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:

De	Feature
A unique identifier for the proposed project. Example:	project_id
Title of the project. E	
• Art Will Make You • First Gr	project_title
Grade level of students for which the project is targeted. One of the enumerate	
 Grades Gra Gra Grac 	project_grade_category
One or more (comma-separated) subject categories for the project following enumerated list (
 Applied L Care & Health & History & Literacy & L Math & Music & 1 Specia 	project_subject_categories
• Music & 1 • Literacy & Language, Math &	
State where school is located (<u>Two-letter U.S. perfections.//en.wikipedia.org/wiki/List of U.S. state abbreviations</u> #Posta	school_state
One or more (comma-separated) subject subcategories for the E: Literature & Writing, Social S	<pre>project_subject_subcategories</pre>
An explanation of the resources needed for the project. I	
 My students need hands on literacy materials to sensory 	project_resource_summary
First applicat	project_essay_1
Second applicat	project_essay_2
Third applicat	project_essay_3
Fourth applicat	project_essay_4
Datetime when project application was submitted. Example: 201 12:45	<pre>project_submitted_datetime</pre>
A unique identifier for the teacher of the proposed project. bdf8baa8fedef6bfeec7ae4ff	teacher_id

Feature De

Teacher's title. One of the following enumerate

teacher_prefix

•

teacher_number_of_previously_posted_projects

Number of project applications previously submitted by the sam

Ev

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

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

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

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

Label

Description

A binary flag indicating whether DonorsChoose approved the project. A value of 0 indicates the

Notes on the Essay Data

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

- __project_essay_1:__ "Introduce us to your classroom"
- project essay 2: "Tell us more about your students"
- __project_essay_3:__ "Describe how your students will use the materials you're requesting"
- __project_essay_3:__ "Close by sharing why your project will make a difference"

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

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

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

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

In [1]:

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

```
C:\Users\hp\Anaconda3\lib\site-packages\gensim\utils.py:1197: UserWarning:
detected Windows; aliasing chunkize to chunkize_serial
  warnings.warn("detected Windows; aliasing chunkize to chunkize serial")
```

1.1 Reading Data

```
In [2]:
```

```
project_data = pd.read_csv('train_data.csv', nrows=50000)
resource_data = pd.read_csv('resources.csv')
```

```
In [3]:
```

1 p069063

```
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 (50000, 17)
The attributes of data : ['Unnamed: 0' 'id' 'teacher_id' 'teacher_prefix'
'school_state'
 'project_submitted_datetime' 'project_grade_category'
 'project_subject_categories' 'project_subject_subcategories'
 'project_title' 'project_essay_1' 'project_essay_2' 'project_essay_3'
 'project essay 4' 'project resource summary'
 'teacher_number_of_previously_posted_projects' 'project_is_approved']
In [4]:
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[4]:
        id
                                         description quantity
                                                             price
0 p233245 LC652 - Lakeshore Double-Space Mobile Drying Rack
                                                         1 149.00
```

14.95

1.2 preprocessing of project_subject_categories

Bouncy Bands for Desks (Blue support pipes)

In [5]:

```
catogories = list(project data['project subject categories'].values)
# remove special characters from list of strings python: https://stackoverflow.com/a/47
301924/4084039
# https://www.geeksforgeeks.org/removing-stop-words-nltk-python/
# https://stackoverflow.com/questions/23669024/how-to-strip-a-specific-word-from-a-stri
ng
# https://stackoverflow.com/questions/8270092/remove-all-whitespace-in-a-string-in-pyth
cat list = []
for i in catogories:
    temp = ""
    # consider we have text like this "Math & Science, Warmth, Care & Hunger"
   for j in i.split(','): # it will split it in three parts ["Math & Science", "Warmt
h", "Care & Hunger"]
        if 'The' in j.split(): # this will split each of the catogory based on space "M
ath & 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 ''(empty) ex:"M
ath & Science"=>"Math&Science"
        temp+=j.strip()+" " #" abc ".strip() will return "abc", remove the trailing spa
ces
        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 [6]:

```
sub catogories = list(project data['project subject subcategories'].values)
# remove special characters from list of strings python: https://stackoverflow.com/a/47
301924/4084039
# https://www.geeksforgeeks.org/removing-stop-words-nltk-python/
# https://stackoverflow.com/questions/23669024/how-to-strip-a-specific-word-from-a-stri
# https://stackoverflow.com/questions/8270092/remove-all-whitespace-in-a-string-in-pyth
sub_cat_list = []
for i in sub catogories:
    temp = ""
    # consider we have text like this "Math & Science, Warmth, Care & Hunger"
   for j in i.split(','): # it will split it in three parts ["Math & Science", "Warmt
h", "Care & Hunger"]
        if 'The' in j.split(): # this will split each of the catogory based on space "M
ath & 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 ''(empty) ex:"M
ath & Science"=>"Math&Science"
        temp +=j.strip()+" "#" abc ".strip() will return "abc", remove the trailing spa
ces
        temp = temp.replace('&','_')
    sub_cat_list.append(temp.strip())
project data['clean subcategories'] = sub cat list
project data.drop(['project subject subcategories'], axis=1, inplace=True)
# count of all the words in corpus python: https://stackoverflow.com/a/22898595/4084039
my counter = Counter()
for word in project_data['clean_subcategories'].values:
    my counter.update(word.split())
sub cat dict = dict(my counter)
sorted_sub_cat_dict = dict(sorted(sub_cat_dict.items(), key=lambda kv: kv[1]))
```

1.3 Text preprocessing

In [7]:

In [8]:

project_data.head(2)

Out[8]:

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

In [9]:

1.4.2.3 Using Pretrained Models: TFIDF weighted W2V

In [10]:

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

My students are English learners that are working on English as their seco nd or third languages. We are a melting pot of refugees, immigrants, and n ative-born 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 repres ented with the families within our school. Each student brings a wealth o f knowledge and experiences to us that open our eyes to new cultures, beli efs, and respect.\"The limits of your language are the limits of your worl d.\"-Ludwig Wittgenstein Our English learner's have a strong support syst em at home that begs for more resources. Many times our parents are learn ing to read and speak English along side of their children. Sometimes thi s creates barriers for parents to be able to help their child learn phonet ics, letter recognition, and other reading skills.\r\n\r\nBy providing the se dvd's and players, students are able to continue their mastery of the E nglish language even if no one at home is able to assist. All families wi th students within the Level 1 proficiency status, will be a offered to be a part of this program. These educational videos will be specially chosen by the English Learner Teacher and will be sent home regularly to watch. The videos are to help the child develop early reading skills.\r\n\r\nPare nts that do not have access to a dvd player will have the opportunity to c heck out a dvd player to use for the year. The plan is to use these video s and educational dvd's for the years to come for other EL students.\r\nna nnan

The 51 fifth grade students that will cycle through my classroom this year all love learning, at least most of the time. At our school, 97.3% of the students receive free or reduced price lunch. Of the 560 students, 97.3% a re minority students. \r\nThe school has a vibrant community that loves to get together and celebrate. Around Halloween there is a whole school parad e to show off the beautiful costumes that students wear. On Cinco de Mayo we put on a big festival with crafts made by the students, dances, and gam es. At the end of the year the school hosts a carnival to celebrate the ha rd work put in during the school year, with a dunk tank being the most pop ular activity. My students will use these five brightly colored Hokki stool s in place of regular, stationary, 4-legged chairs. As I will only have a total of ten in the classroom and not enough for each student to have an i ndividual one, they will be used in a variety of ways. During independent reading time they will be used as special 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 the day they will be used by th e students who need the highest amount of movement in their life in order to stay focused on school.\r\n\r\nWhenever asked what the classroom is mis sing, my students always say more Hokki Stools. They can't get their fill of the 5 stools we already have. When the students are sitting in group wi th me on the Hokki Stools, they are always moving, but at the same time do ing their work. Anytime the students get to pick where they can sit, the H okki Stools are the first to be taken. There are always students who 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 student s to do desk work and move at the same time. These stools will help studen ts to meet their 60 minutes a day of movement by allowing them to activate their core muscles for balance while they sit. For many of my students, th ese chairs will take away the barrier that exists in schools for a child w ho can't sit still.nannan

How do you remember your days of school? Was it in a sterile environment w ith plain walls, rows of desks, and a teacher in front of the room? A typi cal day in our room is nothing like that. I work hard to create a warm inv iting themed 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 r

aces 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 sch ool is an \"open classroom\" concept, which is very unique as there are no walls separating the classrooms. These 9 and 10 year-old students are very eager learners; they are like sponges, absorbing all the information and e xperiences and keep on wanting more. With these resources such as the comfy red throw pillows and the whimsical nautical hanging decor and the blue fi sh nets, I will be able to help create the mood in our classroom setting t o be one of a themed nautical environment. Creating a classroom environmen t is very important in the success in each and every child's education. Th e nautical photo props will be used with each child as they step foot into our classroom for the first time on Meet the Teacher evening. I'll take pi ctures 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 thank you c ards will be used throughout the year by the students as they create thank you cards to their team groups.\r\n\r\nYour generous donations will help m e to help make our classroom a fun, inviting, learning environment from da y one.\r\n\r\nIt costs lost of money out of my own pocket on resources to get 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 language delays, cognitive delays, gross/fine motor delays, to autism. The y are eager beavers and always strive to work their hardest working past t heir limitations. \r\n\r\nThe materials we have are the ones I seek out fo r my students. I teach in a Title I school where most of the students rece ive free or reduced price lunch. Despite their disabilities and limitatio ns, my students love coming to school and come eager to learn and explore. Have you ever felt like you had ants in your pants and you needed to groov e and move as you were in a meeting? This is how my kids feel all the tim e. 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 enha nces gross motor and in Turn fine motor skills. \r\nThey also want to lear n through games, my kids don't want to sit and do worksheets. They want to learn to count by jumping and playing. Physical engagement is the key to o ur 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 [11]:

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

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

# general
    phrase = re.sub(r"n\'t", " not", phrase)
    phrase = re.sub(r"\'re", " are", phrase)
    phrase = re.sub(r"\'s", " is", phrase)
    phrase = re.sub(r"\'d", " would", phrase)
    phrase = re.sub(r"\'d", " 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 [12]:

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

My kindergarten students have varied disabilities ranging from speech and language delays, cognitive delays, gross/fine motor delays, to autism. The y are eager beavers and always strive to work their hardest working past t heir limitations. \r\n\r\nThe materials we have are the ones I seek out fo r my students. I teach in a Title I school where most of the students rece ive free or reduced price lunch. Despite their disabilities and limitatio ns, my students love coming to school and come eager to learn and explore. Have you ever felt like you had ants in your pants and you needed to groov e and move as you were in a meeting? This is how my kids feel all the tim e. 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 enha nces gross motor and in Turn fine motor skills. \r\nThey also want to lear n through games, my kids do not want to sit and do worksheets. They want t o 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 happe n. My students will forget they are doing work and just have the fun a 6 y ear old deserves.nannan

In [13]:

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

My kindergarten students have varied disabilities ranging from speech and language delays, cognitive delays, gross/fine motor delays, to autism. The y are eager beavers and always strive to work their hardest working past t heir limitations. 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 f ree or reduced price lunch. Despite their disabilities and limitations, m y students love coming to school and come eager to learn and explore. Have you ever felt like you had ants in 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 gro ss 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 succes s. The number toss and color and shape mats can make that happen. My stude nts will forget they are doing work and just have the fun a 6 year old des erves.nannan

In [14]:

