\AppData\Roaming\nltk data...
        [nltk data]
        [nltk data]
                      Package vader lexicon is already up-to-date!
```

# **Assignment 5: Logistic Regression**

#### [Task-1] Logistic Regression(either SGDClassifier with log loss, or LogisticRegression) on these feature sets

- Set 1: categorical, numerical features + project\_title(BOW) + preprocessed\_eassay (`BOW with bigrams` with `min\_df=10` and `max\_features=5000`)
- Set 2: categorical, numerical features + project\_title(TFIDF)+ preprocessed\_eassay (`TFIDF with bigrams` with `min\_df=10` and `max\_features=5000`)
- Set 3: categorical, numerical features + project\_title(AVG W2V)+ preprocessed\_eassay (AVG W2V)
- Set 4: categorical, numerical features + project\_title(TFIDF W2V)+ preprocessed\_essay (TFIDF W2V)

#### 2. Hyper paramter tuning (find best hyper parameters corresponding the algorithm that you choose)

- Find the best hyper parameter which will give the maximum <u>AUC</u>
   (<a href="https://www.appliedaicourse.com/course/applied-ai-course-online/lessons/receiver-operating-characteristic-curve-roc-curve-and-auc-1/">https://www.appliedaicourse.com/course/applied-ai-course-online/lessons/receiver-operating-characteristic-curve-roc-curve-and-auc-1/</a>) value
- Find the best hyper paramter using k-fold cross validation or simple cross validation data
- Use gridsearch cv or randomsearch cv or you can also write your own for loops to do this task of hyperparameter tuning

#### 3. Representation of results

 You need to plot the performance of model both on train data and cross validation data for each hyper parameter, like shown in the figure.



 Once after you found the best hyper parameter, you need to train your model with it, and find the AUC on test data and plot the ROC curve on both train and test.



Along with plotting ROC curve, you need to print the <u>confusion matrix</u>
 (<a href="https://www.appliedaicourse.com/course/applied-ai-course-online/lessons/confusion-matrix-tpr-fpr-fnr-tnr-1/">https://www.appliedaicourse.com/course/applied-ai-course-online/lessons/confusion-matrix-tpr-fpr-fnr-tnr-1/</a>) with predicted and original labels of test data points. Please visualize your confusion matrices using <u>seaborn heatmaps</u>.



(https://seaborn.pydata.org/generated/seaborn.heatmap.html)

(https://seaborn.pydata.org/generated/seaborn.heatmap.html)

(https://seaborn.pydata.org/generated/seaborn.heatmap.html)

- 4. [Task-2] Apply Logistic Regression on the below feature set Set 5 by finding the best hyper parameter as suggested in step 2 and step 3.
- 5. Consider these set of features Set 5:
  - school state: categorical data
  - clean\_categories : categorical data
  - clean subcategories : categorical data
  - project grade category :categorical data
  - · teacher prefix : categorical data
  - quantity : numerical data
  - teacher\_number\_of\_previously\_posted\_projects : numerical data
  - price: numerical data
  - sentiment score's of each of the essay : numerical data

- number of words in the title : numerical data
- number of words in the combine essays : numerical data

And apply the Logistic regression on these features by finding the best hyper paramter as suggested in step 2 and step 3

(https://seaborn.pydata.org/generated/seaborn.heatmap.html)

6. Conclusion (https://seaborn.pydata.org/generated/seaborn.heatmap.html)

(https://seaborn.pydata.org/generated/seaborn.heatmap.html)

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
 (<a href="https://seaborn.pydata.org/generated/seaborn.heatmap.html">https://seaborn.pydata.org/generated/seaborn.heatmap.html</a>) link
 (<a href="https://zetcode.com/python/prettytable/">https://zetcode.com/python/prettytable/</a>)



#### 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-leakage, 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 data.
- 4. For more details please go through this <a href="https://soundcloud.com/applied-ai-course/leakage-bow-and-tfidf">https://soundcloud.com/applied-ai-course/leakage-bow-and-tfidf</a>)

# 2. Logistic Regression

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

```
In [2]: | %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 notebook as tqdm1
        from tqdm import tqdm
        import time
        import os
        from plotly import plotly
        import plotly.offline as offline
        import plotly.graph objs as go
        offline.init notebook mode()
        from collections import Counter
        from sklearn.model selection import train test split
```

```
C:\Users\LENOVO\Anaconda3\lib\site-packages\smart_open\ssh.py:34: UserWarnin
g: paramiko missing, opening SSH/SCP/SFTP paths will be disabled. `pip insta
ll paramiko` to suppress
  warnings.warn('paramiko missing, opening SSH/SCP/SFTP paths will be disable
d. `pip install paramiko` to suppress')
C:\Users\LENOVO\Anaconda3\lib\site-packages\gensim\utils.py:1197: UserWarnin
g: detected Windows; aliasing chunkize to chunkize_serial
  warnings.warn("detected Windows; aliasing chunkize to chunkize serial")
```

# Text preprocessing(1)

```
In [5]: catogories = list(project_data['project_subject_categories'].values)
        # remove special characters from list of strings python: https://stackoverflo
        w.com/a/47301924/4084039
        # https://www.geeksforgeeks.org/removing-stop-words-nltk-python/
        # https://stackoverflow.com/questions/23669024/how-to-strip-a-specific-word-fr
        om-a-string
        # https://stackoverflow.com/questions/8270092/remove-all-whitespace-in-a-strin
        g-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 & Scienc"]
        e", "Warmth", "Care & Hunger"]
                if 'The' in j.split(): # this will split each of the catogory based on
        space "Math & Science"=> "Math", "&", "Science"
                    j=j.replace('The','') # if we have the words "The" we are going to
        replace it with ''(i.e removing 'The')
                j = j.replace(' ','') # we are placeing all the ' '(space) with ''(emp
        ty) ex: "Math & Science" => "Math&Science"
                temp+=j.strip()+" " #" abc ".strip() will return "abc", remove the tra
        iling spaces
                temp = temp.replace('&','_') # we are replacing the & value into
            cat list.append(temp.strip())
```

```
In [6]: project_data['clean_categories'] = cat_list
    project_data.drop(['project_subject_categories'], axis=1, inplace=True)
    project_data.head(5)
```

#### Out[6]:

_	Unnamed: 0	id	teacher_id	teacher_prefix	school_state	project_
C	160221	p253737	c90749f5d961ff158d4b4d1e7dc665fc	Mrs.	IN	
1	140945	p258326	897464ce9ddc600bced1151f324dd63a	Mr.	FL	
2	21895	p182444	3465aaf82da834c0582ebd0ef8040ca0	Ms.	AZ	
3	45	p246581	f3cb9bffbba169bef1a77b243e620b60	Mrs.	KY	
4		p104768	be1f7507a41f8479dc06f047086a39ec	Mrs.	TX	

```
In [8]: # dict sort by value python: https://stackoverflow.com/a/613218/4084039
    cat_dict = dict(my_counter)
    sorted_cat_dict = dict(sorted(cat_dict.items(), key=lambda kv: kv[1]))

# ind = np.arange(len(sorted_cat_dict))
# plt.figure(figsize=(20,5))
# p1 = plt.bar(ind, list(sorted_cat_dict.values()))

# plt.ylabel('Projects')
# plt.vtitle('% of projects aproved category wise')
# plt.xticks(ind, list(sorted_cat_dict.keys()))
# plt.show()
# print(sorted_cat_dict)
In [9]: sub_catogories = list(project_data['project_subject_subcategories'].values)
# print(sorted_cat_dict)
```

```
# remove special characters from list of strings python: https://stackoverflo
w.com/a/47301924/4084039
# https://www.geeksforgeeks.org/removing-stop-words-nltk-python/
# https://stackoverflow.com/questions/23669024/how-to-strip-a-specific-word-fr
om-a-string
# https://stackoverflow.com/questions/8270092/remove-all-whitespace-in-a-strin
g-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 & Scienc
e", "Warmth", "Care & Hunger"]
        if 'The' in j.split(): # this will split each of the catogory based on
space "Math & Science"=> "Math", "&", "Science"
            j=j.replace('The','') # if we have the words "The" we are going to
replace it with ''(i.e removing 'The')
        j = j.replace(' ','') # we are placeing all the ' '(space) with ''(emp
ty) ex: "Math & Science" => "Math&Science"
       temp +=j.strip()+" "#" abc ".strip() will return "abc", remove the tra
iling spaces
        temp = temp.replace('&',' ')
   sub cat list.append(temp.strip())
```

```
project_data['clean_subcategories'] = sub cat list
In [10]:
         project_data.drop(['project_subject_subcategories'], axis=1, inplace=True)
         project data.head(2)
Out[10]:
             Unnamed:
                                                  teacher_id teacher_prefix school_state project_:
                           id
                                                                                 IN
               160221 p253737
                               c90749f5d961ff158d4b4d1e7dc665fc
                                                                    Mrs.
               140945 p258326 897464ce9ddc600bced1151f324dd63a
                                                                                 FL
                                                                     Mr.
         # count of all the words in corpus python: https://stackoverflow.com/a/2289859
In [11]:
         5/4084039
         from collections import Counter
         my_counter = Counter()
         for word in project data['clean subcategories'].values:
              my counter.update(word.split())
In [12]: | # dict sort by value python: https://stackoverflow.com/a/613218/4084039
         sub cat dict = dict(my counter)
         sorted sub cat dict = dict(sorted(sub cat dict.items(), key=lambda kv: kv[1]))
         # ind = np.arange(len(sorted sub cat dict))
         # plt.figure(figsize=(20,5))
         # p1 = plt.bar(ind, list(sorted sub cat dict.values()))
         # plt.ylabel('Projects')
         # plt.title('% of projects aproved state wise')
         # plt.xticks(ind, list(sorted sub cat dict.keys()))
         # plt.show()
In [13]: # merge two column text dataframe:
         project_data["essay"] = project_data["project_essay_1"].map(str) +\
                                  project data["project essay 2"].map(str) + \
                                  project_data["project_essay_3"].map(str) + \
                                  project data["project essay 4"].map(str)
```

#### Out[14]:

	id	price	quantity
0	p000001	459.56	7
1	p000002	515.89	21

```
In [15]: # join two dataframes in python:
    project_data = pd.merge(project_data, price_data, on='id', how='left')
```

```
In [16]:
          #presence of the numerical digits in a strings with numeric : https://stackove
          rflow.com/a/19859308/8089731
          def hasNumbers(inputString):
              return any(i.isdigit() for i in inputString)
          p1 = project_data[['id','project_resource_summary']]
          p1 = pd.DataFrame(data=p1)
          p1.columns = ['id','digits_in_summary']
          p1['digits in summary'] = p1['digits in summary'].map(hasNumbers)
          # https://stackoverflow.com/a/17383325/8089731
          p1['digits_in_summary'] = p1['digits_in_summary'].astype(int)
          project_data = pd.merge(project_data, p1, on='id', how='left')
          project data.head(5)
Out[16]:
             Unnamed:
                            id
                                                    teacher_id teacher_prefix school_state project_:
                                 c90749f5d961ff158d4b4d1e7dc665fc
                                                                                    IN
                160221 p253737
                                                                      Mrs.
                140945 p258326 897464ce9ddc600bced1151f324dd63a
                                                                                    FL
                                                                        Mr.
           2
                 21895 p182444 3465aaf82da834c0582ebd0ef8040ca0
                                                                                    ΑZ
                                                                       Ms.
           3
                                f3cb9bffbba169bef1a77b243e620b60
                                                                                    KY
                   45 p246581
                                                                      Mrs.
                                                                                    TX
                172407 p104768
                              be1f7507a41f8479dc06f047086a39ec
                                                                      Mrs.
          5 rows × 21 columns
```

# Text preprocessing(2)

```
In [17]: # 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"\'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"\'ve", " have", phrase)
    phrase = re.sub(r"\'ve", " am", phrase)
    return phrase
```

```
In [18]: # 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', 'it
          self', 'they', 'them', 'their',\
                      'theirs', 'themselves', 'what', 'which', 'who', 'whom', 'this', 't
          hat', "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', 'becau
          se', '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', 'a
         11', 'any', 'both', 'each', 'few', 'more', \
                      'most', 'other', 'some', 'such', 'only', 'own', 'same', 'so', 'tha
          n', 'too', 'very', \
                      's', 't', 'can', 'will', 'just', 'don', "don't", 'should', "shoul
         d've", 'now', 'd', 'll', 'm', 'o', 're', \
          've', 'y', 'ain', 'aren', "aren't", 'couldn', "couldn't", 'didn', "didn't", 'doesn', "doesn't", 'hadn',\
                      "hadn't", 'hasn', "hasn't", 'haven', "haven't", 'isn', "isn't", 'm
          a', 'mightn', "mightn't", 'mustn',\
                      "mustn't", 'needn', "needn't", 'shan', "shan't", 'shouldn', "shoul
          dn't", 'wasn', "wasn't", 'weren', "weren't", \
                      'won', "won't", 'wouldn', "wouldn't"]
```

```
In [19]: # Combining all the above statemennts
    from tqdm import tqdm
    preprocessed_essays = []
    # tqdm is for printing the status bar
    for sentance in tqdm1(project_data['essay'].values):
        sent = decontracted(sentance)
        sent = sent.replace('\\r', '')
        sent = sent.replace('\\r', '')
        sent = sent.replace('\\n', '')
        sent = re.sub('[^A-Za-z0-9]+', '', sent)
        sent = re.sub('nannan', '', sent)
        # https://gist.github.com/sebleier/554280
        sent = ''.join(e for e in sent.split() if e not in stopwords)
        preprocessed_essays.append(sent.lower().strip())
```

```
In [20]: from tqdm import tqdm
    preprocessed_titles = []
# tqdm is for printing the status bar
    for title in tqdm1(project_data['project_title'].values):
        _title = decontracted(title)
        _title = _title.replace('\\r', ' ')
        _title = _title.replace('\\", ' ')
        _title = _title.replace('\\", ' ')
        _title = _title.replace('\\n', ' ')
        _title = re.sub('[^A-Za-z0-9]+', ' ', _title)
        # https://gist.github.com/sebleier/554280
        _title = ' '.join(e for e in _title.split() if e not in stopwords)
        preprocessed_titles.append(_title.lower().strip())
```

```
In [21]: preprocessed_titles[1000]
```

Out[21]: 'sailing into super 4th grade year'

```
In [22]: | project grade categories = list(project data['project grade category'].values)
         # remove special characters from list of strings python: https://stackoverflo
         w.com/a/47301924/4084039
         # https://www.geeksforgeeks.org/removing-stop-words-nltk-python/
         # https://stackoverflow.com/questions/23669024/how-to-strip-a-specific-word-fr
         om-a-string
         # https://stackoverflow.com/questions/8270092/remove-all-whitespace-in-a-strin
         g-in-python
         project grade cat list = []
         for i in tqdm1(project_grade_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 & Scienc
         e", "Warmth", "Care & Hunger"]
                 if 'The' in j.split(): # this will split each of the catogory based on
         space "Math & Science"=> "Math", "&", "Science"
                     j=j.replace('The','') # if we have the words "The" we are going to
         replace it with ''(i.e removing 'The')
                 j = j.replace(' ','') # we are placeing all the ' '(space) with ''(emp
         ty) ex:"Math & Science"=>"Math&Science"
                 temp +=j.strip()+" "#" abc ".strip() will return "abc", remove the tra
         iling spaces
                 temp = temp.replace('&',' ')
             project grade cat list.append(temp.strip())
```

#### Out[23]:

Unnamed: id teacher\_id teacher\_prefix school\_state project\_:

0 160221 p253737 c90749f5d961ff158d4b4d1e7dc665fc Mrs. IN

1 140945 p258326 897464ce9ddc600bced1151f324dd63a Mr. FL

2 rows × 21 columns

**◆** 

```
In [24]:
         project_data.drop(['project_essay_1','project_essay_2','project_essay_3','proj
          ect_essay_4'], axis=1, inplace=True)
          project data.head(2)
Out[24]:
             Unnamed:
                           id
                                                  teacher_id teacher_prefix school_state project_:
                               c90749f5d961ff158d4b4d1e7dc665fc
                                                                                 IN
               160221 p253737
                                                                    Mrs.
               140945 p258326 897464ce9ddc600bced1151f324dd63a
                                                                                 FL
          1
                                                                     Mr.
                                                                                         In [25]:
         #Replacing Nan's with maximum occured value: https://stackoverflow.com/a/51053
          916/8089731
          project_data['teacher_prefix'].value_counts().argmax()
          project data.fillna(value=project data['teacher prefix'].value counts().argmax
          (),axis=1,inplace=True)
         project data['preprocessed essays'] = preprocessed essays
In [26]:
          project data['preprocessed titles'] = preprocessed titles
In [27]: project data.columns
Out[27]: Index(['Unnamed: 0', 'id', 'teacher_id', 'teacher_prefix', 'school_state',
                 'project_submitted_datetime', 'project_title',
                 'project_resource_summary',
                 'teacher number of previously posted projects', 'project is approved',
                 'clean_categories', 'clean_subcategories', 'essay', 'price', 'quantit
         у',
                 'digits_in_summary', 'clean_project_grade_category',
                 'preprocessed_essays', 'preprocessed_titles'],
                dtype='object')
In [ ]:
```

# 2.2 Make Data Model Ready: encoding numerical, categorical features

```
In [30]: X_train, X_test, y_train, y_test = train_test_split(project_data, project_data[
    'project_is_approved'], test_size=0.33, stratify = project_data['project_is_approved'])
# X_train, X_cv, y_train, y_cv = train_test_split(X_train, y_train, test_size=
    0.33, stratify=y_train)

X_train.drop(['project_is_approved'], axis=1, inplace=True)
X_test.drop(['project_is_approved'], axis=1, inplace=True)
# X_cv.drop(['project_is_approved'], axis=1, inplace=True)
print(X_train.shape)
print(X_test.shape)

(73196, 18)
(36052, 18)
```

### 1.4.1 Vectorizing Categorical data

In [31]: # we use count vectorizer to convert the values into one hot encoded features

```
from sklearn.feature extraction.text import CountVectorizer
         vectorizer cat = CountVectorizer(vocabulary=list(sorted cat dict.keys()), lowe
         rcase=False, binary=True)
         vectorizer cat.fit(X train['clean categories'].values)
         print(vectorizer_cat.get_feature_names())
         categories one hot train = vectorizer cat.transform(X train['clean categories'
         ].values)
         # categories one hot cv = vectorizer cat.transform(X cv['clean categories'].va
         Lues)
         categories_one_hot_test = vectorizer_cat.transform(X_test['clean_categories'].
         values)
         print("Shape of matrix after one hot encodig train ", categories one hot train.
         shape)
         # print("Shape of matrix after one hot encodig cv ",categories one hot cv.shap
         print("Shape of matrix after one hot encodig_test ",categories_one_hot_test.sh
         ape)
         ['Warmth', 'Care_Hunger', 'History_Civics', 'Music_Arts', 'AppliedLearning',
         'SpecialNeeds', 'Health Sports', 'Math Science', 'Literacy Language']
         Shape of matrix after one hot encodig train (73196, 9)
         Shape of matrix after one hot encoding test (36052, 9)
In [32]: # we use count vectorizer to convert the values into one hot encoded features
         vectorizer sub cat = CountVectorizer(vocabulary=list(sorted sub cat dict.keys
         ()), lowercase=False, binary=True)
         vectorizer sub cat.fit(X train['clean subcategories'].values)
         print(vectorizer_sub_cat.get_feature_names())
         sub categories one hot train = vectorizer sub cat.transform(X train['clean sub
         categories'].values)
         # sub categories one hot cv = vectorizer sub cat.transform(X cv['clean subcate
         gories'].values)
         sub_categories_one_hot_test = vectorizer_sub_cat.transform(X_test['clean_subca')
         tegories'l.values)
         print("Shape of matrix after one hot encodig train ", sub categories one hot tr
         ain.shape)
         # print("Shape of matrix after one hot encodig cv ",sub categories one hot cv.
         shape)
         print("Shape of matrix after one hot encodig_test ",sub_categories_one_hot_tes
         t.shape)
         ['Economics', 'CommunityService', 'FinancialLiteracy', 'ParentInvolvement',
         'Extracurricular', 'Civics_Government', 'ForeignLanguages', 'NutritionEducati
         on', 'Warmth', 'Care_Hunger', 'SocialSciences', 'PerformingArts', 'CharacterE
         ducation', 'TeamSports', 'Other', 'College CareerPrep', 'Music', 'History Geo
         graphy', 'Health_LifeScience', 'EarlyDevelopment', 'ESL', 'Gym_Fitness', 'Env
         ironmentalScience', 'VisualArts', 'Health_Wellness', 'AppliedSciences', 'Spec
         ialNeeds', 'Literature_Writing', 'Mathematics', 'Literacy']
         Shape of matrix after one hot encodig_train (73196, 30)
         Shape of matrix after one hot encodig test (36052, 30)
```

```
In [33]: # we use count vectorizer to convert the values into one hot encoded features
         from sklearn.feature extraction.text import CountVectorizer
         vectorizer state = CountVectorizer( lowercase=False, binary=True)
         vectorizer state.fit(X train['school state'].values)
         print(vectorizer state.get feature names())
         school_state_one_hot_train = vectorizer_state.transform(X_train['school_state'
         1.values)
         # school state one hot cv = vectorizer state.transform(X cv['school state'].va
         Lues)
         school state one hot test = vectorizer state.transform(X test['school state'].
         values)
         print("Shape of matrix after one hot encodig train ", school state one hot trai
         n.shape)
         # print("Shape of matrix after one hot encodig_cv ",school_state_one_hot_cv.sh
         ape)
         print("Shape of matrix after one hot encodig test ", school state one hot test.
         shape)
         ['AK', 'AL', 'AR', 'AZ', 'CA', 'CO', 'CT', 'DC', 'DE', 'FL', 'GA', 'HI', 'I
         Ā', 'ID', 'IL', 'IN', 'KS', 'KY', 'LA', 'MA', 'MD', 'ME', 'MI', 'MN', 'MO',
         'MS', 'MT', 'NC', 'ND', 'NE', 'NH', 'NJ', 'NM', 'NV', 'NY', 'OH', 'OK', 'OR',
         'PA', 'RI', 'SC', 'SD', 'TN', 'TX', 'UT', 'VA', 'VT', 'WA', 'WI', 'WV', 'WY']
         Shape of matrix after one hot encoding train (73196, 51)
         Shape of matrix after one hot encodig_test (36052, 51)
In [34]:
         # we use count vectorizer to convert the values into one hot encoded features
         from sklearn.feature extraction.text import CountVectorizer
         vectorizer teacherprefix = CountVectorizer( lowercase=False, binary=True)
         vectorizer teacherprefix.fit(X train['teacher prefix'].values.astype('U'))
         print(vectorizer_teacherprefix.get_feature_names())
         #https://stackoverflow.com/a/39308809/8089731
         teacher prefix one hot train = vectorizer teacherprefix.transform(X train['tea
         cher prefix'].values.astype('U'))
         # teacher prefix one hot cv = vectorizer teacherprefix.transform(X cv['teacher
         prefix'].values.astype('U'))
         teacher prefix one hot test = vectorizer teacherprefix.transform(X test['teach
         er prefix'].values.astype('U'))
         print("Shape of matrix after one hot encodig train ", teacher prefix one hot tr
         ain.shape)
         # print("Shape of matrix after one hot encodig cv ",teacher prefix one hot cv.
         shape)
         print("Shape of matrix after one hot encodig_test ",teacher_prefix_one_hot_tes
         t[:5,:1)
         # print(X train['teacher prefix'].value counts())
         ['Dr', 'Mr', 'Mrs', 'Ms', 'Teacher']
         Shape of matrix after one hot encodig train (73196, 5)
         Shape of matrix after one hot encodig test
                                                       (0, 3)
           (1, 2)
                         1
           (2, 2)
                         1
           (3, 2)
                         1
           (4, 1)
                         1
```