```
#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 language delays cognitive delays gross fine motor delays to autism They ar e eager beavers and always strive to work their hardest working past their limitations 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 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 like you had ants in your pants and you needed to groove and move as you w ere 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 lo ve 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 no t 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 [15]:

```
# 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'r
e", "you've",\
            "you'll", "you'd", 'your', 'yours', 'yourself', 'yourselves', 'he', 'him',
'his', 'himself', \
            'she', "she's", 'her', 'hers', 'herself', 'it', "it's", 'its', 'itself', 't
hey', 'them', 'their',\
            'theirs', 'themselves', 'what', 'which', 'who', 'whom', 'this', 'that', "th
at'll", 'these', 'those', \
            'am', 'is', 'are', 'was', 'were', 'be', 'been', 'being', 'have', 'has', 'ha
d', 'having', 'do', 'does', \
            'did', 'doing', 'a', 'an', 'the', 'and', 'but', 'if', 'or', 'because', 'as'
, 'until', 'while', 'of', \
            'at', 'by', 'for', 'with', 'about', 'against', 'between', 'into', 'through'
 'during', 'before', 'after',\
            'above', 'below', 'to', 'from', 'up', 'down', 'in', 'out', 'on', 'off', 'ov
er', 'under', 'again', 'further',\
            'then', 'once', 'here', 'there', 'when', 'where', 'why', 'how', 'all', 'an
                   'few', 'more',\
y', 'both', 'each',
            'most', 'other', 'some', 'such', 'only', 'own', 'same', 'so', 'than', 'too'
, 'very', \
            's', 't', 'can', 'will', 'just', 'don', "don't", 'should', "should've", 'no
w', '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", 'ma', 'migh
tn', "mightn't", 'mustn',\
            "mustn't", 'needn', "needn't", 'shan', "shan't", 'shouldn', "shouldn't". 'w
asn', "wasn't", 'weren', "weren't", \
            'won', "won't", 'wouldn', "wouldn't"]
```

In [16]:

```
# 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('\\n', '')
    sent = re.sub('[^A-Za-z0-9]+', '', sent)
    # https://gist.github.com/sebleier/554280
    sent = ''.join(e for e in sent.split() if e.lower() not in stopwords)
    preprocessed_essays.append(sent.lower().strip())
```

```
100%| 50000/50000 [00:52<00:00, 944.22it/s]
```

In [17]:

```
# after preprocesing
preprocessed_essays[20000]
```

Out[17]:

'kindergarten students varied disabilities ranging speech language delays cognitive delays gross fine motor delays autism eager beavers always striv e work hardest working past limitations materials ones seek students teach title school students receive free reduced price lunch despite disabilities limitations students love coming school come eager learn explore ever felt like ants pants needed groove move meeting kids feel time want able move learn say wobble chairs answer love develop core enhances gross motor turn fine motor skills also want learn games kids not want sit worksheets want learn count jumping playing physical engagement key success number toss color shape mats make happen students forget work fun 6 year old deserves nannan'

In [18]:

```
project_data['essay'] = preprocessed_essays
```

1.4 Preprocessing of `project_title`

In [19]:

```
# similarly you can preprocess the titles also
```

In [20]:

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

```
100%| 50000/50000 [00:02<00:00, 17612.70it/s]
```

In [21]:

```
project_data['project_title'] = preprocessed_titles
```

In [22]:

```
#Preprocessing project_grade_category

#reference link: https://stackoverflow.com/questions/28986489/python-pandas-how-to-repl
ace-a-characters-in-a-column-of-a-dataframe

project_data['project_grade_category'] = project_data['project_grade_category'].str.rep
lace('-', '_')
project_data['project_grade_category'] = project_data['project_grade_category'].str.rep
lace(' ', '_')
```

1.5 Preparing data for models

```
In [23]:
project_data.columns
Out[23]:
Index(['Unnamed: 0', 'id', 'teacher_id', 'teacher_prefix', 'school_state',
       'project_submitted_datetime', 'project_grade_category', 'project_ti
tle',
       'project_essay_1', 'project_essay_2', 'project_essay_3',
       'project_essay_4', 'project_resource_summary',
       'teacher_number_of_previously_posted_projects', 'project_is_approve
d',
       'clean_categories', 'clean_subcategories', 'essay'],
      dtype='object')
we are going to consider
      - school_state : categorical data
      - clean_categories : categorical data
      - clean subcategories : categorical data
       - project_grade_category : categorical data
      - teacher prefix : categorical data
      - project_title : text data
       - text : text data
      - project_resource_summary: text data (optinal)
      - quantity : numerical (optinal)
      - teacher_number_of_previously_posted_projects : numerical
```

1.5.1 Vectorizing Categorical data

- price : numerical

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

In [24]:

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

['Warmth', 'Care_Hunger', 'History_Civics', 'Music_Arts', 'AppliedLearning', 'SpecialNeeds', 'Health_Sports', 'Math_Science', 'Literacy_Language'] Shape of matrix after one hot encodig (50000, 9)

In [25]:

```
# we use count vectorizer to convert the values into one
vectorizer = CountVectorizer(vocabulary=list(sorted_sub_cat_dict.keys()), lowercase=Fal
se, binary=True)
sub_categories_one_hot = vectorizer.fit_transform(project_data['clean_subcategories'].v
alues)
print(vectorizer.get_feature_names())
print("Shape of matrix after one hot encodig ",sub_categories_one_hot.shape)
```

```
['Economics', 'CommunityService', 'FinancialLiteracy', 'ParentInvolvemen t', 'Extracurricular', 'Civics_Government', 'ForeignLanguages', 'Nutrition Education', 'Warmth', 'Care_Hunger', 'SocialSciences', 'PerformingArts', 'CharacterEducation', 'TeamSports', 'Other', 'College_CareerPrep', 'Musi c', 'History_Geography', 'Health_LifeScience', 'EarlyDevelopment', 'ESL', 'Gym_Fitness', 'EnvironmentalScience', 'VisualArts', 'Health_Wellness', 'A ppliedSciences', 'SpecialNeeds', 'Literature_Writing', 'Mathematics', 'Literacy']
Shape of matrix after one hot encodig (50000, 30)
```

In [26]:

you can do the similar thing with state, teacher_prefix and project_grade_category al so

In [27]:

In [28]:

```
#teacher_prefix
vectorizer = CountVectorizer(lowercase=False, binary=True)
vectorizer.fit(project_data['teacher_prefix'].values.astype('U'))
#While running this i got an error:np.nan is an invalid document, expected byte or unic
ode string.
#I fixed it by using stackoverflow.com
#https://stackoverflow.com/questions/39303912/tfidfvectorizer-in-scikit-learn-valueerro
r-np-nan-is-an-invalid-document
print(vectorizer.get_feature_names())

teacher_prefix_one_hot = vectorizer.transform(project_data['teacher_prefix'].values.ast
ype('U')) #here also i did same as above
print('Shape of matrix of one hot encoding', teacher_prefix_one_hot.shape)
```

```
['Dr', 'Mr', 'Mrs', 'Ms', 'Teacher', 'nan'] Shape of matrix of one hot encoding (50000, 6)
```

In [29]:

```
#project_grade_category
vectorizer = CountVectorizer(lowercase=False, binary=True)
vectorizer.fit(project_data['project_grade_category'].values.astype('U'))
print(vectorizer.get_feature_names())

project_grade_category_one_hot = vectorizer.fit_transform(project_data['project_grade_category'].values.astype('U'))
print('Shape of matrix of one hot encoding', project_grade_category_one_hot.shape)
```

```
['Grades_3_5', 'Grades_6_8', 'Grades_9_12', 'Grades_PreK_2'] Shape of matrix of one hot encoding (50000, 4)
```

1.5.2 Vectorizing Text data

1.5.2.1 Bag of words

In [30]:

```
# We are considering only the words which appeared in at least 10 documents(rows or pro
jects).
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 (50000, 12101)

In [31]:

```
# you can vectorize the title also
# before you vectorize the title make sure you preprocess it
```

In [32]:

```
vectorizer = CountVectorizer(min_df=10)
title_bow = vectorizer.fit_transform(preprocessed_titles)
print("Shape of matrix after one hot encodig ",title_bow.shape)
```

Shape of matrix after one hot encodig (50000, 2039)

1.5.2.2 TFIDF vectorizer

In [33]:

```
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 (50000, 12101)

In [34]:

```
vectorizer = TfidfVectorizer(min_df=10)
title_tfidf = vectorizer.fit_transform(preprocessed_titles)
print("Shape of matrix after one hot encodig ",title_tfidf.shape)
```

Shape of matrix after one hot encodig (50000, 2039)

1.5.2.3 Using Pretrained Models: Avg W2V

In [35]:

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

Out[35]:

```
'\n# Reading glove vectors in python: https://stackoverflow.com/a/3823034
9/4084039\ndef loadGloveModel(gloveFile):\n
                                            print ("Loading Glove Mode
                                                     model = {} \n
        f = open(gloveFile,\'r\', encoding="utf8")\n
or line in tqdm(f):\n
                          splitLine = line.split()\n
                                                           word = spli
                 embedding = np.array([float(val) for val in splitLine
tLine[0]\n
                                          print ("Done.",len(model)," w
[1:]])\n
               model[word] = embedding\n
ords loaded!")\n
                  return model\nmodel = loadGloveModel(\'glove.42B.300d.
txt\')\n\n# ========\nOutput:\n
                                                    \nLoading Glove Mod
el\n1917495it [06:32, 4879.69it/s]\nDone. 1917495 words loaded!\n\n# ====
ds.extend(i.split(\' \'))\n\nfor i in preproced_titles:\n
                                                         words.extend
(i.split(\' \'))\nprint("all the words in the coupus", len(words))\nwords
= set(words)\nprint("the unique words in the coupus", len(words))\n\ninter
_words = set(model.keys()).intersection(words)\nprint("The number of words
that are present in both glove vectors and our coupus",
                                                          len(inter wo
rds),"(",np.round(len(inter_words)/len(words)*100,3),"%)")\n\nwords_courpu
s = {}\nwords_glove = set(model.keys())\nfor i in words:\n
                                                          if i in word
s glove:\n
                words_courpus[i] = model[i]\nprint("word 2 vec length",
len(words courpus))\n\n# stronging variables into pickle files python: h
ttp://www.jessicayung.com/how-to-use-pickle-to-save-and-load-variables-in-
python/\n\nimport pickle\nwith open(\'glove_vectors\', \'wb\') as f:\n
pickle.dump(words_courpus, f)\n\n'
```

In [36]:

```
# stronging variables into pickle files python: http://www.jessicayung.com/how-to-use-p
ickle-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())
```

In [37]:

```
# 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[0]))
```

```
100%
```

| 50000/50000 [00:25<00:00, 1961.42it/s]

50000 300

1.5.2.3 Using Pretrained Models: TFIDF weighted W2V

In [38]:

```
# 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 [39]:

```
# average Word2Vec
# compute average word2vec for each review.
tfidf_w2v_vectors = []; # the avg-w2v for each sentence/review is stored in this list
for sentence in tqdm(preprocessed_essays): # for each review/sentence
    vector = np.zeros(300) # as word vectors are of zero length
    tf_idf_weight =0; # num of words with a valid vector in the sentence/review
    for word in sentence.split(): # for each word in a review/sentence
        if (word in glove words) and (word in tfidf words):
            vec = model[word] # getting the vector for each word
            # here we are multiplying idf value(dictionary[word]) and the tf value((sen
tence.count(word)/len(sentence.split())))
            tf_idf = dictionary[word]*(sentence.count(word)/len(sentence.split())) # ge
tting 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]))
```

100%|

| 50000/50000 [03:02<00:00, 273.46it/s]

50000 300

In [40]:

Similarly you can vectorize for title also

In [41]:

```
# average Word2Vec
# compute average word2vec for each review.
tfidf_w2v_vectors_titles = []; # the avg-w2v for each sentence/review is stored in this
List
for sentence in tqdm(preprocessed_titles): # for each review/sentence
    vector = np.zeros(300) # as word vectors are of zero length
    tf_idf_weight =0; # num of words with a valid vector in the sentence/review
    for word in sentence.split(): # for each word in a review/sentence
        if (word in glove_words) and (word in tfidf_words):
            vec = model[word] # getting the vector for each word
            # here we are multiplying idf value(dictionary[word]) and the tf value((sen
tence.count(word)/len(sentence.split())))
            tf_idf = dictionary[word]*(sentence.count(word)/len(sentence.split())) # ge
tting 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_titles.append(vector)
print(len(tfidf_w2v_vectors_titles))
print(len(tfidf w2v vectors titles[0]))
```

```
100%| 50000/50000 [00:02<00:00, 17749.10it/s]
```

1.5.3 Vectorizing Numerical features

In [42]:

300

```
price_data = resource_data.groupby('id').agg({'price':'sum', 'quantity':'sum'}).reset_i
ndex()
project_data = pd.merge(project_data, price_data, on='id', how='left')
```

In [43]:

```
# check this one: https://www.youtube.com/watch?v=0HOqOcln3Z4&t=530s
# standardization sklearn: https://scikit-learn.org/stable/modules/generated/sklearn.pr
eprocessing.StandardScaler.html
from sklearn.preprocessing import StandardScaler
# price_standardized = standardScalar.fit(project_data['price'].values)
# this will rise the error
# ValueError: Expected 2D array, got 1D array instead: array=[725.05 213.03 329.
       287.73 5.5 ].
# Reshape your data either using array.reshape(-1, 1)
price scalar = StandardScaler()
price_scalar.fit(project_data['price'].values.reshape(-1,1)) # finding the mean and sta
ndard deviation of this data
print(f"Mean : {price_scalar.mean_[0]}, Standard deviation : {np.sqrt(price_scalar.var_
[0])}")
# Now standardize the data with above maen and variance.
price_standardized = price_scalar.transform(project_data['price'].values.reshape(-1, 1
))
```

Mean: 299.33367619999996, Standard deviation: 378.20927190421384

In [44]:

```
price_standardized

Out[44]:
array([[-0.38268146],
```

1.5.4 Merging all the above features

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

In [45]:

```
print(categories_one_hot.shape)
print(sub_categories_one_hot.shape)
print(text_bow.shape)
print(price_standardized.shape)

(50000, 9)
(50000, 30)
(50000, 12101)
(50000, 1)
```

In [46]:

```
# 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 dense matirx
:)
X = hstack((categories_one_hot, sub_categories_one_hot, text_bow, price_standardized))
X.shape
```

Out[46]:

(50000, 12141)

In [47]:

Computing Sentiment Scores

In [48]:

```
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 multi
ple intelligences i use a wide range\
of techniques to help all my students succeed students in my class come from a variety
of different backgrounds which makes\
for wonderful sharing of experiences and cultures including native americans our school
is a caring community of successful \
learners which can be seen through collaborative student project based learning in and
out of the classroom kindergarteners \
in my class love to work with hands on materials and have many different opportunities
to practice a skill before it is\
mastered having the social skills to work cooperatively with friends is a crucial aspec
t of the kindergarten curriculum\
montana is the perfect place to learn about agriculture and nutrition my students love
to role play in our pretend kitchen\
in the early childhood classroom i have had several kids ask me can we try cooking with
real food i will take their idea \
and create common core cooking lessons where we learn important math and writing concep
ts while cooking delicious healthy \
food for snack time my students will have a grounded appreciation for the work that wen
t into making the food and knowledge \
of where the ingredients came from as well as how it is healthy for their bodies this p
roject would expand our learning of \
nutrition and agricultural cooking recipes by having us peel our own apples to make hom
emade applesauce make our own bread \
and mix up healthy plants from our classroom garden in the spring we will also create o
ur own cookbooks to be printed and \
shared with families students will gain math and literature skills as well as a life lo
ng 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,

Assignment 8: DT

1. Apply Decision Tree Classifier(DecisionTreeClassifier) on these feature sets

- Set 1: categorical, numerical features + project_title(BOW) + preprocessed_eassay (BOW)
- Set 2: categorical, numerical features + project_title(TFIDF)+ preprocessed_eassay (TFIDF)
- Set 3: categorical, numerical features + project_title(AVG W2V)+ preprocessed_eassay (AVG W2V)
- Set 4: categorical, numerical features + project_title(TFIDF W2V)+ preprocessed_eassay (TFIDF W2V)

2. Hyper paramter tuning (best `depth` in range [1, 5, 10, 50, 100, 500, 100], and the best `min_samples_split` in range [5, 10, 100, 500])

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

3. Graphviz

- Visualize your decision tree with Graphviz. It helps you to understand how a decision is being made, given a new vector.
- Since feature names are not obtained from word2vec related models, visualize only BOW & TFIDF decision trees using Graphviz
- Make sure to print the words in each node of the decision tree instead of printing its index.
- Just for visualization purpose, limit max_depth to 2 or 3 and either embed the generated images of graphviz in your notebook, or directly upload them as .png files.

4. 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>
 (https://www.appliedaicourse.com/course/applied-ai-course-online/lessons/confusion-matrix-tpr-fpr-fnr-tnr-1/) with predicted and original labels of test data points



- Once after you plot the confusion matrix with the test data, get all the `false positive data points`
 - Plot the WordCloud <u>WordCloud (https://www.geeksforgeeks.org/generating-word-cloud-python/)</u>
 - Plot the box plot with the 'price' of these 'false positive data points'
 - Plot the pdf with the `teacher_number_of_previously_posted_projects` of these `false positive data points`

5. **[Task-2]**

Select 5k best features from features of Set 2 using <u>`feature_importances_` (https://scikit-learn.org/stable/modules/generated/sklearn.tree.DecisionTreeClassifier.html</u>), discard all the other remaining features and then apply any of the model of you choice i.e. (Dession tree, Logistic

Regression, Linear SVM), you need to do hyperparameter tuning corresponding to the model you selected and procedure in step 2 and step 3

6. Conclusion

You need to summarize the results at the end of the notebook, summarize it in the table format. To
print out a table please refer to this prettytable library <u>link (http://zetcode.com/python/prettytable/)</u>



2. Decision Tree

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

In [49]:

```
# please write all the code with proper documentation, and proper titles for each subse
ction
# 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 debugging your
code
# when you plot any graph make sure you use
# a. Title, that describes your plot, this will be very helpful to the reader
# b. Legends if needed
# c. X-axis label
# d. Y-axis label
```

In [50]:

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

In [51]:

```
from sklearn.model_selection import train_test_split

x_train, x_test, y_train, y_test = train_test_split(x, y, test_size=0.33, stratify=y)
print(x_train.shape)
print(y_train.shape)
print(y_train.shape)
print(y_test.shape)

(33500, 19)
(16500, 19)
(33500,)
(16500,)
```

2.2 Make Data Model Ready: encoding numerical, categorical features

In [52]:

```
# please write all the code with proper documentation, and proper titles for each subse
ction
# 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 debugging 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 reader
# b. Legends if needed
# c. X-axis label
# d. Y-axis label
```

In [53]:

#Here I'm not encoding numerical features because in decision tress, there's no need to normalize/standardize data.
#Reshaping numerical features and printing their shapes

In [54]:

```
#Price

x_train_price = x_train['price'].values.reshape(1, -1)
x_test_price = x_test['price'].values.reshape(1, -1)

print(x_train_price.shape)
print(x_test_price.shape)
```