```
In [35]: | print(project data['clean project grade category'].unique())
         ['GradesPreK-2' 'Grades6-8' 'Grades3-5' 'Grades9-12']
         # we use count vectorizer to convert the values into one hot encoded features
In [36]:
         from sklearn.feature extraction.text import CountVectorizer
         # https://stackoverflow.com/a/38161028/8089731
         pattern = "(?u) \setminus b[\setminus w-] + \setminus b"
         vectorizer projectgrade = CountVectorizer(token pattern=pattern, lowercase=Fal
         se, binary=True)
         vectorizer projectgrade.fit(X train['clean project grade category'].values)
         print(vectorizer projectgrade.get feature names())
         #https://stackoverflow.com/a/39308809/8089731
         project_grade_category_one_hot_train = vectorizer_projectgrade.transform(X_tra
         in['clean_project_grade_category'].values)
         # project grade category one hot cv = vectorizer projectgrade.transform(X cv
          ['clean project grade category'].values)
         project_grade_category_one_hot_test = vectorizer_projectgrade.transform(X_test
         ['clean project grade category'].values)
         print("Shape of matrix after one hot encodig train ",project grade category on
         e hot train.shape)
         # print("Shape of matrix after one hot encodig cv ",project grade category one
          hot cv.shape)
         print("Shape of matrix after one hot encodig_test ",project_grade_category_one
          hot test[:5,:])
         ['Grades3-5', 'Grades6-8', 'Grades9-12', 'GradesPreK-2']
         Shape of matrix after one hot encodig train (73196, 4)
         Shape of matrix after one hot encodig test
                                                         (0, 1)
           (1, 0)
                          1
           (2, 0)
                          1
           (3, 0)
                          1
           (4, 0)
                          1
```

### **Vectorizing Numerical features**

```
In [37]: # check this one: https://www.youtube.com/watch?v=0HOqOcln3Z4&t=530s
         # standardization sklearn: https://scikit-learn.org/stable/modules/generated/s
         klearn.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 32
              ... 399.
                         287.73 5.5 ].
         # Reshape your data either using array.reshape(-1, 1)
         price scalar = StandardScaler()
         price_scalar.fit(X_train['price'].values.reshape(-1,1)) # finding the mean and
         standard deviation of this data
         print(f"Mean : {price scalar.mean [0]}, Standard deviation : {np.sqrt(price sc
         alar.var_[0])}")
         # Now standardize the data with above maen and variance.
         price_standardized_train = price_scalar.transform(X_train['price'].values.resh
         ape(-1, 1)
         # price standardized cv = price scalar.transform(X cv['price'].values.reshape
         (-1, 1)
         price standardized test = price scalar.transform(X test['price'].values.reshap
         e(-1, 1)
         print(price standardized train.shape)
         # print(price standardized cv.shape)
         print(price standardized test.shape)
```

Mean: 297.77730654680585, Standard deviation: 363.9022509013626 (73196, 1) (36052, 1)

```
In [38]: # check this one: https://www.youtube.com/watch?v=0HOqOcLn3Z4&t=530s
         # standardization sklearn: https://scikit-learn.org/stable/modules/generated/s
         klearn.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 32
              ... 399.
                         287.73 5.5 ].
         # Reshape your data either using array.reshape(-1, 1)
         quantity scalar = StandardScaler()
         quantity_scalar.fit(X_train['quantity'].values.reshape(-1,1)) # finding the me
         an and standard deviation of this data
         # print(f"Mean : {quantity scalar.mean [0]}, Standard deviation : {np.sqrt(qua
         ntity_scalar.var_[0])}")
         # Now standardize the data with above maen and variance.
         quantity standardized train = quantity scalar.transform(X train['quantity'].va
         lues.reshape(-1, 1))
         # quantity standardized cv = quantity scalar.transform(X cv['quantity'].value
         s.reshape(-1, 1))
         quantity standardized test = quantity scalar.transform(X test['quantity'].valu
         es.reshape(-1, 1)
         print(quantity standardized train.shape)
         # print(quantity standardized cv.shape)
         print(quantity standardized test.shape)
```

C:\Users\LENOVO\Anaconda3\lib\site-packages\sklearn\utils\validation.py:595:
DataConversionWarning:

Data with input dtype int64 was converted to float64 by StandardScaler.

C:\Users\LENOVO\Anaconda3\lib\site-packages\sklearn\utils\validation.py:595:
DataConversionWarning:

Data with input dtype int64 was converted to float64 by StandardScaler.

C:\Users\LENOVO\Anaconda3\lib\site-packages\sklearn\utils\validation.py:595:
DataConversionWarning:

Data with input dtype int64 was converted to float64 by StandardScaler.

```
(73196, 1)
(36052, 1)
```

```
In [39]: # check this one: https://www.youtube.com/watch?v=0HOqOcln3Z4&t=530s
         # standardization sklearn: https://scikit-learn.org/stable/modules/generated/s
         klearn.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 32
              ... 399.
                         287.73 5.5 ].
         # Reshape your data either using array.reshape(-1, 1)
         teacher_number_of_previously_posted_projects_scalar = StandardScaler()
         teacher_number_of_previously_posted_projects_scalar.fit(X_train['teacher_numbe
         r_of_previously_posted_projects'].values.reshape(-1,1)) # finding the mean and
         standard deviation of this data
         # print(f"Mean : {teacher_number_of_previously_posted_projects_scalar.mean_
         [0]}, Standard deviation : {np.sqrt(teacher number of previously posted projec
         ts_scalar.var_[0])}")
         # Now standardize the data with above maen and variance.
         teacher number of previously posted projects standardized train = teacher numb
         er_of_previously_posted_projects_scalar.transform(X_train['teacher_number_of_p
         reviously posted projects'].values.reshape(-1, 1))
         # teacher_number_of_previously_posted_projects_standardized_cv = teacher_numbe
         r of previously posted projects scalar.transform(X cv['teacher number of previ
         ously posted projects'].values.reshape(-1, 1))
         teacher number of previously posted projects standardized test = teacher numbe
         r_of_previously_posted_projects_scalar.transform(X_test['teacher_number_of_pre
         viously posted projects'].values.reshape(-1, 1))
         print(teacher number of previously posted projects standardized train.shape)
         # print(teacher_number_of_previously_posted_projects_standardized_cv.shape)
         print(teacher_number_of_previously_posted_projects_standardized_test.shape)
         C:\Users\LENOVO\Anaconda3\lib\site-packages\sklearn\utils\validation.py:595:
```

C:\Users\LENOVO\Anaconda3\lib\site-packages\sklearn\utils\validation.py:595:
DataConversionWarning:

Data with input dtype int64 was converted to float64 by StandardScaler.

C:\Users\LENOVO\Anaconda3\lib\site-packages\sklearn\utils\validation.py:595:
DataConversionWarning:

Data with input dtype int64 was converted to float64 by StandardScaler.

C:\Users\LENOVO\Anaconda3\lib\site-packages\sklearn\utils\validation.py:595:
DataConversionWarning:

Data with input dtype int64 was converted to float64 by StandardScaler.

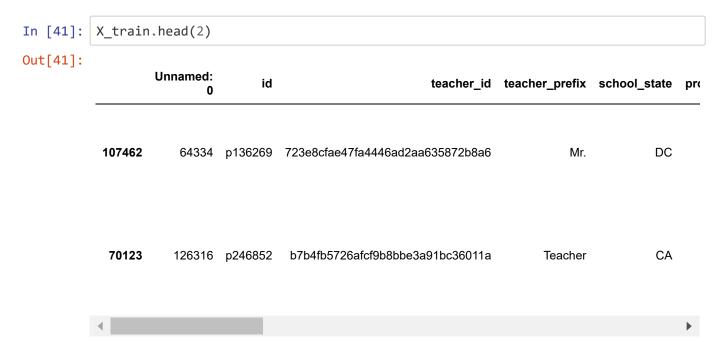
```
(73196, 1)
(36052, 1)
```

```
In [ ]:
```

```
In [40]: # please write all the code with proper documentation, and proper titles for e
    ach subsection
    # go through documentations and blogs before you start coding
    # first figure out what to do, and then think about how to do.
    # reading and understanding error messages will be very much helpfull in debug
    ging your code
    # make sure you featurize train and test data separatly

# when you plot any graph make sure you use
    # a. Title, that describes your plot, this will be very helpful to the rea
    der
        # b. Legends if needed
        # c. X-axis label
        # d. Y-axis label
```

### 2.3 Make Data Model Ready: encoding eassay, and project\_title



Bag of Words(BOW) on project\_TEXT/ESSAYS (Train,Cv,Test)

```
In [42]: # We are considering only the words which appeared in at least 10 documents(ro
    ws or projects).
    vectorizer_bow_essays = CountVectorizer(min_df=10,max_features=5000,ngram_rang
    e=(1,2))
    vectorizer_bow_essays.fit(X_train['preprocessed_essays'])

    text_bow_train = vectorizer_bow_essays.transform(X_train['preprocessed_essays'])

    # text_bow_cv = vectorizer_bow_essays.transform(X_cv['preprocessed_essays'])
    text_bow_test = vectorizer_bow_essays.transform(X_test['preprocessed_essays'])
    print("Shape of matrix after BOW_text_train ",text_bow_train.shape)

# print("Shape of matrix after BOW_text_cv ",text_bow_cv.shape)
    print("Shape of matrix after BOW_text_test ",text_bow_test.shape)

Shape of matrix after BOW_text_train (73196, 5000)
```

Shape of matrix after BOW\_text\_test (36052, 5000)

### Bag of Words(BOW) on project\_title (Train,Cv,Test)

```
In [43]: # We are considering only the words which appeared in at least 10 documents(ro
    ws or projects).
    vectorizer_bow_titles = CountVectorizer(min_df=10)
    vectorizer_bow_titles.fit(X_train['preprocessed_titles'])

    title_bow_train = vectorizer_bow_titles.transform(X_train['preprocessed_title
    s'])
    # title_bow_cv = vectorizer_bow_titles.transform(X_cv['preprocessed_titles'])
    title_bow_test = vectorizer_bow_titles.transform(X_test['preprocessed_titles'])
    print("Shape of matrix after BOW_title_train ",title_bow_train.shape)
    # print("Shape of matrix after BOW_title_cv ",title_bow_cv.shape)
    print("Shape of matrix after BOW_title_test ",title_bow_test.shape)
Shape of matrix after BOW_title_train (73196, 2621)
Shape of matrix after BOW_title_test (36052, 2621)
```

### TFIDF Vectorizer on project\_TEXT/ESSAYS (Train,Cv,Test)

```
In [44]: from sklearn.feature_extraction.text import TfidfVectorizer
    vectorizer_tfidf_essays = TfidfVectorizer(min_df=10,max_features=5000,ngram_ra
    nge=(1,2))
    vectorizer_tfidf_essays.fit(X_train['preprocessed_essays'])

    text_tfidf_train = vectorizer_tfidf_essays.transform(X_train['preprocessed_essays'])

# text_tfidf_cv = vectorizer_tfidf_essays.transform(X_cv['preprocessed_essays'])

text_tfidf_test = vectorizer_tfidf_essays.transform(X_test['preprocessed_essays'])

print("Shape of matrix after tfidf_text_train ",text_tfidf_train.shape)

# print("Shape of matrix after tfidf_text_cv ",text_tfidf_cv.shape)

print("Shape of matrix after tfidf_text_test ",text_tfidf_test.shape)

Shape of matrix after tfidf_text_train (73196, 5000)
Shape of matrix after tfidf text test (36052, 5000)
```

# TFIDF Vectorizer on project title (Train, Cv, Test)

```
In [45]: from sklearn.feature extraction.text import TfidfVectorizer
         vectorizer tfidf title = TfidfVectorizer(min df=10)
         vectorizer_tfidf_title.fit(X_train['preprocessed_titles'])
         title tfidf train = vectorizer tfidf title.transform(X train['preprocessed tit
         les'])
         # title tfidf cv = vectorizer tfidf title.transform(X cv['preprocessed title
         title_tfidf_test = vectorizer_tfidf_title.transform(X_test['preprocessed_title
         s'])
         print("Shape of matrix after tfidf title train ",title tfidf train.shape)
         # print("Shape of matrix after tfidf title cv ",title tfidf cv.shape)
         print("Shape of matrix after tfidf_title_test ",title_tfidf_test.shape)
         Shape of matrix after tfidf_title_train (73196, 2621)
         Shape of matrix after tfidf title test (36052, 2621)
In [46]: # stronging variables into pickle files python: http://www.jessicayung.com/how
         -to-use-pickle-to-save-and-load-variables-in-python/
         # make sure you have the glove vectors file
         with open('../glove vectors', 'rb') as f:
             model = pickle.load(f)
             glove words = set(model.keys())
```

## Avg W2V on TEXT/ESSAYS(Train,cv,test)

```
In [47]: # average Word2Vec
         # compute average word2vec for each review.
         avg w2v essays vectors train = []; # the avg-w2v for each sentence/review is s
         tored in this list
         for sentence in tqdm1(X train['preprocessed essays']): # for each review/sente
         nce
             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 essays vectors train.append(vector)
         # avg w2v essays vectors cv = []; # the avg-w2v for each sentence/review is st
         ored in this list
         # for sentence in tqdm1(X cv['preprocessed essays']): # for each review/senten
         #
               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
         #
               avq w2v essays vectors cv.append(vector)
         avg w2v essays vectors test = []; # the avg-w2v for each sentence/review is st
         ored in this list
         for sentence in tqdm1(X test['preprocessed essays']): # for each review/senten
         ce
             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 essays vectors test.append(vector)
         print(len(avg w2v essays vectors train))
         # print(len(avg w2v essays vectors cv))
         print(len(avg_w2v_essays_vectors_test))
         print(len(avg w2v essays vectors train[0]))
         # print(len(avg w2v essays vectors cv[0]))
         print(len(avg w2v essays vectors test[0]))
```

# Avg W2V on TITLES(Train,cv,test)

```
In [48]: # average Word2Vec
         # compute average word2vec for each review.
         avg w2v titles vectors train = []; # the avg-w2v for each sentence/review is s
         tored in this list
         for sentence in tqdm1(X train['preprocessed titles']): # for each review/sente
         nce
             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 titles vectors train.append(vector)
         # avg w2v titles vectors cv = []; # the avg-w2v for each sentence/review is st
         ored in this list
         # for sentence in tqdm1(X cv['preprocessed titles']): # for each review/senten
         #
               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 titles vectors cv.append(vector)
         avg w2v titles vectors test = []; # the avg-w2v for each sentence/review is st
         ored in this list
         for sentence in tqdm1(X test['preprocessed titles']): # for each review/senten
         ce
             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 titles vectors test.append(vector)
         print(len(avg w2v titles vectors train))
         # print(len(avg w2v titles vectors cv))
         print(len(avg_w2v_titles_vectors_test))
         print(len(avg w2v titles vectors train[0]))
         # print(len(avg w2v titles vectors cv[0]))
         print(len(avg w2v titles vectors test[0]))
```

# TFIDF weighted W2V on TEXT/ESSAYS(Train,cv,test)

```
In [49]: \# 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())
         # average Word2Vec
         # compute average word2vec for each review.
         tfidf_w2v_essays_vectors_train = []; # the avg-w2v for each sentence/review is
         stored in this list
         for sentence in tqdm1(X train['preprocessed essays']): # for each review/sente
             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/revie
             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 v
         alue((sentence.count(word)/len(sentence.split())))
                     tf idf = dictionary[word]*(sentence.count(word)/len(sentence.split
         ())) # getting the tfidf 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_essays_vectors_train.append(vector)
         # # average Word2Vec
         # # compute average word2vec for each review.
         # tfidf_w2v_essays_vectors_cv = []; # the avg-w2v for each sentence/review is
          stored in this list
         # for sentence in tqdm1(X cv['preprocessed essays']): # for each review/senten
         ce
         #
               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/rev
         iew
               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.spl
         it())) # getting the tfidf 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_essays_vectors_cv.append(vector)
         # average Word2Vec
```

```
# compute average word2vec for each review.
tfidf_w2v_essays_vectors_test = []; # the avg-w2v for each sentence/review is
stored in this list
for sentence in tqdm1(X test['preprocessed essays']): # for each review/senten
    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/revie
    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 v
alue((sentence.count(word)/len(sentence.split())))
            tf idf = dictionary[word]*(sentence.count(word)/len(sentence.split
())) # getting the tfidf 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_essays_vectors_test.append(vector)
print(len(tfidf w2v essays vectors train))
# print(len(tfidf w2v essays vectors cv))
print(len(tfidf w2v essays vectors test))
print(len(tfidf_w2v_essays_vectors_train[0]))
# print(len(tfidf w2v essays vectors cv[0]))
print(len(tfidf w2v essays vectors test[0]))
```

### TFIDF weighted W2V on TITLES(Train,cv,test)

```
In [50]: \# S = ["abc def pqr", "def def def abc", "pqr pqr def"]
         tfidf model = TfidfVectorizer()
         tfidf model.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 model.get feature names(), list(tfidf model.idf
         )))
         tfidf words = set(tfidf model.get feature names())
         # average Word2Vec
         # compute average word2vec for each review.
         tfidf_w2v_titles_vectors_train = []; # the avg-w2v for each sentence/review is
         stored in this list
         for sentence in tqdm1(X train['preprocessed titles']): # for each review/sente
             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/revie
             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 v
         alue((sentence.count(word)/len(sentence.split())))
                     tf idf = dictionary[word]*(sentence.count(word)/len(sentence.split
         ())) # getting the tfidf 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_titles_vectors_train.append(vector)
         # # average Word2Vec
         # # compute average word2vec for each review.
         # tfidf_w2v_titles_vectors_cv = []; # the avg-w2v for each sentence/review is
          stored in this list
         # for sentence in tqdm1(X cv['preprocessed titles']): # for each review/senten
         ce
         #
               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/rev
         iew
               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.spl
         it())) # getting the tfidf 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_titles_vectors_cv.append(vector)
         # average Word2Vec
```

```
# compute average word2vec for each review.
tfidf_w2v_titles_vectors_test = []; # the avg-w2v for each sentence/review is
stored in this list
for sentence in tqdm1(X test['preprocessed titles']): # for each review/senten
   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/revie
W
   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 v
alue((sentence.count(word)/len(sentence.split())))
            tf idf = dictionary[word]*(sentence.count(word)/len(sentence.split
())) # getting the tfidf 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_titles_vectors_test.append(vector)
print(len(tfidf w2v titles vectors train))
# print(len(tfidf w2v titles vectors cv))
print(len(tfidf w2v titles vectors test))
print(len(tfidf_w2v_titles_vectors_train[0]))
# print(len(tfidf w2v titles vectors cv[0]))
print(len(tfidf w2v titles vectors test[0]))
```

```
In [1]: import dill
    # dill.dump_session('notebook_env.db')
    dill.load_session('notebook_env.db')
```

C:\Users\LENOVO\Anaconda3\lib\site-packages\smart\_open\ssh.py:34: UserWarnin
g: paramiko missing, opening SSH/SCP/SFTP paths will be disabled. `pip insta
ll paramiko` to suppress
 warnings.warn('paramiko missing, opening SSH/SCP/SFTP paths will be disable
d. `pip install paramiko` to suppress')
C:\Users\LENOVO\Anaconda3\lib\site-packages\gensim\utils.py:1197: UserWarnin
g: detected Windows; aliasing chunkize to chunkize\_serial
 warnings.warn("detected Windows; aliasing chunkize to chunkize\_serial")

```
In [2]: project data.columns
Out[2]: Index(['Unnamed: 0', 'id', 'teacher_id', 'teacher_prefix', 'school_state',
                 'project_submitted_datetime', 'project_title',
                 'project_resource_summary',
                 'teacher_number_of_previously_posted_projects', 'project_is_approved',
                 'clean_categories', 'clean_subcategories', 'essay', 'price', 'quantit
         у',
                'digits in summary', 'clean project grade category',
                 'preprocessed_essays', 'preprocessed_titles'],
               dtype='object')
In [53]: # please write all the code with proper documentation, and proper titles for e
         ach subsection
         # go through documentations and blogs before you start coding
         # first figure out what to do, and then think about how to do.
         # reading and understanding error messages will be very much helpfull in debug
         ging your code
         # make sure you featurize train and test data separatly
         # when you plot any graph make sure you use
             # a. Title, that describes your plot, this will be very helpful to the rea
         der
             # b. Legends if needed
             # c. X-axis label
             # d. Y-axis label
```

# 2.4 Appling Logistic Regression on different kind of featurization as mentioned in the instructions

Apply Logistic Regression 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 instrucations

## 2.4.1 Applying Logistic Regression on BOW, SET 1

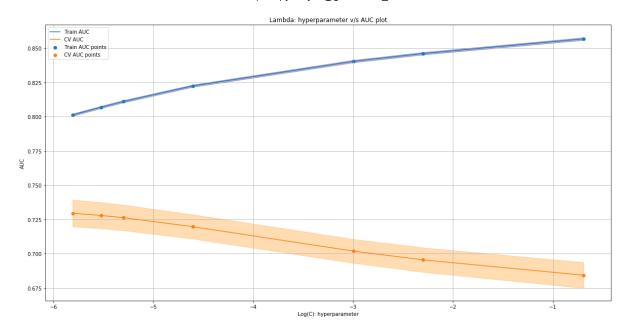
```
In [3]: # merge two sparse matrices: https://stackoverflow.com/a/19710648/4084039
        from scipy.sparse import hstack
        X tr = hstack((categories one hot train, sub categories one hot train, school st
        ate one hot train, teacher prefix one hot train
                        ,project_grade_category_one_hot_train,price_standardized_train,
        quantity_standardized_train
                        ,teacher number of previously posted projects standardized trai
        n,text bow train,title bow train)).tocsr()
        # X cr = hstack((categories one hot cv,sub categories one hot cv,school state
        one_hot_cv,teacher_prefix_one_hot_cv
                          ,project grade category one hot cv,price standardized cv,quan
        tity standardized cv
                          ,teacher_number_of_previously_posted_projects_standardized_c
        v,text bow cv,title bow cv)).tocsr()
        X te = hstack((categories one hot test, sub categories one hot test, school stat
        e_one_hot_test,teacher_prefix_one_hot_test
                        ,project grade category one hot test,price standardized test,qu
        antity standardized test
                        ,teacher_number_of_previously_posted_projects_standardized_test
        ,text_bow_test,title_bow_test)).tocsr()
        print("Final Data matrix on BOW")
        print(X tr.shape, y train.shape)
        # print(X_cr.shape, y_cv.shape)
        print(X_te.shape, y_test.shape)
        print("="*100)
        Final Data matrix on BOW
```

```
Final Data matrix on BOW (73196, 7723) (73196,) (36052, 7723) (36052,)
```

\_\_\_\_\_

#### 1.1 Method 1: GridSearchCV

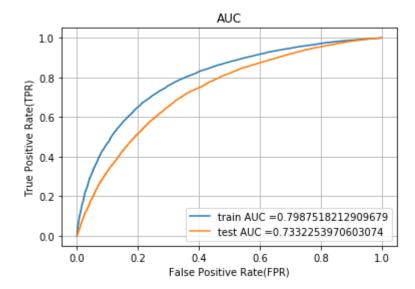
```
In [8]: # https://machinelearningmastery.com/how-to-fix-futurewarning-messages-in-scik
        it-learn/
        from warnings import simplefilter
        # ignore all future warnings
        simplefilter(action='ignore', category=FutureWarning)
        from sklearn.model selection import GridSearchCV
        import matplotlib.pvplot as plt
        from sklearn.linear model import LogisticRegression
        from sklearn.metrics import roc auc score
        C = [0.003, 0.004, 0.005, 0.01, 0.05, 0.1, 0.5]
        logi = LogisticRegression(class weight='balanced')
        parameters = {'C':C}
        clf = GridSearchCV(logi, parameters, cv= 10, scoring='roc auc')
        clf.fit(X tr, y train)
        train_auc= clf.cv_results_['mean_train_score']
        train_auc_std= clf.cv_results_['std_train_score']
        cv auc = clf.cv results ['mean test score']
        cv auc std= clf.cv results ['std test score']
        plt.figure(figsize=(20,10))
        plt.plot(np.log(parameters['C']), train_auc, label='Train AUC')
        # https://stackoverflow.com/a/48803361/4084039
        plt.gca().fill between(np.log(parameters['C']),train auc - train auc std,train
         auc + train auc std,alpha=0.3,color='darkblue')
        plt.plot(np.log(parameters['C']), cv_auc, label='CV AUC')
        # https://stackoverflow.com/a/48803361/4084039
        plt.gca().fill_between(np.log(parameters['C']),cv_auc - cv_auc_std,cv_auc + cv
         _auc_std,alpha=0.3,color='darkorange')
        plt.scatter(np.log(parameters['C']), train auc, label='Train AUC points')
        plt.scatter(np.log(parameters['C']), cv auc, label='CV AUC points')
        plt.legend()
        plt.xlabel("Log(C): hyperparameter")
        plt.ylabel("AUC")
        plt.title("Lambda: hyperparameter v/s AUC plot")
        plt.grid()
```