(1, 33500)
(1, 16500)

In [55]:

```
#teacher_number_of_previously_posted_projects

x_train_previous_projects = x_train['teacher_number_of_previously_posted_projects'].val
ues.reshape(1, -1)
x_test_previous_projects = x_test['teacher_number_of_previously_posted_projects'].value
s.reshape(1, -1)

print(x_train_previous_projects.shape)
print(x_test_previous_projects.shape)
```

(1, 33500)
(1, 16500)

In [56]:

```
#quantity
x_train_quantity = x_train['quantity'].values.reshape(1, -1)
x_test_quantity = x_test['quantity'].values.reshape(1, -1)

print(x_train_quantity.shape)
print(x_test_quantity.shape)

(1, 33500)
(1, 16500)
```

In [57]:

#Encoding Categorical Features

In [109]:

```
#school_state
state_vectorizer = CountVectorizer()
state_vectorizer.fit(x_train['school_state'].values)

x_train_state_one_hot = state_vectorizer.transform(x_train['school_state'].values)
x_test_state_one_hot = state_vectorizer.transform(x_test['school_state'].values)

print("After Vectorizations:")
print(x_train_state_one_hot.shape, y_train.shape)
print(x_test_state_one_hot.shape, y_test.shape)
print(state_vectorizer.get_feature_names())
```

```
After Vectorizations:
(33500, 51) (33500,)
(16500, 51) (16500,)
['ak', 'al', 'ar', 'az', 'ca', 'co', 'ct', 'dc', 'de', 'fl', 'ga', 'hi',
'ia', 'id', 'il', 'in', 'ks', 'ky', 'la', 'ma', 'md', 'me', 'mi', 'mn', 'm
o', 'ms', 'mt', 'nc', 'nd', 'ne', 'nh', 'nj', 'nm', 'nv', 'ny', 'oh', 'o
k', 'or', 'pa', 'ri', 'sc', 'sd', 'tn', 'tx', 'ut', 'va', 'vt', 'wa', 'w
i', 'wv', 'wy']
```

In [110]:

```
#teacher_prefix

teacher_vectorizer = CountVectorizer()
teacher_vectorizer.fit(x_train['teacher_prefix'].values.astype('U'))

x_train_teacher_one_hot = teacher_vectorizer.transform(x_train['teacher_prefix'].values
.astype('U'))

x_test_teacher_one_hot = teacher_vectorizer.transform(x_test['teacher_prefix'].values.a
stype('U'))

print("After Vectorizations:")
print(x_train_teacher_one_hot.shape, y_train.shape)
print(x_test_teacher_one_hot.shape, y_test.shape)
```

```
After Vectorizations: (33500, 5) (33500,) (16500, 5) (16500,)
```

In [111]:

```
#Project_grade_category
grade vectorizer = CountVectorizer()
grade_vectorizer.fit(x_train['project_grade_category'].values)
x_train_grade_one_hot = grade_vectorizer.transform(x_train['project_grade_category'].va
x_test_grade_one_hot = grade_vectorizer.transform(x_test['project_grade_category'].valu
es)
print("After Vectorizations:")
print(x_train_grade_one_hot.shape, y_train.shape)
print(x_test_grade_one_hot.shape, y_test.shape)
After Vectorizations:
(33500, 4) (33500,)
(16500, 4) (16500,)
In [112]:
#project_subject_categories
categories_vectorizer = CountVectorizer()
categories_vectorizer.fit(x_train['clean_categories'].values)
```

x_train_categories_one_hot = categories_vectorizer.transform(x_train['clean_categories'

x_test_categories_one_hot = categories_vectorizer.transform(x_test['clean_categories'].

print(x_test_categories_one_hot.shape, y_test.shape) After Vectorizations: (33500, 9) (33500,)

(16500, 9) (16500,)

print("After Vectorizations:")

print(x_train_categories_one_hot.shape, y_train.shape)

1.values)

values)

In [113]:

```
#project_subject_subcategories

subcategories_vectorizer = CountVectorizer()
subcategories_vectorizer.fit(x_train['clean_subcategories'].values)

x_train_subcategories_one_hot = subcategories_vectorizer.transform(x_train['clean_subcategories'].values)

x_test_subcategories_one_hot = subcategories_vectorizer.transform(x_test['clean_subcategories'].values)

print("After Vectorizations:")
print(x_train_subcategories_one_hot.shape, y_train.shape)
print(x_test_subcategories_one_hot.shape, y_test.shape)

After Vectorizations:
(33500, 30) (33500,)
(16500, 30) (16500,)
```

2.3 Make Data Model Ready: encoding eassay, and project_title

In [63]:

BOW: title, essay

In [114]:

```
#Project title
title_bow_vectorizer = CountVectorizer(min_df=10, ngram_range=(2, 2), max_features=5000
title_bow_vectorizer.fit(x_train['project_title'].values)
x_train_title_bow = title_bow_vectorizer.transform(x_train['project_title'].values)
x_test_title_bow = title_bow_vectorizer.transform(x_test['project_title'].values)
print("After Vectorizations:")
print(x_train_title_bow.shape, y_train.shape)
print(x_test_title_bow.shape, y_test.shape)
After Vectorizations:
(33500, 769) (33500,)
(16500, 769) (16500,)
In [115]:
#Essay
essay bow vectorizer = CountVectorizer(min df=10, ngram range=(2, 2), max features=5000
essay_bow_vectorizer.fit(x_train['essay'].values)
x_train_essay_bow = essay_bow_vectorizer.transform(x_train['essay'].values)
```

x_test_essay_bow = essay_bow_vectorizer.transform(x_test['essay'].values)

```
After Vectorizations:
(33500, 5000) (33500,)
(16500, 5000) (16500,)
```

print("After Vectorizations:")

print(x_train_essay_bow.shape, y_train.shape) print(x_test_essay_bow.shape, y_test.shape)

TFIDF: ttitle, essay

In [145]:

```
#project_title
title_tfidf_vectorizer = TfidfVectorizer(min_df=10, ngram_range=(2, 2), max_features=50
00)
title_tfidf_vectorizer.fit(x_train['project_title'].values)
x_train_title_tfidf = title_tfidf_vectorizer.transform(x_train['project_title'].values)
x_test_title_tfidf = title_tfidf_vectorizer.transform(x_test['project_title'].values)
print("After Vectorizations:")
print(x_train_title_tfidf.shape, y_train.shape)
print(x_test_title_tfidf.shape, y_test.shape)
After Vectorizations:
(33500, 769) (33500,)
```

(16500, 769) (16500,)

In [116]:

```
#essay
from sklearn.feature_extraction.text import TfidfVectorizer
essay_tfidf_vectorizer = TfidfVectorizer(min_df=10, ngram_range=(2, 2), max_features=50
00)
essay_tfidf_vectorizer.fit(x_train['essay'].values)

x_train_essay_tfidf = essay_tfidf_vectorizer.transform(x_train['essay'].values)

x_test_essay_tfidf = essay_tfidf_vectorizer.transform(x_test['essay'].values)

print("After Vectorizations:")
print(x_train_essay_tfidf.shape, y_train.shape)
print(x_test_essay_tfidf.shape, y_test.shape)

After Vectorizations:
(33500, 5000) (33500,)
(16500, 5000) (16500,)
```

AVG W2V: ttitle, essay

In [68]:

```
with open('glove_vectors', 'rb') as f:
  model = pickle.load(f)
  glove_words = set(model.keys())
```

Project_title

In [69]:

```
x_train_title_avg_w2v = []; # the avg-w2v for each sentence/review is stored in this li
st

for sentence in tqdm(x_train['project_title'].values): # 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
        x_train_title_avg_w2v.append(vector)

print(len(x_train_title_avg_w2v))
print(len(x_train_title_avg_w2v[0]))
```

```
100%
```

| 33500/33500 [00:00<00:00, 38280.78it/s]

33500 300

In [70]:

```
x_test_title_avg_w2v = []; # the avg-w2v for each sentence/review is stored in this lis
t
for sentence in tqdm(x_test['project_title'].values): # 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
    x_test_title_avg_w2v.append(vector)

print(len(x_test_title_avg_w2v))
    print(len(x_test_title_avg_w2v[0]))
```

100%|

| 16500/16500 [00:00<00:00, 33589.94it/s]

16500 300

Essay

In [71]:

```
# average Word2Vec
# compute average word2vec for each review.
x train essay avg w2v = []; # the avg-w2v for each sentence/review is stored in this li
st
for sentence in tqdm(x_train['essay'].values): # 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
    x_train_essay_avg_w2v.append(vector)
print(len(x train essay avg w2v))
print(len(x train essay avg w2v[0]))
```

100%|

| 33500/33500 [00:17<00:00, 1872.58it/s]

33500 300

In [72]:

```
x_test_essay_avg_w2v = []; # the avg-w2v for each sentence/review is stored in this lis
t

for sentence in tqdm(x_test['essay'].values): # 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
        x_test_essay_avg_w2v.append(vector)
```

100%

| 16500/16500 [00:09<00:00, 1831.98it/s]

TFIDF W2V: ttitle, essay

Project_title

In [73]:

```
#project_title

tfidf_model = TfidfVectorizer()
  tfidf_model.fit(x_train['project_title'])
# 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 [74]:

```
x train title tfidf w2v = []; # the avg-w2v for each sentence/review is stored in this
list
for sentence in tqdm(x_train['project_title']): # for each review/sentence
    vector = np.zeros(300) # as word vectors are of zero Length
    tf_idf_weight =0; # num of words with a valid vector in the sentence/review
    for word in sentence.split(): # for each word in a review/sentence
        if (word in glove_words) and (word in tfidf_words):
            vec = model[word] # getting the vector for each word
            # here we are multiplying idf value(dictionary[word]) and the tf value((sen
tence.count(word)/len(sentence.split())))
            tf_idf = dictionary[word]*(sentence.count(word)/len(sentence.split())) # ge
tting 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
    x train title tfidf w2v.append(vector)
print(len(x_train_title_tfidf_w2v))
```

100%

33500/33500 [00:02<00:00, 16147.76it/s]

33500

In [75]:

```
x_{test_{title}} tfidf_w2v = []; # the avg-<math>w2v for each sentence/review is stored in this l
ist
for sentence in tqdm(x_test['project_title']): # for each review/sentence
    vector = np.zeros(300) # as word vectors are of zero length
    tf_idf_weight =0; # num of words with a valid vector in the sentence/review
    for word in sentence.split(): # for each word in a review/sentence
        if (word in glove_words) and (word in tfidf_words):
            vec = model[word] # getting the vector for each word
            # here we are multiplying idf value(dictionary[word]) and the tf value((sen
tence.count(word)/len(sentence.split())))
            tf idf = dictionary[word]*(sentence.count(word)/len(sentence.split())) # qe
tting 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
    x_test_title_tfidf_w2v.append(vector)
```

100%

16500/16500 [00:00<00:00, 18058.84it/s]

Essay

In [76]:

```
#essay

tfidf_model = TfidfVectorizer()

tfidf_model.fit(x_train['essay'])

# 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 [77]:

```
x_train_essay_tfidf_w2v = []; # the avg-w2v for each sentence/review is stored in this
list
for sentence in tqdm(x_train['essay']): # for each review/sentence
    vector = np.zeros(300) # as word vectors are of zero length
    tf_idf_weight =0; # num of words with a valid vector in the sentence/review
    for word in sentence.split(): # for each word in a review/sentence
        if (word in glove_words) and (word in tfidf_words):
            vec = model[word] # getting the vector for each word
            # here we are multiplying idf value(dictionary[word]) and the tf value((sen
tence.count(word)/len(sentence.split())))
            tf_idf = dictionary[word]*(sentence.count(word)/len(sentence.split())) # ge
tting 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
    x train essay tfidf w2v.append(vector)
print(len(x_train_essay_tfidf_w2v))
```

100%|

| 33500/33500 [02:02<00:00, 272.46it/s]

33500

In [78]:

```
x_test_essay_tfidf_w2v = []; # the avg-w2v for each sentence/review is stored in this l
ist
for sentence in tqdm(x test['essay']): # for each review/sentence
    vector = np.zeros(300) # as word vectors are of zero length
    tf idf weight =0; # num of words with a valid vector in the sentence/review
    for word in sentence.split(): # for each word in a review/sentence
        if (word in glove_words) and (word in tfidf_words):
            vec = model[word] # getting the vector for each word
            # here we are multiplying idf value(dictionary[word]) and the tf value((sen
tence.count(word)/len(sentence.split())))
            tf_idf = dictionary[word]*(sentence.count(word)/len(sentence.split())) # ge
tting 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
    x test essay tfidf w2v.append(vector)
```

100%

16500/16500 [01:01<00:00, 267.83it/s]

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

Apply Decision Tree 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

In [79]:

```
# please write all the code with proper documentation, and proper titles for each subse
ction
# 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 debugging your
code
# when you plot any graph make sure you use
# a. Title, that describes your plot, this will be very helpful to the reader
# b. Legends if needed
# c. X-axis label
# d. Y-axis label
```

In [80]:

```
#Before merging, re-shape some features. if we dont reshape, we'll get error
#Re-shaping
x_train_price = x_train_price.reshape(-1,1)
x_train_previous_projects = x_train_previous_projects.reshape(-1,1)
x_train_quantity = x_train_quantity.reshape(-1, 1)

x_test_price = x_test_price.reshape(-1,1)
x_test_previous_projects = x_test_previous_projects.reshape(-1,1)
x_test_quantity = x_test_quantity.reshape(-1, 1)
```

2.4.1 Applying Decision Trees on BOW, SET 1

In []:

```
# Please write all the code with proper documentation
```

In [82]:

In [83]:

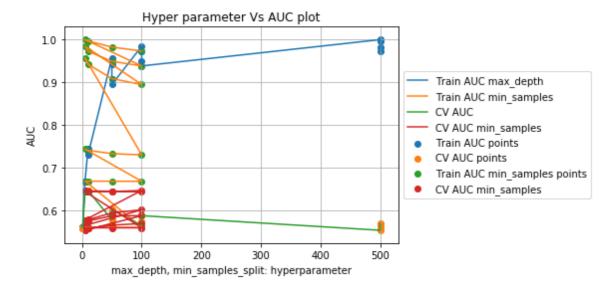
```
from sklearn.model_selection import GridSearchCV
from sklearn.tree import DecisionTreeClassifier
tuned\_parameters = \{ 'max\_depth': [1, 5, 10], 'min\_samples\_split': [5, 10, 50] \}
DTC = DecisionTreeClassifier(class_weight='balanced')
clf = GridSearchCV(DTC, tuned_parameters, cv=10, scoring='roc_auc')
clf.fit(x_train_bow, y_train)
results = pd.DataFrame.from_dict(clf.cv_results_)
max_depth_results = results.sort_values(['param_max_depth'])
train_auc= results['mean_train_score']
train_auc_std= results['std_train_score']
cv_auc = results['mean_test_score']
cv_auc_std= results['std_test_score']
max_depth = results['param_max_depth']
plt.plot(max_depth, train_auc, label='Train AUC')
# this code is copied from here: https://stackoverflow.com/a/48803361/4084039
# plt.gca().fill_between(K, train_auc - train_auc_std,train_auc + train_auc_std,alpha=
0.2,color='darkblue')
plt.plot(max_depth, cv_auc, label='CV AUC')
# this code is copied from here: https://stackoverflow.com/a/48803361/4084039
# plt.gca().fill_between(K, cv_auc - cv_auc_std,cv_auc + cv_auc_std,alpha=0.2,color='da
rkorange')
plt.scatter(max_depth, train_auc, label='Train AUC points')
plt.scatter(max_depth, cv_auc, label='CV AUC points')
plt.legend()
plt.xlabel("max_depth: hyperparameter")
plt.ylabel("AUC")
plt.title("Hyper parameter Vs AUC plot")
plt.grid()
plt.show()
results.head()
```

Out[83]:

'\nfrom sklearn.model selection import GridSearchCV\nfrom sklearn.tree imp ort DecisionTreeClassifier\n\ntuned parameters = {\'max depth\':[1, 5, 1 0], \'min_samples_split\':[5, 10, 50]}\nDTC = DecisionTreeClassifier(class _weight=\'balanced\')\nclf = GridSearchCV(DTC, tuned_parameters, cv=10, sc oring=\'roc_auc\')\nclf.fit(x_train_bow, y_train)\n\nresults = pd.DataFram e.from_dict(clf.cv_results_)\nmax_depth_results = results.sort_values([\'p aram_max_depth\'])\n\ntrain_auc= results[\'mean_train_score\']\ntrain_auc_ std= results[\'std_train_score\']\ncv_auc = results[\'mean_test_score\'] \ncv_auc_std= results[\'std_test_score\']\nmax_depth = results[\'param_ma x_depth\']\n\nplt.plot(max_depth, train_auc, label=\'Train AUC\')\n# this code is copied from here: https://stackoverflow.com/a/48803361/4084039\n# plt.gca().fill_between(K, train_auc - train_auc_std,train_auc + train_auc_ std,alpha=0.2,color=\'darkblue\')\n\nplt.plot(max_depth, cv_auc, label= \'CV AUC\')\n# this code is copied from here: https://stackoverflow.com/a/ 48803361/4084039\n# plt.gca().fill_between(K, cv_auc - cv_auc_std,cv_auc + cv_auc_std,alpha=0.2,color=\'darkorange\')\n\nplt.scatter(max_depth, train _auc, label=\'Train AUC points\')\nplt.scatter(max_depth, cv_auc, label= \'CV AUC points\')\n\nplt.legend()\nplt.xlabel("max_depth: hyperparamete r")\nplt.ylabel("AUC")\nplt.title("Hyper parameter Vs AUC plot")\nplt.grid ()\nplt.show()\n\nresults.head()\n'

In [84]:

```
from sklearn.model selection import GridSearchCV
from sklearn.tree import DecisionTreeClassifier
tuned parameters = {'max depth':[1, 5, 10, 50, 100, 500], 'min samples split':[5, 10, 5
0, 100]}
DTC = DecisionTreeClassifier(class_weight='balanced')
clf = GridSearchCV(DTC, tuned_parameters, cv=10, scoring='roc_auc')
clf.fit(x_train_bow, y_train)
results = pd.DataFrame.from dict(clf.cv results )
max_depth_results = results.sort_values(['param_max_depth'])
min_samples_split_results = results.sort_values(['param_min_samples_split'])
#For max depth
train_auc_max_depth= results['mean_train_score']
train_auc_std_max_depth= results['std_train_score']
cv_auc_max_depth = results['mean_test_score']
cv_auc_std_max_depth= results['std_test_score']
max_depth = results['param_max_depth']
#For min_samples_split
train_auc_min_samples = results['mean_train_score']
train auc std min samples = results['std train score']
cv_auc_min_samples = results['mean_test_score']
cv_auc_std_min_samples = results['std_test_score']
min_samples_split = results['param_min_samples_split']
plt.plot(max_depth, train_auc_max_depth, label='Train AUC max_depth')
# this code is copied from here: https://stackoverflow.com/a/48803361/4084039
# plt.gca().fill_between(K, train_auc - train_auc_std,train_auc + train_auc_std,alpha=
0.2, color='darkblue')
plt.plot(min samples split, train auc min samples, label='Train AUC min samples') #Plot
ting min samples-split curve
plt.plot(max_depth, cv_auc_max_depth, label='CV AUC')
# this code is copied from here: https://stackoverflow.com/a/48803361/4084039
# plt.gca().fill_between(K, cv_auc - cv_auc_std,cv_auc + cv_auc_std,alpha=0.2,color='da
rkorange')
plt.plot(min samples split, cv auc min samples, label='CV AUC min samples')
plt.scatter(max depth, train auc max depth, label='Train AUC points')
plt.scatter(max_depth, cv_auc_max_depth, label='CV AUC points')
plt.scatter(min samples split, train auc min samples, label=('Train AUC min samples poi
nts'))
plt.scatter(min samples split, cv auc min samples, label=('CV AUC min samples'))
plt.legend(loc='center left', bbox_to_anchor=(1, 0.5))
plt.xlabel("max depth, min samples split: hyperparameter")
plt.ylabel("AUC")
plt.title("Hyper parameter Vs AUC plot")
plt.grid()
plt.show()
results.head()
```