```
In [5]: def batch_predict(clf, data):
    # roc_auc_score(y_true, y_score) the 2nd parameter should be probability e
    stimates of the positive class
    # not the predicted outputs

y_data_pred = []
    tr_loop = data.shape[0] - data.shape[0]%1000
    # consider you X_tr shape is 49041, then your cr_loop will be 49041 - 4904
1%1000 = 49000
    # in this for loop we will iterate unti the last 1000 multiplier
    for i in range(0, tr_loop, 1000):
        y_data_pred.extend(clf.predict_proba(data[i:i+1000])[:,1])
    # we will be predicting for the last data points
    y_data_pred.extend(clf.predict_proba(data[tr_loop:])[:,1])
    return y_data_pred
```

```
# https://scikit-learn.org/stable/modules/generated/sklearn.metrics.roc_curve.
In [11]:
         html#sklearn.metrics.roc curve
         from sklearn.metrics import roc curve, auc
         logi = LogisticRegression(C=0.003,class_weight='balanced')
         logi.fit(X_tr[:,:], y_train[:])
         # roc auc score(y true, y score) the 2nd parameter should be probability estim
         ates of the positive class
         # not the predicted outputs
         y_train_pred = batch_predict(logi, X_tr[:,:])
         y_test_pred = batch_predict(logi, X_te[:])
         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_tp
         r)))
         plt.plot(test fpr, test tpr, label="test AUC ="+str(auc(test fpr, test tpr)))
         plt.legend()
         plt.xlabel("False Positive Rate(FPR)")
         plt.ylabel("True Positive Rate(TPR)")
         plt.title("AUC")
         plt.grid()
         plt.show()
```



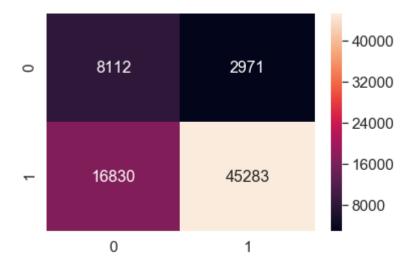
```
In [13]: print("="*100)
    from sklearn.metrics import confusion_matrix
    print("Train confusion matrix")
    print(confusion_matrix(y_train[:], predict(y_train_pred, tr_thresholds, train_fpr, train_tpr)))
    print("Test confusion matrix")
    print(confusion_matrix(y_test[:], predict(y_test_pred, tr_thresholds, test_fpr, test_tpr)))
```

```
Train confusion matrix
the maximum value of tpr*(1-fpr) 0.5336091821331846 for threshold 0.49
[[ 8112 2971]
  [16830 45283]]
Test confusion matrix
the maximum value of tpr*(1-fpr) 0.4623413482199421 for threshold 0.572
[[ 4188 1271]
  [13277 17316]]
```

```
In [14]: conf_matr_df_train = pd.DataFrame(confusion_matrix(y_train[:], predict(y_train_pred, tr_thresholds, train_fpr, train_tpr)))
    sns.set(font_scale=1.4)#for label size
    sns.heatmap(conf_matr_df_train, annot=True,annot_kws={"size": 16}, fmt='g')
```

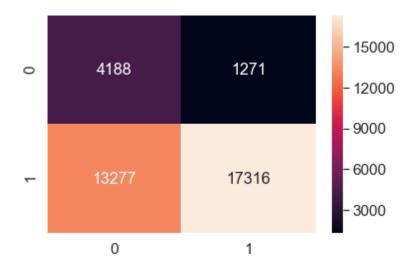
the maximum value of tpr\*(1-fpr) 0.5336091821331846 for threshold 0.49

Out[14]: <matplotlib.axes. subplots.AxesSubplot at 0x2d1552e3898>



the maximum value of tpr\*(1-fpr) 0.4623413482199421 for threshold 0.572

Out[15]: <matplotlib.axes. subplots.AxesSubplot at 0x2d155944ac8>



### 2.4.2 Applying Logistic Regression on TFIDF, SET 2

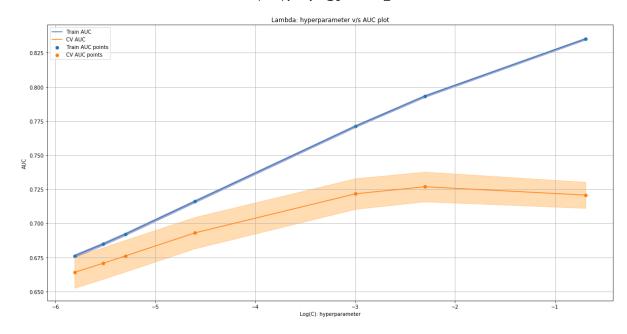
```
In [4]: # merge two sparse matrices: https://stackoverflow.com/a/19710648/4084039
        from scipy.sparse import hstack
        X tr = hstack((categories one hot train, sub categories one hot train, school st
        ate one hot train, teacher prefix one hot train
                        ,project grade category one hot train, price standardized train,
        quantity_standardized_train
                        ,teacher number of previously posted projects standardized trai
        n,text tfidf train,title tfidf train)).tocsr()
        # X cr = hstack((categories one hot cv,sub categories one hot cv,school state
        one_hot_cv,teacher_prefix_one_hot_cv
                          ,project grade category one hot cv,price standardized cv,quan
        tity standardized cv
                          ,teacher_number_of_previously_posted_projects_standardized_c
        v,text tfidf cv,title tfidf cv)).tocsr()
        X te = hstack((categories one hot test, sub categories one hot test, school stat
        e_one_hot_test,teacher_prefix_one_hot_test
                        ,project grade category one hot test,price standardized test,qu
        antity standardized test
                        ,teacher_number_of_previously_posted_projects_standardized_test
        ,text tfidf test,title tfidf test)).tocsr()
        print("Final Data matrix on TFIDF")
        print(X tr.shape, y train.shape)
        # print(X_cr.shape, y_cv.shape)
        print(X_te.shape, y_test.shape)
        print("="*100)
```

Final Data matrix on TFIDF (73196, 7723) (73196,) (36052, 7723) (36052,)

\_\_\_\_\_\_

#### 1.1 Method 1: GridSearchCV

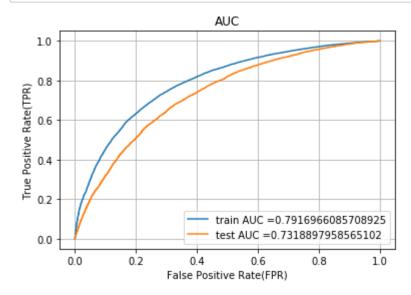
```
In [5]: # https://machinelearningmastery.com/how-to-fix-futurewarning-messages-in-scik
        it-learn/
        from warnings import simplefilter
        # ignore all future warnings
        simplefilter(action='ignore', category=FutureWarning)
        from sklearn.model selection import GridSearchCV
        import matplotlib.pvplot as plt
        from sklearn.linear model import LogisticRegression
        from sklearn.metrics import roc auc score
        C = [0.003, 0.004, 0.005, 0.01, 0.05, 0.1, 0.5]
        logi = LogisticRegression(class weight='balanced')
        parameters = {'C':C}
        clf = GridSearchCV(logi, parameters, cv= 10, scoring='roc auc')
        clf.fit(X tr, y train)
        train auc= clf.cv results ['mean train score']
        train_auc_std= clf.cv_results_['std_train_score']
        cv auc = clf.cv results ['mean test score']
        cv auc std= clf.cv results ['std test score']
        plt.figure(figsize=(20,10))
        plt.plot(np.log(parameters['C']), train_auc, label='Train AUC')
        # https://stackoverflow.com/a/48803361/4084039
        plt.gca().fill between(np.log(parameters['C']),train auc - train auc std,train
         auc + train auc std,alpha=0.3,color='darkblue')
        plt.plot(np.log(parameters['C']), cv_auc, label='CV AUC')
        # https://stackoverflow.com/a/48803361/4084039
        plt.gca().fill_between(np.log(parameters['C']),cv_auc - cv_auc_std,cv_auc + cv
         _auc_std,alpha=0.3,color='darkorange')
        plt.scatter(np.log(parameters['C']), train auc, label='Train AUC points')
        plt.scatter(np.log(parameters['C']), cv auc, label='CV AUC points')
        plt.legend()
        plt.xlabel("Log(C): hyperparameter")
        plt.ylabel("AUC")
        plt.title("Lambda: hyperparameter v/s AUC plot")
        plt.grid()
```



```
In [6]: def batch_predict(clf, data):
    # roc_auc_score(y_true, y_score) the 2nd parameter should be probability e
    stimates of the positive class
    # not the predicted outputs

y_data_pred = []
    tr_loop = data.shape[0] - data.shape[0]%1000
    # consider you X_tr shape is 49041, then your cr_loop will be 49041 - 4904
1%1000 = 49000
    # in this for loop we will iterate unti the last 1000 multiplier
    for i in range(0, tr_loop, 1000):
        y_data_pred.extend(clf.predict_proba(data[i:i+1000])[:,1])
    # we will be predicting for the last data points
    y_data_pred.extend(clf.predict_proba(data[tr_loop:])[:,1])
    return y_data_pred
```

```
# https://scikit-learn.org/stable/modules/generated/sklearn.metrics.roc_curve.
In [11]:
         html#sklearn.metrics.roc curve
         from sklearn.metrics import roc curve, auc
         logi = LogisticRegression(C=0.1,class_weight='balanced')
         logi.fit(X_tr[:,:], y_train[:])
         # roc auc score(y true, y score) the 2nd parameter should be probability estim
         ates of the positive class
         # not the predicted outputs
         y_train_pred = batch_predict(logi, X_tr[:,:])
         y_test_pred = batch_predict(logi, X_te[:])
         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_tp
         r)))
         plt.plot(test fpr, test tpr, label="test AUC ="+str(auc(test fpr, test tpr)))
         plt.legend()
         plt.xlabel("False Positive Rate(FPR)")
         plt.ylabel("True Positive Rate(TPR)")
         plt.title("AUC")
         plt.grid()
         plt.show()
```



```
Train confusion matrix
the maximum value of tpr*(1-fpr) 0.5230910980980099 for threshold 0.493
[[ 7966 3117]
       [16909 45204]]
Test confusion matrix
the maximum value of tpr*(1-fpr) 0.4563546597548524 for threshold 0.571
[[ 4262 1197]
       [14052 16541]]
```

```
In [14]: conf_matr_df_train = pd.DataFrame(confusion_matrix(y_train[:], predict(y_train_pred, tr_thresholds, train_fpr, train_tpr)))
    sns.set(font_scale=1.4)#for label size
    sns.heatmap(conf_matr_df_train, annot=True,annot_kws={"size": 16}, fmt='g')
```

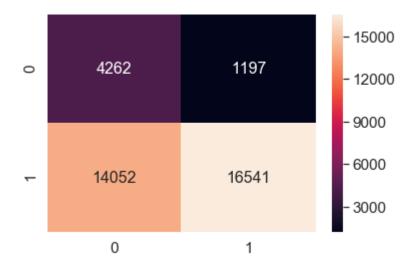
the maximum value of tpr\*(1-fpr) 0.5230910980980099 for threshold 0.493

Out[14]: <matplotlib.axes. subplots.AxesSubplot at 0x2a2d2a3e0b8>



the maximum value of tpr\*(1-fpr) 0.4563546597548524 for threshold 0.571

Out[15]: <matplotlib.axes. subplots.AxesSubplot at 0x2a2d30d6a20>

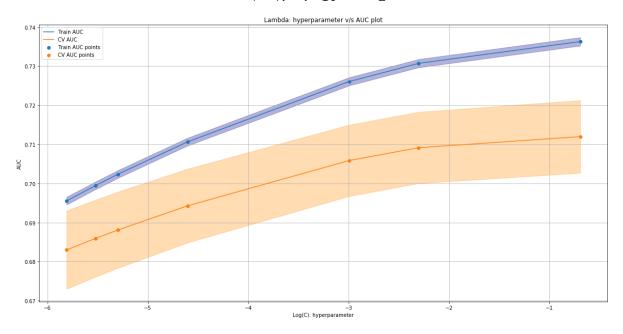


### 2.4.3 Applying Logistic Regression on AVG W2V, SET 3

```
In [2]: # merge two sparse matrices: https://stackoverflow.com/a/19710648/4084039
        from scipy.sparse import hstack
        X tr = hstack((categories one hot train, sub categories one hot train, school st
        ate one hot train, teacher prefix one hot train
                        ,project_grade_category_one_hot_train,price_standardized_train,
        quantity_standardized_train
                        ,teacher_number_of_previously_posted_projects_standardized_trai
        n,avg w2v essays vectors train
                        ,avg w2v titles vectors train)).tocsr()
        # X_cr = hstack((categories_one_hot_cv,sub_categories_one_hot_cv,school_state_
        one hot cv, teacher prefix one hot cv
                          ,project_grade_category_one_hot_cv,price_standardized_cv,quan
        tity_standardized cv
                          ,teacher number of previously posted projects standardized c
        v,avg w2v essays vectors cv
                          ,avg_w2v_titles_vectors_cv)).tocsr()
        X te = hstack((categories one hot test, sub categories one hot test, school stat
        e_one_hot_test,teacher_prefix_one_hot_test
                        ,project_grade_category_one_hot_test,price_standardized_test,qu
        antity standardized test
                        ,teacher number of previously posted projects standardized test
         ,avg_w2v_essays_vectors_test
                        ,avg w2v titles vectors test)).tocsr()
        print("Final Data matrix on AVGW2V")
        print(X tr.shape, y train.shape)
        # print(X cr.shape, y cv.shape)
        print(X_te.shape, y_test.shape)
        print("="*100)
        Final Data matrix on AVGW2V
        (73196, 702) (73196,)
        (36052, 702) (36052,)
```

### 1.1 Method 1: GridSearchCV

```
In [3]: # https://machinelearningmastery.com/how-to-fix-futurewarning-messages-in-scik
        it-learn/
        from warnings import simplefilter
        # ignore all future warnings
        simplefilter(action='ignore', category=FutureWarning)
        from sklearn.model selection import GridSearchCV
        import matplotlib.pvplot as plt
        from sklearn.linear model import LogisticRegression
        from sklearn.metrics import roc auc score
        C = [0.003, 0.004, 0.005, 0.01, 0.05, 0.1, 0.5]
        logi = LogisticRegression(class weight='balanced')
        parameters = {'C':C}
        clf = GridSearchCV(logi, parameters, cv= 10, scoring='roc auc')
        clf.fit(X tr, y train)
        train auc= clf.cv results ['mean train score']
        train_auc_std= clf.cv_results_['std_train_score']
        cv auc = clf.cv results ['mean test score']
        cv auc std= clf.cv results ['std test score']
        plt.figure(figsize=(20,10))
        plt.plot(np.log(parameters['C']), train_auc, label='Train AUC')
        # https://stackoverflow.com/a/48803361/4084039
        plt.gca().fill between(np.log(parameters['C']),train auc - train auc std,train
         auc + train auc std,alpha=0.3,color='darkblue')
        plt.plot(np.log(parameters['C']), cv_auc, label='CV AUC')
        # https://stackoverflow.com/a/48803361/4084039
        plt.gca().fill_between(np.log(parameters['C']),cv_auc - cv_auc_std,cv_auc + cv
         _auc_std,alpha=0.3,color='darkorange')
        plt.scatter(np.log(parameters['C']), train auc, label='Train AUC points')
        plt.scatter(np.log(parameters['C']), cv auc, label='CV AUC points')
        plt.legend()
        plt.xlabel("Log(C): hyperparameter")
        plt.ylabel("AUC")
        plt.title("Lambda: hyperparameter v/s AUC plot")
        plt.grid()
```

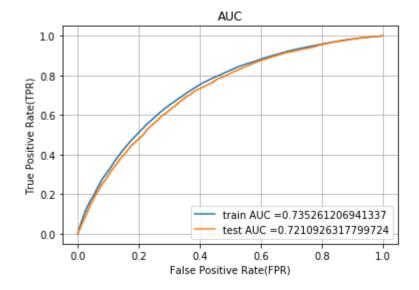


```
In [4]: def batch_predict(clf, data):
    # roc_auc_score(y_true, y_score) the 2nd parameter should be probability e
    stimates of the positive class
    # not the predicted outputs

    y_data_pred = []
    tr_loop = data.shape[0] - data.shape[0]%1000
    # consider you X_tr shape is 49041, then your cr_loop will be 49041 - 4904

1%1000 = 49000
    # in this for loop we will iterate unti the last 1000 multiplier
    for i in range(0, tr_loop, 1000):
        y_data_pred.extend(clf.predict_proba(data[i:i+1000])[:,1])
    # we will be predicting for the last data points
    y_data_pred.extend(clf.predict_proba(data[tr_loop:])[:,1])
    return y_data_pred
```

```
# https://scikit-learn.org/stable/modules/generated/sklearn.metrics.roc_curve.
In [7]:
        html#sklearn.metrics.roc curve
        from sklearn.metrics import roc curve, auc
        logi = LogisticRegression(C=0.5,class_weight='balanced')
        logi.fit(X_tr[:,:], y_train[:])
        # roc auc score(y true, y score) the 2nd parameter should be probability estim
        ates of the positive class
        # not the predicted outputs
        y_train_pred = batch_predict(logi, X_tr[:,:])
        y_test_pred = batch_predict(logi, X_te[:])
        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_tp
        r)))
        plt.plot(test fpr, test tpr, label="test AUC ="+str(auc(test fpr, test tpr)))
        plt.legend()
        plt.xlabel("False Positive Rate(FPR)")
        plt.ylabel("True Positive Rate(TPR)")
        plt.title("AUC")
        plt.grid()
        plt.show()
```



```
In [9]: print("="*100)
    from sklearn.metrics import confusion_matrix
    print("Train confusion matrix")
    print(confusion_matrix(y_train[:], predict(y_train_pred, tr_thresholds, train_fpr, train_tpr)))
    print("Test confusion matrix")
    print(confusion_matrix(y_test[:], predict(y_test_pred, tr_thresholds, test_fpr, test_tpr)))
```

-----

```
Train confusion matrix
the maximum value of tpr*(1-fpr) 0.45839193499902187 for threshold 0.491
[[ 7417 3666]
  [19568 42545]]
Test confusion matrix
the maximum value of tpr*(1-fpr) 0.44642125491282 for threshold 0.602
[[ 4451 1008]
  [16576 14017]]
```

```
In [10]: conf_matr_df_train = pd.DataFrame(confusion_matrix(y_train[:], predict(y_train_pred, tr_thresholds, train_fpr, train_tpr)))
    sns.set(font_scale=1.4)#for label size
    sns.heatmap(conf_matr_df_train, annot=True,annot_kws={"size": 16}, fmt='g')
```

the maximum value of tpr\*(1-fpr) 0.45839193499902187 for threshold 0.491

Out[10]: <matplotlib.axes. subplots.AxesSubplot at 0x2c2578c1908>



```
In [11]: conf_matr_df_test = pd.DataFrame(confusion_matrix(y_test[:], predict(y_test_p red, tr_thresholds, test_fpr, test_tpr)))
    sns.set(font_scale=1.4)#for label size
    sns.heatmap(conf_matr_df_test, annot=True,annot_kws={"size": 16}, fmt='g')
```

the maximum value of tpr\*(1-fpr) 0.44642125491282 for threshold 0.602

Out[11]: <matplotlib.axes.\_subplots.AxesSubplot at 0x2c257c9dac8>



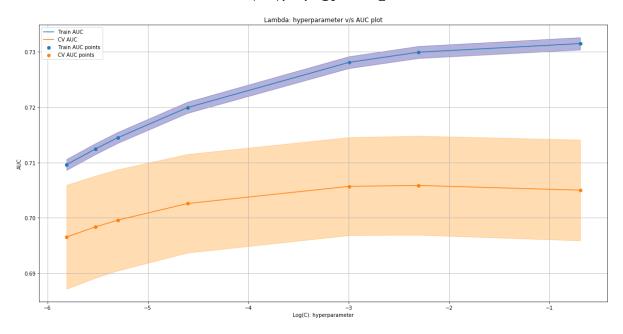
## 2.4.4 Applying Logistic Regression on TFIDF W2V, SET 4

```
In [2]: # merge two sparse matrices: https://stackoverflow.com/a/19710648/4084039
        from scipy.sparse import hstack
        X tr = hstack((categories one hot train, sub categories one hot train, school st
        ate one hot train, teacher prefix one hot train
                       ,project grade category one hot train, price standardized train,
        quantity_standardized_train
                       ,teacher number of previously posted projects standardized trai
        n,tfidf_w2v_essays_vectors_train
                       ,tfidf w2v titles vectors train)).tocsr()
        # X_cr = hstack((categories_one_hot_cv,sub_categories_one_hot_cv,school_state_
        one hot cv, teacher prefix one hot cv
                        ,project_grade_category_one_hot_cv,price_standardized_cv,quan
        tity_standardized cv
                        ,teacher number of previously posted projects standardized c
        v,tfidf w2v essays vectors cv
                        ,tfidf_w2v_titles_vectors_cv)).tocsr()
        X te = hstack((categories one hot test, sub categories one hot test, school stat
        e_one_hot_test,teacher_prefix_one_hot_test
                       ,project_grade_category_one_hot_test,price_standardized_test,qu
        antity standardized test
                       ,teacher number of previously posted projects standardized test
        ,tfidf_w2v_essays_vectors_test
                      ,tfidf w2v titles vectors test)).tocsr()
        print("Final Data matrix on TFIDF W2V")
        print(X tr.shape, y train.shape)
        # print(X cr.shape, y cv.shape)
        print(X_te.shape, y_test.shape)
        print("="*100)
        Final Data matrix on TFIDF W2V
        (73196, 702) (73196,)
        (36052, 702) (36052,)
        In [3]: len(tfidf_w2v_titles_vectors_test)
```

### 1.1 Method 1: GridSearchCV

Out[3]: 36052

```
In [4]: # https://machinelearningmastery.com/how-to-fix-futurewarning-messages-in-scik
        it-learn/
        from warnings import simplefilter
        # ignore all future warnings
        simplefilter(action='ignore', category=FutureWarning)
        from sklearn.model selection import GridSearchCV
        import matplotlib.pvplot as plt
        from sklearn.linear model import LogisticRegression
        from sklearn.metrics import roc auc score
        C = [0.003, 0.004, 0.005, 0.01, 0.05, 0.1, 0.5]
        logi = LogisticRegression(class weight='balanced')
        parameters = {'C':C}
        clf = GridSearchCV(logi, parameters, cv= 10, scoring='roc auc')
        clf.fit(X tr, y train)
        train auc= clf.cv results ['mean train score']
        train_auc_std= clf.cv_results_['std_train_score']
        cv auc = clf.cv results ['mean test score']
        cv auc std= clf.cv results ['std test score']
        plt.figure(figsize=(20,10))
        plt.plot(np.log(parameters['C']), train_auc, label='Train AUC')
        # https://stackoverflow.com/a/48803361/4084039
        plt.gca().fill between(np.log(parameters['C']),train auc - train auc std,train
         auc + train auc std,alpha=0.3,color='darkblue')
        plt.plot(np.log(parameters['C']), cv_auc, label='CV AUC')
        # https://stackoverflow.com/a/48803361/4084039
        plt.gca().fill_between(np.log(parameters['C']),cv_auc - cv_auc_std,cv_auc + cv
         _auc_std,alpha=0.3,color='darkorange')
        plt.scatter(np.log(parameters['C']), train auc, label='Train AUC points')
        plt.scatter(np.log(parameters['C']), cv auc, label='CV AUC points')
        plt.legend()
        plt.xlabel("Log(C): hyperparameter")
        plt.ylabel("AUC")
        plt.title("Lambda: hyperparameter v/s AUC plot")
        plt.grid()
```

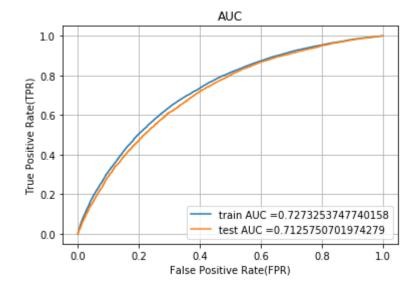


```
In [5]: def batch_predict(clf, data):
    # roc_auc_score(y_true, y_score) the 2nd parameter should be probability e
    stimates of the positive class
    # not the predicted outputs

    y_data_pred = []
    tr_loop = data.shape[0] - data.shape[0]%1000
    # consider you X_tr shape is 49041, then your cr_loop will be 49041 - 4904

1%1000 = 49000
    # in this for loop we will iterate unti the last 1000 multiplier
    for i in range(0, tr_loop, 1000):
        y_data_pred.extend(clf.predict_proba(data[i:i+1000])[:,1])
    # we will be predicting for the last data points
    y_data_pred.extend(clf.predict_proba(data[tr_loop:])[:,1])
    return y_data_pred
```

```
# https://scikit-learn.org/stable/modules/generated/sklearn.metrics.roc_curve.
In [10]:
         html#sklearn.metrics.roc curve
         from sklearn.metrics import roc curve, auc
         logi = LogisticRegression(C=0.05,class_weight='balanced')
         logi.fit(X_tr[:,:], y_train[:])
         # roc auc score(y true, y score) the 2nd parameter should be probability estim
         ates of the positive class
         # not the predicted outputs
         y_train_pred = batch_predict(logi, X_tr[:,:])
         y_test_pred = batch_predict(logi, X_te[:])
         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_tp
         r)))
         plt.plot(test fpr, test tpr, label="test AUC ="+str(auc(test fpr, test tpr)))
         plt.legend()
         plt.xlabel("False Positive Rate(FPR)")
         plt.ylabel("True Positive Rate(TPR)")
         plt.title("AUC")
         plt.grid()
         plt.show()
```



```
In [12]: print("="*100)
    from sklearn.metrics import confusion_matrix
    print("Train confusion matrix")
    print(confusion_matrix(y_train[:], predict(y_train_pred, tr_thresholds, train_fpr, train_tpr)))
    print("Test confusion matrix")
    print(confusion_matrix(y_test[:], predict(y_test_pred, tr_thresholds, test_fpr, test_tpr)))
```

-----

```
Train confusion matrix
the maximum value of tpr*(1-fpr) 0.45055004407556865 for threshold 0.489
[[ 7430  3653]
  [20369 41744]]
Test confusion matrix
the maximum value of tpr*(1-fpr) 0.4334934400158479 for threshold 0.586
[[ 4444  1015]
  [16791 13802]]
```

```
In [13]: conf_matr_df_train = pd.DataFrame(confusion_matrix(y_train[:], predict(y_train_pred, tr_thresholds, train_fpr, train_tpr)))
    sns.set(font_scale=1.4)#for label size
    sns.heatmap(conf_matr_df_train, annot=True,annot_kws={"size": 16}, fmt='g')
```

the maximum value of tpr\*(1-fpr) 0.45055004407556865 for threshold 0.489

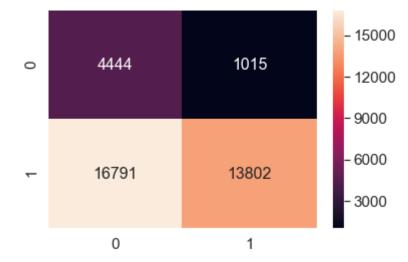
Out[13]: <matplotlib.axes.\_subplots.AxesSubplot at 0x1705a0093c8>



```
In [14]: conf_matr_df_test = pd.DataFrame(confusion_matrix(y_test[:], predict(y_test_p)
    red, tr_thresholds, test_fpr, test_tpr)))
    sns.set(font_scale=1.4)#for label size
    sns.heatmap(conf_matr_df_test, annot=True,annot_kws={"size": 16}, fmt='g')
```

the maximum value of tpr\*(1-fpr) 0.4334934400158479 for threshold 0.586

Out[14]: <matplotlib.axes.\_subplots.AxesSubplot at 0x17058bdd630>



```
In [0]: # please write all the code with proper documentation, and proper titles for e
        ach subsection
        # go through documentations and blogs before you start coding
        # first figure out what to do, and then think about how to do.
        # reading and understanding error messages will be very much helpfull in debug
        ging your code
        # when you plot any graph make sure you use
            # a. Title, that describes your plot, this will be very helpful to the rea
        der
            # b. Legends if needed
            # c. X-axis label
            # d. Y-axis Label
In [1]: | import dill
        # dill.dump session('notebook env.db')
        dill.load_session('notebook_env.db')
        C:\Users\LENOVO\Anaconda3\lib\site-packages\smart_open\ssh.py:34: UserWarnin
        g: paramiko missing, opening SSH/SCP/SFTP paths will be disabled.
        ll paramiko` to suppress
          warnings.warn('paramiko missing, opening SSH/SCP/SFTP paths will be disable
            `pip install paramiko` to suppress')
        C:\Users\LENOVO\Anaconda3\lib\site-packages\gensim\utils.py:1197: UserWarnin
        g: detected Windows; aliasing chunkize to chunkize_serial
          warnings.warn("detected Windows; aliasing chunkize to chunkize serial")
```

# 2.5 Logistic Regression with added Features 'Set 5'

### Adding number of words in TITLES feature

### Word counts(TITLES)

```
In [6]: | title wordcount train = []
        title_train = list(X_train['preprocessed_titles'])
        for i in tqdm1(title train):
            b = len(str(i).split())
            title wordcount train.append(b)
        title_wordcount_train = np.array(title_wordcount_train)
        # title wordcount cv = []
        # title_cv = list(X_cv['preprocessed_titles'])
        # for i in tqdm1(title_cv):
              b = len(str(i).split())
              title_wordcount_cv.append(b)
        # title_wordcount_cv = np.array(title_wordcount_cv)
        title wordcount test = []
        title_test = list(X_test['preprocessed_titles'])
        for i in tqdm1(title test):
            b = len(str(i).split())
            title_wordcount_test.append(b)
        title wordcount test = np.array(title wordcount test)
        print(title_wordcount_train.shape)
        # print(title wordcount cv.shape)
        print(title wordcount test.shape)
```

(73196,) (36052,)

#### **Standardizing Word counts(TITLES)**

```
In [7]: from sklearn.preprocessing import StandardScaler

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

    title_wordcount_standardized_train = title_wordcount_scalar.transform(title_wordcount_train.reshape(-1,1))

# title_wordcount_standardized_cv = title_wordcount_scalar.transform(title_wordcount_cv.reshape(-1,1))

title_wordcount_standardized_test = title_wordcount_scalar.transform(title_wordcount_test.reshape(-1,1))

print(title_wordcount_standardized_train.shape)

# print(title_wordcount_standardized_cv.shape)

print(title_wordcount_standardized_test.shape)
```

C:\Users\LENOVO\Anaconda3\lib\site-packages\sklearn\utils\validation.py:595:
DataConversionWarning:

Data with input dtype int32 was converted to float64 by StandardScaler.

C:\Users\LENOVO\Anaconda3\lib\site-packages\sklearn\utils\validation.py:595:
DataConversionWarning:

Data with input dtype int32 was converted to float64 by StandardScaler.

C:\Users\LENOVO\Anaconda3\lib\site-packages\sklearn\utils\validation.py:595:
DataConversionWarning:

Data with input dtype int32 was converted to float64 by StandardScaler.

(73196, 1) (36052, 1)

#### Word counts(ESSAYS)

```
In [8]: | essay wordcount train = []
        essay_train = list(X_train['preprocessed_essays'])
        for i in tqdm1(essay train):
            b = len(str(i).split())
            essay wordcount train.append(b)
        essay_wordcount_train = np.array(essay_wordcount_train)
        # essay wordcount cv = []
        # essay_cv = list(X_cv['preprocessed_essays'])
        # for i in tqdm1(essay_cv):
              b = len(str(i).split())
              essay_wordcount_cv.append(b)
        # essay_wordcount_cv = np.array(essay_wordcount_cv)
        essay wordcount test = []
        essay_test = list(X_test['preprocessed_titles'])
        for i in tqdm1(essay test):
            b = len(str(i).split())
            essay_wordcount_test.append(b)
        essay wordcount test = np.array(essay wordcount test)
        print(essay_wordcount_train.shape)
        # print(essay wordcount cv.shape)
        print(essay_wordcount_test.shape)
```

(73196,) (36052,)

### Standardizing Word counts(ESSAYS)

```
In [9]: from sklearn.preprocessing import StandardScaler
        essay wordcount scalar = StandardScaler()
        essay wordcount scalar.fit(essay wordcount train.reshape(-1,1))
        essay_wordcount_standardized_train = essay_wordcount_scalar.transform(essay_wo
        rdcount train.reshape(-1,1))
        # essay wordcount standardized cv = essay wordcount scalar.transform(essay wor
        dcount cv.reshape(-1,1))
        essay_wordcount_standardized_test = essay_wordcount_scalar.transform(essay_wor
        dcount test.reshape(-1,1))
        print(essay_wordcount_standardized_train.shape)
        # print(essay wordcount standardized cv.shape)
        print(essay wordcount standardized test.shape)
        C:\Users\LENOVO\Anaconda3\lib\site-packages\sklearn\utils\validation.py:595:
        DataConversionWarning:
        Data with input dtype int32 was converted to float64 by StandardScaler.
        C:\Users\LENOVO\Anaconda3\lib\site-packages\sklearn\utils\validation.py:595:
        DataConversionWarning:
        Data with input dtype int32 was converted to float64 by StandardScaler.
        C:\Users\LENOVO\Anaconda3\lib\site-packages\sklearn\utils\validation.py:595:
        DataConversionWarning:
```

Data with input dtype int32 was converted to float64 by StandardScaler.

(73196, 1) (36052, 1)

# Sentiment scores for each essay

```
In [10]:
         import nltk
         from nltk.sentiment.vader import SentimentIntensityAnalyzer
         import nltk
         # nltk.download('vader lexicon')
         sid = SentimentIntensityAnalyzer()
         for sentiment = '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 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 o
         ur school is a caring community of successful \
         learners which can be seen through collaborative student project based learnin
         g in and out of the classroom kindergarteners \
         in my class love to work with hands on materials and have many different oppor
         tunities to practice a skill before it is\
         mastered having the social skills to work cooperatively with friends is a cruc
         ial aspect of the kindergarten curriculum\
         montana is the perfect place to learn about agriculture and nutrition my stude
         nts 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 writi
         ng 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 knowledge \
         of where the ingredients came from as well as how it is healthy for their bodi
         es 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 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'
         ss = sid.polarity scores(for sentiment)
         # for k in ss:
               print('{0}: {1}, '.format(k, ss[k]), end='')
         print(ss)
         {'neg': 0.01, 'neu': 0.745, 'pos': 0.245, 'compound': 0.9975}
In [11]: | lst = []
         list(ss.values())
Out[11]: [0.01, 0.745, 0.245, 0.9975]
```

```
In [12]:
         import nltk
         from nltk.sentiment.vader import SentimentIntensityAnalyzer
         import nltk
         # nltk.download('vader lexicon')
         sid = SentimentIntensityAnalyzer()
         essay sentscore train = []
         essay_train = list(X_train['preprocessed_essays'])
         for i in tqdm1(essay train):
             ss = sid.polarity_scores(str(i))
             essay_sentscore_train.append(ss)
         essay sentscore train = np.array(essay sentscore train)
         essay_negscore_train = []
         essay neuscore train = []
         essay_posscore_train = []
         essay_compoundscore_train = []
         for it in essay sentscore train:
             a = it['neg']
             essay_negscore_train.append(a)
             b = it['neu']
             essay_neuscore_train.append(b)
             c = it['pos']
             essay posscore train.append(c)
             d = it['compound']
             essay_compoundscore_train.append(d)
         essay_negscore_train = np.array(essay_negscore_train).reshape(-1,1)
         essay_neuscore_train = np.array(essay_neuscore_train).reshape(-1,1)
         essay_posscore_train = np.array(essay_posscore_train).reshape(-1,1)
         essay compoundscore train = np.array(essay compoundscore train).reshape(-1,1)
         print((essay_negscore_train.shape))
         print((essay neuscore train.shape))
         print((essay posscore train.shape))
         print((essay_compoundscore_train.shape))
         # essay sentscore cv = []
         # essay_cv = list(X_cv['preprocessed_essays'])
         # for i in tqdm1(essay_cv):
              ss = sid.polarity scores(str(i))
               essay sentscore cv.append(ss)
         # essay_sentscore_cv = np.array(essay_sentscore_cv)
         \# essay negscore cv = []
         # essay neuscore cv = []
         # essay posscore cv = []
         # essay compoundscore cv = []
         # for it in essay_sentscore_cv:
              a = it['neg']
              essay negscore cv.append(a)
               b = it['neu']
```

```
#
     essay neuscore cv.append(b)
#
     c = it['pos']
     essay_posscore_cv.append(c)
     d = it['compound']
     essay compoundscore cv.append(d)
# essay negscore cv = np.array(essay negscore cv).reshape(-1,1)
# essay_neuscore_cv = np.array(essay_neuscore_cv).reshape(-1,1)
# essay_posscore_cv = np.array(essay_posscore_cv).reshape(-1,1)
# essay compoundscore cv = np.array(essay compoundscore cv).reshape(-1,1)
# print((essay negscore cv.shape))
# print((essay neuscore cv.shape))
# print((essay_posscore_cv.shape))
# print((essay compoundscore cv.shape))
essay sentscore test = []
essay_test = list(X_test['preprocessed_essays'])
for i in tqdm1(essay test):
   ss = sid.polarity scores(str(i))
   essay sentscore test.append(ss)
essay_sentscore_test = np.array(essay_sentscore_test)
essay_negscore_test = []
essay neuscore test = []
essay_posscore_test = []
essay compoundscore test = []
for it in essay sentscore test:
   a = it['neg']
   essay negscore test.append(a)
   b = it['neu']
   essay neuscore test.append(b)
   c = it['pos']
   essay posscore test.append(c)
   d = it['compound']
   essay compoundscore test.append(d)
essay negscore test = np.array(essay negscore test).reshape(-1,1)
essay neuscore test = np.array(essay neuscore test).reshape(-1,1)
essay_posscore_test = np.array(essay_posscore_test).reshape(-1,1)
essay_compoundscore_test = np.array(essay_compoundscore_test).reshape(-1,1)
print((essay negscore test.shape))
print((essay neuscore test.shape))
print((essay_posscore_test.shape))
print((essay compoundscore test.shape))
```

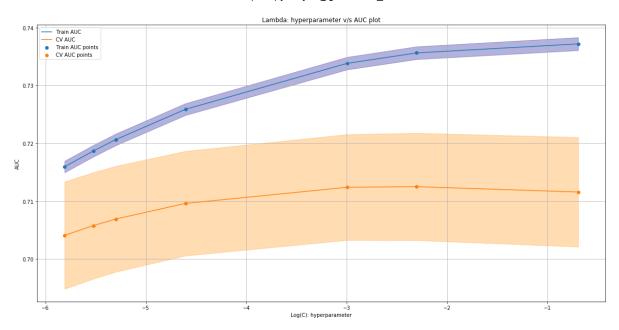
```
(73196, 1)
         (73196, 1)
         (73196, 1)
         (73196, 1)
         (36052, 1)
         (36052, 1)
         (36052, 1)
         (36052, 1)
In [13]: # merge two sparse matrices: https://stackoverflow.com/a/19710648/4084039
         from scipy.sparse import hstack
         X tr = hstack((categories one hot train, sub categories one hot train, school st
         ate one hot train, teacher prefix one hot train
                         ,project grade category one hot train, price standardized train,
         quantity_standardized_train
                         ,teacher_number_of_previously_posted_projects_standardized_trai
         n,tfidf w2v essays vectors train
                         tfidf w2v titles vectors train, title wordcount standardized tr
         ain,essay_wordcount_standardized_train
                         ,essay_negscore_train,essay_neuscore_train,essay_posscore_train
         ,essay_compoundscore_train)).tocsr()
         # X_cr = hstack((categories_one_hot_cv,sub_categories_one_hot_cv,school_state_
         one hot cv, teacher prefix one hot cv
                           ,project grade category one hot cv,price standardized cv,quan
         tity standardized cv
                           ,teacher number of previously posted projects standardized c
         v,tfidf w2v essays vectors cv
                           ,tfidf_w2v_titles_vectors_cv,title_wordcount_standardized_cv,
         essay wordcount standardized cv
                           ,essay_negscore_cv,essay_neuscore_cv,essay_posscore_cv,essay_
         compoundscore cv)).tocsr()
         X te = hstack((categories one hot test, sub categories one hot test, school stat
         e one hot test, teacher prefix one hot test
                         ,project_grade_category_one_hot_test,price_standardized_test,qu
         antity standardized test
                         ,teacher number of previously posted projects standardized test
         tfidf w2v essays vectors test
                         ,tfidf_w2v_titles_vectors_test,title_wordcount_standardized_tes
         t,essay wordcount standardized test
                         ,essay negscore test,essay neuscore test,essay posscore test,es
         say compoundscore test)).tocsr()
         print("Final Data matrix on TFIDF W2V")
         print(X_tr.shape, y_train.shape)
         # print(X cr.shape, y cv.shape)
         print(X te.shape, y test.shape)
         print("="*100)
         Final Data matrix on TFIDF W2V
```

```
Final Data matrix on TFIDF W2V (73196, 708) (73196,) (36052, 708) (36052,)
```

-----

# 1.1 Method 1: GridSearchCV

In [14]: | # https://machinelearningmastery.com/how-to-fix-futurewarning-messages-in-scik it-learn/ from warnings import simplefilter # ignore all future warnings simplefilter(action='ignore', category=FutureWarning) from sklearn.model selection import GridSearchCV import matplotlib.pvplot as plt from sklearn.linear model import LogisticRegression from sklearn.metrics import roc auc score C = [0.003, 0.004, 0.005, 0.01, 0.05, 0.1, 0.5]logi = LogisticRegression(class weight='balanced') parameters = {'C':C} clf = GridSearchCV(logi, parameters, cv= 10, scoring='roc auc') clf.fit(X tr, y train) train auc= clf.cv results ['mean train score'] train\_auc\_std= clf.cv\_results\_['std\_train\_score'] cv auc = clf.cv results ['mean test score'] cv auc std= clf.cv results ['std test score'] plt.figure(figsize=(20,10)) plt.plot(np.log(parameters['C']), train\_auc, label='Train AUC') # https://stackoverflow.com/a/48803361/4084039 plt.gca().fill between(np.log(parameters['C']),train auc - train auc std,train auc + train auc std,alpha=0.3,color='darkblue') plt.plot(np.log(parameters['C']), cv\_auc, label='CV AUC') # https://stackoverflow.com/a/48803361/4084039 plt.gca().fill\_between(np.log(parameters['C']),cv\_auc - cv\_auc\_std,cv\_auc + cv \_auc\_std,alpha=0.3,color='darkorange') plt.scatter(np.log(parameters['C']), train auc, label='Train AUC points') plt.scatter(np.log(parameters['C']), cv auc, label='CV AUC points') plt.legend() plt.xlabel("Log(C): hyperparameter") plt.ylabel("AUC") plt.title("Lambda: hyperparameter v/s AUC plot") plt.grid()

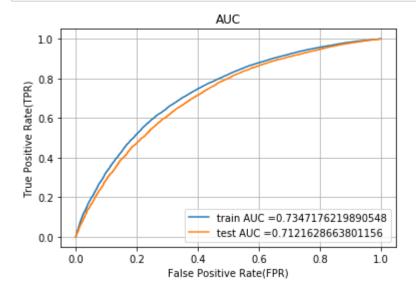


```
In [15]: def batch_predict(clf, data):
    # roc_auc_score(y_true, y_score) the 2nd parameter should be probability e
    stimates of the positive class
    # not the predicted outputs

y_data_pred = []
    tr_loop = data.shape[0] - data.shape[0]%1000
    # consider you X_tr shape is 49041, then your cr_loop will be 49041 - 4904

1%1000 = 49000
    # in this for loop we will iterate unti the last 1000 multiplier
    for i in range(0, tr_loop, 1000):
        y_data_pred.extend(clf.predict_proba(data[i:i+1000])[:,1])
    # we will be predicting for the last data points
    y_data_pred.extend(clf.predict_proba(data[tr_loop:])[:,1])
    return y_data_pred
```

```
In [17]: # https://scikit-learn.org/stable/modules/generated/sklearn.metrics.roc_curve.
         html#sklearn.metrics.roc curve
         from sklearn.metrics import roc curve, auc
         logi = LogisticRegression(C=0.1,class_weight='balanced')
         logi.fit(X_tr[:,:], y_train[:])
         # roc auc score(y true, y score) the 2nd parameter should be probability estim
         ates of the positive class
         # not the predicted outputs
         y_train_pred = batch_predict(logi, X_tr[:,:])
         y_test_pred = batch_predict(logi, X_te[:])
         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_tp
         r)))
         plt.plot(test fpr, test tpr, label="test AUC ="+str(auc(test fpr, test tpr)))
         plt.legend()
         plt.xlabel("False Positive Rate(FPR)")
         plt.ylabel("True Positive Rate(TPR)")
         plt.title("AUC")
         plt.grid()
         plt.show()
```



```
In [19]: print("="*100)
    from sklearn.metrics import confusion_matrix
    print("Train confusion matrix")
    print(confusion_matrix(y_train[:], predict(y_train_pred, tr_thresholds, train_fpr, train_tpr)))
    print("Test confusion matrix")
    print(confusion_matrix(y_test[:], predict(y_test_pred, tr_thresholds, test_fpr, test_tpr)))
```

\_\_\_\_\_\_

```
Train confusion matrix
the maximum value of tpr*(1-fpr) 0.4577614555945954 for threshold 0.496
[[ 7646  3437]
  [20899  41214]]
Test confusion matrix
the maximum value of tpr*(1-fpr) 0.43409173762084863 for threshold 0.602
[[ 5262  197]
  [26664  3929]]
```

```
In [20]: conf_matr_df_train = pd.DataFrame(confusion_matrix(y_train[:], predict(y_train_pred, tr_thresholds, train_fpr, train_tpr)))
    sns.set(font_scale=1.4)#for label size
    sns.heatmap(conf_matr_df_train, annot=True,annot_kws={"size": 16}, fmt='g')
```

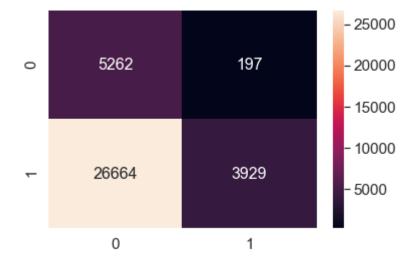
the maximum value of tpr\*(1-fpr) 0.4577614555945954 for threshold 0.496

Out[20]: <matplotlib.axes. subplots.AxesSubplot at 0x26f5ae47160>



the maximum value of tpr\*(1-fpr) 0.43409173762084863 for threshold 0.602

Out[21]: <matplotlib.axes.\_subplots.AxesSubplot at 0x26f5aecb898>



### 3. Conclusion

```
In [22]: from prettytable import PrettyTable
#If you get a ModuleNotFoundError error , install prettytable using: pip3 inst
all prettytable
x = PrettyTable()
x.field_names = ["Vectorizer", "Model", "Hyper Parameter", "AUC"]
x.add_row(["BOW", "Logistic Regression", 0.003, 0.73])
x.add_row(["TFIDF", "Logistic Regression", 0.1, 0.73])
x.add_row(["AVG W2V", "Logistic Regression", 0.5, 0.72])
x.add_row(["TFIDF W2V", "Logistic Regression", 0.05, 0.71])
x.add_row(["ADDITIONAL FEATURES", "Logistic Regression", 0.1, 0.71])
print(x)
```

Vectorizer	Model	Hyper Parameter	++   AUC
BOW TFIDF AVG W2V TFIDF W2V ADDITIONAL FEATURES	Logistic Regression Logistic Regression Logistic Regression Logistic Regression Logistic Regression	0.003 0.1 0.5 0.05 0.1	0.73     0.73     0.72     0.71

In [ ]: #