Out[84]:

	mean_fit_time	std_fit_time	mean_score_time	std_score_time	param_max_depth	param_n
0	0.142558	0.029244	0.008374	0.001310	1	
1	0.136062	0.013681	0.008222	0.000558	1	
2	0.144735	0.030765	0.009876	0.003091	1	
3	0.136866	0.020790	0.008470	0.001016	1	
4	0.669828	0.029657	0.008010	0.003563	5	

5 rows × 32 columns

→

In [85]:

clf.best_params_

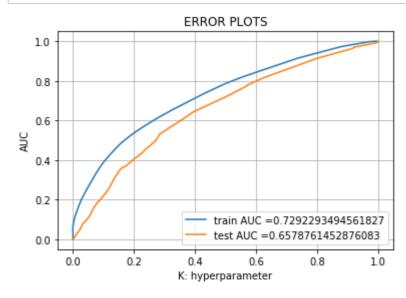
Out[85]:

{'max_depth': 10, 'min_samples_split': 100}

In []:

In [86]:

```
from sklearn.metrics import roc_curve, auc
DTC = DecisionTreeClassifier(max_depth=5, min_samples_split=100, class_weight='balance
d')
DTC.fit(x_train_bow, y_train)
y_train_pred = clf.predict_proba(x_train_bow)[:, 1]
y_test_pred = clf.predict_proba(x_test_bow)[:, 1]
train fpr, train_tpr, tr_thresholds = roc_curve(y_train, y_train_pred)
test_fpr, test_tpr, te_thresholds = roc_curve(y_test, y_test_pred)
plt.grid()
plt.plot(train_fpr, train_tpr, label="train AUC ="+str(auc(train_fpr, train_tpr)))
plt.plot(test_fpr, test_tpr, label="test AUC ="+str(auc(test_fpr, test_tpr)))
plt.legend()
plt.xlabel("K: hyperparameter")
plt.ylabel("AUC")
plt.title("ERROR PLOTS")
plt.show()
```



In [87]:

```
# we are writing our own function for predict, with defined thresould
# we will pick a threshold that will give the least fpr
def find_best_threshold(threshould, fpr, tpr):
    t = threshould[np.argmax(tpr*(1-fpr))]
    # (tpr*(1-fpr)) will be maximum if your fpr is very low and tpr is very high
    print("the maximum value of tpr*(1-fpr)", max(tpr*(1-fpr)), "for threshold", np.rou
nd(t,3))
    return t
def predict with best t(proba, threshould):
    predictions = []
    for i in proba:
        if i>=threshould:
            predictions.append(1)
        else:
            predictions.append(0)
    return predictions
```

In [88]:

```
#Confusion matrix

from sklearn.metrics import confusion_matrix
best_t = find_best_threshold(tr_thresholds, train_fpr, train_tpr)
print("Train confusion matrix")
print(confusion_matrix(y_train, predict_with_best_t(y_train_pred, best_t)))
print("Test confusion matrix")
print(confusion_matrix(y_test, predict_with_best_t(y_test_pred, best_t)))

the maximum value of tpr*(1-fpr) 0.4425887798107272 for threshold 0.485
Train confusion matrix
[[ 3750 1418]
```

```
Train confusion matrix
[[ 3750     1418]
     [11051     17281]]
Test confusion matrix
[[1655     891]
     [5634     8320]]
```

In [89]:

```
import seaborn as sns

#Train Confusion matrix

#Reference- https://www.kaggle.com/agungor2/various-confusion-matrix-plots

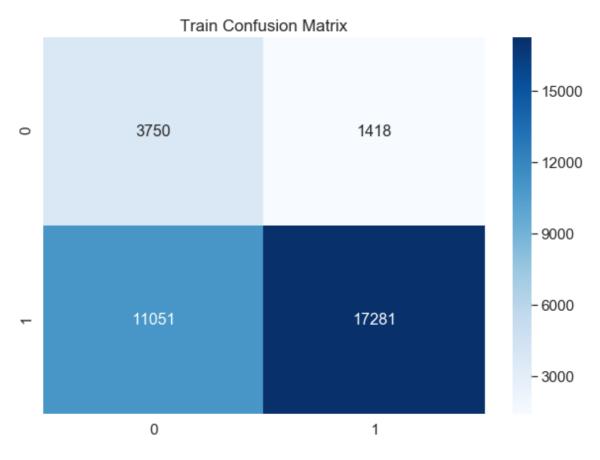
y_train_predicted= predict_with_best_t(y_train_pred, best_t)

df_cm = pd.DataFrame(confusion_matrix(y_train,y_train_predicted), columns=np.unique(y_t rain), index = np.unique(y_train))

plt.figure(figsize = (10,7))
    sns.set(font_scale=1.4)#for label size
    plt.title('Train Confusion Matrix')
    sns.heatmap(df_cm, cmap="Blues", annot=True, fmt='g', annot_kws={"size": 16})
```

Out[89]:

<matplotlib.axes._subplots.AxesSubplot at 0x1b55ca55940>



In [90]:

```
train_fpr = []
for i in range(len(y_train)):
    if y_train[i] == 0 and y_train_predicted[i] == 1:
        train_fpr.append(y_train[i])
len(train_fpr)
```

Out[90]:

1418

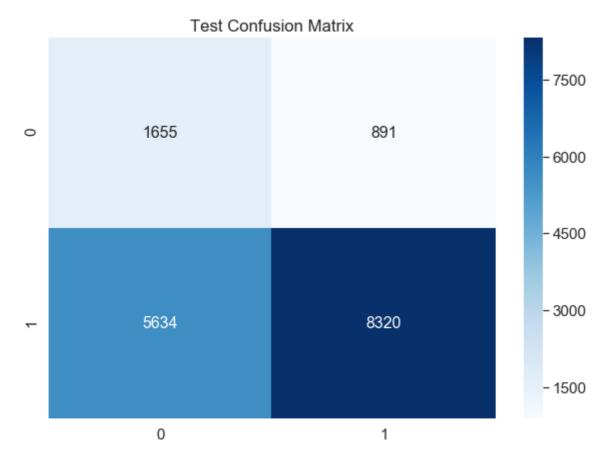
In [91]:

```
#Test Confusion matrix
y_test_predicted=predict_with_best_t(y_test_pred, best_t)
df_cm = pd.DataFrame(confusion_matrix(y_test,y_test_predicted ), columns=np.unique(y_test), index = np.unique(y_test))

plt.figure(figsize = (10,7))
sns.set(font_scale=1.4)#for label size
plt.title('Test Confusion Matrix')
sns.heatmap(df_cm, cmap="Blues", annot=True, fmt='g', annot_kws={"size": 16})
```

Out[91]:

<matplotlib.axes._subplots.AxesSubplot at 0x1b55ca55f28>



In [92]:

```
#Plotting wordcloud using essay
```

In [93]:

```
test_fpr = []
for i in range(len(y_test)):
    if y_test[i]==0 and y_test_predicted[i]==1:
        test_fpr.append(i)
```

```
In [94]:
```

```
essay_fpr = [x_test['essay'].iloc[i] for i in test_fpr]
print(len(essay_fpr))
```

891

In [95]:

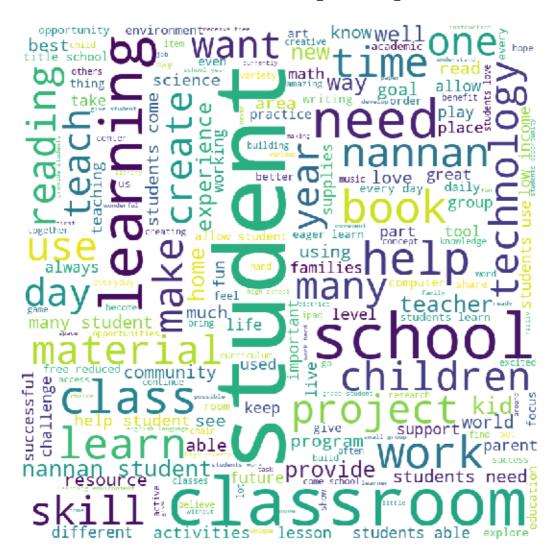
```
word=[]
for i in essay_fpr:
    word.append(i.split())
```

In [96]:

```
import nltk
from wordcloud import WordCloud, STOPWORDS
```

In [97]:

```
comment words = ' '
stopwords = set(STOPWORDS)
for val in essay_fpr:
    # typecaste each val to string
   val = str(val)
    # split the value
   tokens = val.split()
    # Converts each token into Lowercase
    for i in range(len(tokens)):
        tokens[i] = tokens[i].lower()
    for words in tokens:
        comment_words = comment_words + words + ' '
wordcloud = WordCloud(width = 800, height = 800,
                background_color ='white',
                stopwords = stopwords,
                min_font_size = 10).generate(comment_words)
# plot the WordCloud image
plt.figure(figsize = (8, 8), facecolor = None)
plt.imshow(wordcloud)
plt.axis("off")
plt.tight_layout(pad = 0)
plt.show()
```



In [98]:

#Plotting the boxplot with price

In [99]:

```
price_fpr = [x_test['price'].iloc[i] for i in test_fpr]
print(len(price_fpr))
```

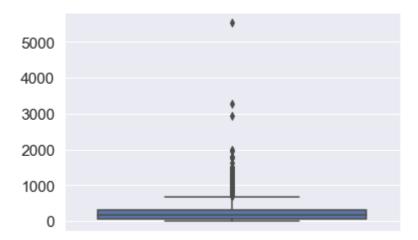
891

In [100]:

```
sns.boxplot(y=price_fpr)
```

Out[100]:

<matplotlib.axes._subplots.AxesSubplot at 0x1b52734e048>



In [101]:

#plotting the pdf with teacher_number_of_previously_posted_projects

In [102]:

```
values = x_test['teacher_number_of_previously_posted_projects'].values
teacher_previous_projects_fpr = []
for i in range(len(values)):
    if y_test[i]==0 and y_test_predicted[i]==1:
        teacher_previous_projects_fpr.append(values[i])

print(len(teacher_previous_projects_fpr))
```

891

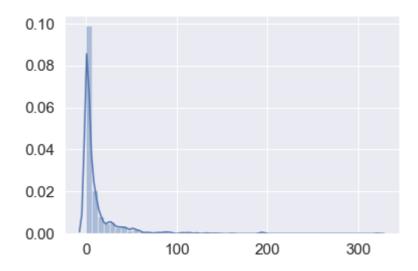
In [103]:

#Reference https://stackoverflow.com/questions/52908925/add-a-standard-normal-pdf-overa-seaborn-histogram

pdf = sns.distplot(teacher_previous_projects_fpr, norm_hist=True, kde=True)
pdf

Out[103]:

<matplotlib.axes._subplots.AxesSubplot at 0x1b524281f60>



2.4.1.1 Graphviz visualization of Decision Tree on BOW, SET 1

In [104]:

Please write all the code with proper documentation

```
In [105]:
```

In [117]:

```
#Getting feature names
features = []
features.append(['price'])
features.append(['teacher_number_of_previously_posted_projects'])
features.append(['quantity'])
features.extend(state_vectorizer.get_feature_names())
features.extend(teacher_vectorizer.get_feature_names())
features.extend(grade_vectorizer.get_feature_names())
features.extend(categories_vectorizer.get_feature_names())
features.extend(subcategories_vectorizer.get_feature_names())
features.extend(essay_bow_vectorizer.get_feature_names())
features.extend(title_bow_vectorizer.get_feature_names())
len(features)
```

Out[117]:

5871

In []:

In [120]:

```
from sklearn import tree
import pydot
from io import StringIO

#References https://stackoverflow.com/questions/27817994/visualizing-decision-tree-in-
scikit-learn

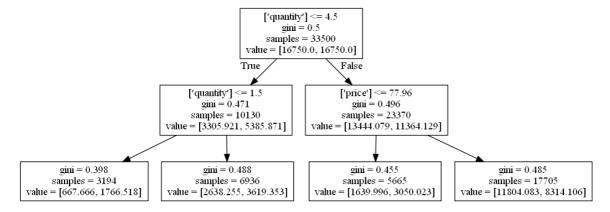
tree.export_graphviz(DTC, out_file='tree.dot') #produces dot file

dotfile = StringIO()
tree.export_graphviz(DTC,feature_names=features, out_file=dotfile)
(graph, ) = pydot.graph_from_dot_data(dotfile.getvalue())
graph.write_png("dtree1.png")
```

In [121]:

```
from IPython.display import Image
Image('dtree1.png')
```

Out[121]:



2.4.2 Applying Decision Trees on TFIDF, SET 2

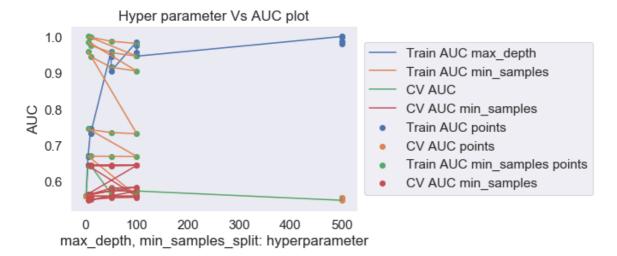
In [122]:

Please write all the code with proper documentation

In [123]:

In [124]:

```
from sklearn.model selection import GridSearchCV
from sklearn.tree import DecisionTreeClassifier
tuned parameters = {'max depth':[1, 5, 10, 50, 100, 500], 'min samples split':[5, 10, 5
0, 100]}
DTC = DecisionTreeClassifier(class_weight='balanced')
clf = GridSearchCV(DTC, tuned_parameters, cv=10, scoring='roc_auc')
clf.fit(x_train_tfidf, y_train)
results = pd.DataFrame.from dict(clf.cv results )
max_depth_results = results.sort_values(['param_max_depth'])
min_samples_split_results = results.sort_values(['param_min_samples_split'])
#For max depth
train_auc_max_depth= results['mean_train_score']
train_auc_std_max_depth= results['std_train_score']
cv_auc_max_depth = results['mean_test_score']
cv_auc_std_max_depth= results['std_test_score']
max_depth = results['param_max_depth']
#For min_samples_split
train_auc_min_samples = results['mean_train_score']
train auc std min samples = results['std train score']
cv_auc_min_samples = results['mean_test_score']
cv_auc_std_min_samples = results['std_test_score']
min_samples_split = results['param_min_samples_split']
plt.plot(max_depth, train_auc_max_depth, label='Train AUC max_depth')
# this code is copied from here: https://stackoverflow.com/a/48803361/4084039
# plt.gca().fill_between(K, train_auc - train_auc_std,train_auc + train_auc_std,alpha=
0.2, color='darkblue')
plt.plot(min samples split, train auc min samples, label='Train AUC min samples') #Plot
ting min samples-split curve
plt.plot(max_depth, cv_auc_max_depth, label='CV AUC')
# this code is copied from here: https://stackoverflow.com/a/48803361/4084039
# plt.gca().fill_between(K, cv_auc - cv_auc_std,cv_auc + cv_auc_std,alpha=0.2,color='da
rkorange')
plt.plot(min samples split, cv auc min samples, label='CV AUC min samples')
plt.scatter(max depth, train auc max depth, label='Train AUC points')
plt.scatter(max_depth, cv_auc_max_depth, label='CV AUC points')
plt.scatter(min samples split, train auc min samples, label=('Train AUC min samples poi
nts'))
plt.scatter(min samples split, cv auc min samples, label=('CV AUC min samples'))
plt.legend(loc='center left', bbox_to_anchor=(1, 0.5))
plt.xlabel("max depth, min samples split: hyperparameter")
plt.ylabel("AUC")
plt.title("Hyper parameter Vs AUC plot")
plt.grid()
plt.show()
results.head()
```



Out[124]:

	mean_fit_time	std_fit_time	mean_score_time	std_score_time	param_max_depth	param_n
0	0.340137	0.072396	0.006927	0.002456	1	
1	0.384475	0.103413	0.011899	0.003835	1	
2	0.342468	0.061538	0.008327	0.001032	1	
3	0.324950	0.032711	0.008123	0.001047	1	
4	1.440195	0.068820	0.008581	0.001148	5	

5 rows × 32 columns

In [125]:

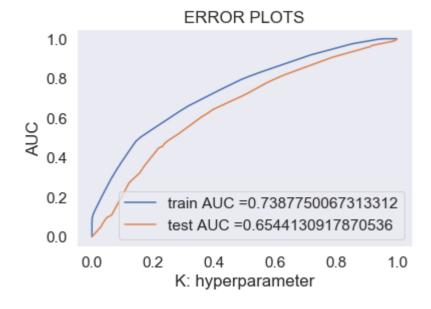
clf.best_params_

Out[125]:

{'max_depth': 10, 'min_samples_split': 5}

In [126]:

```
from sklearn.metrics import roc_curve, auc
DTC_set2 = DecisionTreeClassifier(max_depth=10, min_samples_split=5, class_weight='bala
nced')
DTC_set2.fit(x_train_tfidf, y_train)
y_train_pred = clf.predict_proba(x_train_tfidf)[:, 1]
y_test_pred = clf.predict_proba(x_test_tfidf)[:, 1]
train fpr, train tpr, tr thresholds = roc curve(y train, y train pred)
test_fpr, test_tpr, te_thresholds = roc_curve(y_test, y_test_pred)
plt.grid()
plt.plot(train_fpr, train_tpr, label="train AUC ="+str(auc(train_fpr, train_tpr)))
plt.plot(test_fpr, test_tpr, label="test AUC ="+str(auc(test_fpr, test_tpr)))
plt.legend()
plt.xlabel("K: hyperparameter")
plt.ylabel("AUC")
plt.title("ERROR PLOTS")
plt.show()
```



In [127]:

```
# we are writing our own function for predict, with defined thresould
# we will pick a threshold that will give the least fpr
def find_best_threshold(threshould, fpr, tpr):
    t = threshould[np.argmax(tpr*(1-fpr))]
    # (tpr*(1-fpr)) will be maximum if your fpr is very low and tpr is very high
    print("the maximum value of tpr*(1-fpr)", max(tpr*(1-fpr)), "for threshold", np.rou
nd(t,3))
    return t
def predict with best t(proba, threshould):
    predictions = []
    for i in proba:
        if i>=threshould:
            predictions.append(1)
        else:
            predictions.append(0)
    return predictions
```

In [128]:

```
#Confusion matrix

from sklearn.metrics import confusion_matrix
best_t = find_best_threshold(tr_thresholds, train_fpr, train_tpr)
print("Train confusion matrix")
print(confusion_matrix(y_train, predict_with_best_t(y_train_pred, best_t)))
print("Test confusion matrix")
print(confusion_matrix(y_test, predict_with_best_t(y_test_pred, best_t)))

the maximum value of tpr*(1-fpr) 0.4517061957532294 for threshold 0.5
Train confusion matrix
[[ 3642  1526]
       [10172  18160]]
Test confusion matrix
[[1584  962]
       [5271  8683]]
```

In [129]:

```
import seaborn as sns

#Train Confusion matrix

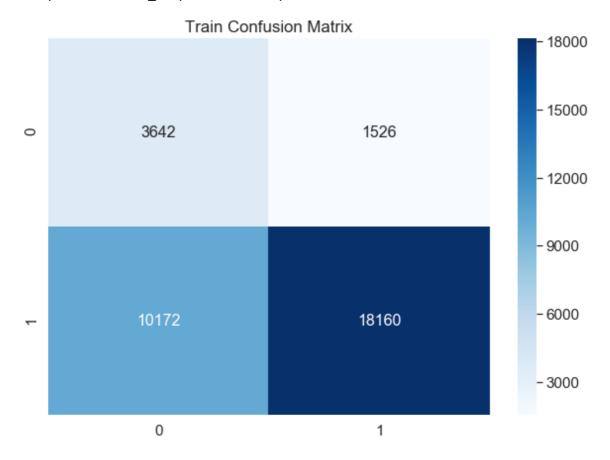
#Reference- https://www.kaggle.com/agungor2/various-confusion-matrix-plots

df_cm = pd.DataFrame(confusion_matrix(y_train, predict_with_best_t(y_train_pred, best_t)), columns=np.unique(y_train), index = np.unique(y_train))

plt.figure(figsize = (10,7))
sns.set(font_scale=1.4)#for Label size
plt.title('Train Confusion Matrix')
sns.heatmap(df_cm, cmap="Blues", annot=True, fmt='g', annot_kws={"size": 16})
```

Out[129]:

<matplotlib.axes._subplots.AxesSubplot at 0x1b55c559d30>



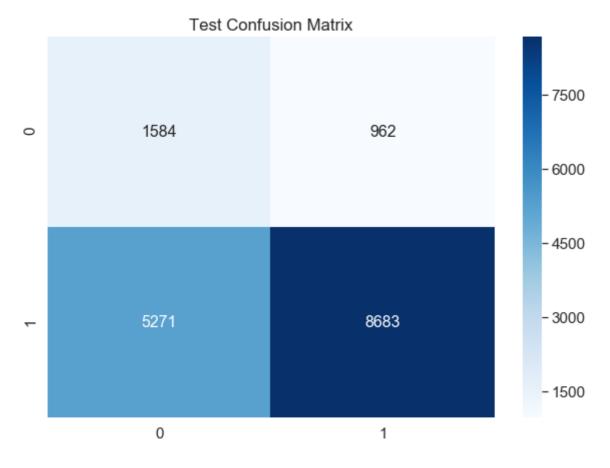
In [130]:

```
#Test Confusion matrix
df_cm = pd.DataFrame(confusion_matrix(y_test, predict_with_best_t(y_test_pred, best_t
)), columns=np.unique(y_test), index = np.unique(y_test))

plt.figure(figsize = (10,7))
sns.set(font_scale=1.4)#for label size
plt.title('Test Confusion Matrix')
sns.heatmap(df_cm, cmap="Blues", annot=True, fmt='g', annot_kws={"size": 16})
```

Out[130]:

<matplotlib.axes._subplots.AxesSubplot at 0x1b500383dd8>



In [131]:

```
#Getting all the false positive points

test_fpr = []
for i in range(len(y_test)):
    if y_test[i]==0 and y_test_predicted[i]==1:
        test_fpr.append(i)
```

In [132]:

```
#Plotting the word cloud
```

In [133]:

```
values = x_test['essay'].values
essay_fpr = []
for i in range(len(values)):
    if y_test[i]==0 and y_test_predicted[i]==1:
        essay_fpr.append(values[i])

print(len(essay_fpr))
```

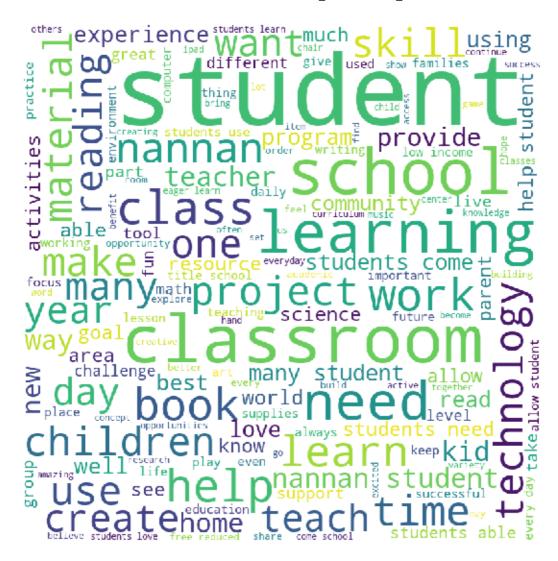
891

In [134]:

```
word=[]
for i in essay_fpr:
    word.append(i.split())
```

In [135]:

```
import nltk
from wordcloud import WordCloud, STOPWORDS
comment words = ' '
stopwords = set(STOPWORDS)
for val in essay_fpr:
    # typecaste each val to string
   val = str(val)
    # split the value
   tokens = val.split()
    # Converts each token into Lowercase
    for i in range(len(tokens)):
        tokens[i] = tokens[i].lower()
    for words in tokens:
        comment_words = comment_words + words + ' '
wordcloud = WordCloud(width = 800, height = 800,
                background_color ='white',
                stopwords = stopwords,
                min_font_size = 10).generate(comment_words)
# plot the WordCloud image
plt.figure(figsize = (8, 8), facecolor = None)
plt.imshow(wordcloud)
plt.axis("off")
plt.tight_layout(pad = 0)
plt.show()
```



In [136]:

#Plotting the boxplot with price

In [137]:

```
values = x_test['price'].values
price_fpr = []
for i in range(len(values)):
    if y_test[i]==0 and y_test_predicted[i]==1:
        price_fpr.append(values[i])
print(len(price_fpr))
```

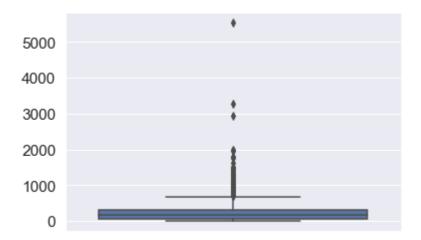
891

In [138]:

```
sns.boxplot(y=price_fpr)
```

Out[138]:

<matplotlib.axes._subplots.AxesSubplot at 0x1b524c444e0>



In [139]:

#plotting the pdf with teacher_number_of_previously_posted_projects

In [140]:

```
values = x_test['teacher_number_of_previously_posted_projects'].values
teacher_previous_projects_fpr = []
for i in range(len(values)):
    if y_test[i]==0 and y_test_predicted[i]==1:
        teacher_previous_projects_fpr.append(values[i])

print(len(teacher_previous_projects_fpr))
```

891

In [141]:

```
#Reference https://stackoverflow.com/questions/15415455/plotting-probability-density-fu
nction-by-sample-with-matplotLib
from scipy.stats.kde import gaussian_kde
from numpy import linspace

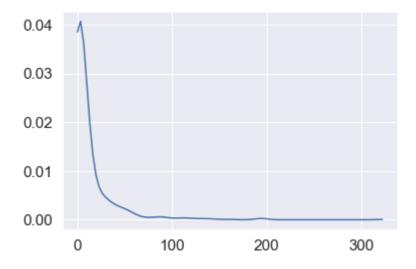
kde = gaussian_kde(teacher_previous_projects_fpr)

dist_space = linspace( min(teacher_previous_projects_fpr), max(teacher_previous_project
s_fpr), 100 )

plt.plot( dist_space, kde(dist_space) )
```

Out[141]:

[<matplotlib.lines.Line2D at 0x1b52718d278>]



In []:

2.4.2.1 Graphviz visualization of Decision Tree on TFIDF, SET 2

In [142]:

Please write all the code with proper documentation

In [143]:

In [146]:

```
#getting feature names
features = []

features.append(['price'])
features.append(['teacher_number_of_previously_posted_projects'])
features.append(['quantity'])
features.extend(state_vectorizer.get_feature_names())
features.extend(teacher_vectorizer.get_feature_names())
features.extend(grade_vectorizer.get_feature_names())
features.extend(categories_vectorizer.get_feature_names())
features.extend(subcategories_vectorizer.get_feature_names())
features.extend(essay_tfidf_vectorizer.get_feature_names())
features.extend(title_tfidf_vectorizer.get_feature_names())
len(features)
```

Out[146]:

5871

In [150]:

```
import pydot
from io import StringIO
from sklearn import tree

#References https://stackoverflow.com/questions/27817994/visualizing-decision-tree-in-scikit-learn

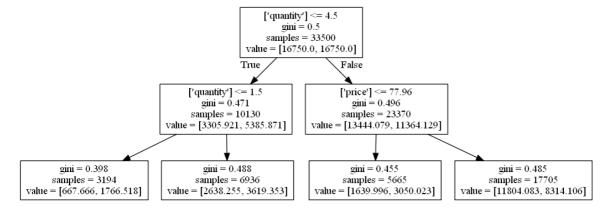
tree.export_graphviz(DTC, out_file='tree.dot')

dotfile = StringIO()
tree.export_graphviz(DTC, feature_names=features, out_file=dotfile)
(graph, ) = pydot.graph_from_dot_data(dotfile.getvalue())
graph.write_png("dtree2.png")
```

In [151]:

```
from IPython.display import Image
Image('dtree2.png')
```

Out[151]:



2.4.3 Applying Decision Trees on AVG W2V, SET 3

In [152]:

Please write all the code with proper documentation

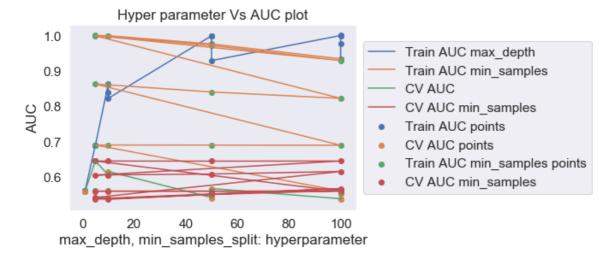
In [153]:

In [156]:

```
from sklearn.model selection import GridSearchCV
from sklearn.tree import DecisionTreeClassifier
tuned_parameters = {'max_depth':[1, 5, 10, 50, 100], 'min_samples_split':[5, 10, 50, 10
0]}
DTC = DecisionTreeClassifier(class_weight='balanced')
clf = GridSearchCV(DTC, tuned_parameters, cv=10, scoring='roc_auc')
clf.fit(x_train_avg_w2v, y_train)
Out[156]:
GridSearchCV(cv=10, error_score='raise-deprecating',
      estimator=DecisionTreeClassifier(class_weight='balanced', criterion
='gini',
            max depth=None, max features=None, max leaf nodes=None,
            min_impurity_decrease=0.0, min_impurity_split=None,
            min_samples_leaf=1, min_samples_split=2,
            min_weight_fraction_leaf=0.0, presort=False, random_state=Non
e,
            splitter='best'),
       fit_params=None, iid='warn', n_jobs=None,
       param_grid={'max_depth': [1, 5, 10, 50, 100], 'min_samples_split':
[5, 10, 50, 100]},
       pre_dispatch='2*n_jobs', refit=True, return_train_score='warn',
       scoring='roc_auc', verbose=0)
```

In [157]:

```
results = pd.DataFrame.from dict(clf.cv results )
max_depth_results = results.sort_values(['param_max_depth'])
min_samples_split_results = results.sort_values(['param_min_samples_split'])
#For max depth
train_auc_max_depth= results['mean_train_score']
train_auc_std_max_depth= results['std_train_score']
cv_auc_max_depth = results['mean_test_score']
cv_auc_std_max_depth= results['std_test_score']
max depth = results['param max depth']
#For min samples split
train_auc_min_samples = results['mean_train_score']
train_auc_std_min_samples = results['std_train_score']
cv_auc_min_samples = results['mean_test_score']
cv auc std min samples = results['std test score']
min_samples_split = results['param_min_samples_split']
plt.plot(max_depth, train_auc_max_depth, label='Train AUC max_depth')
# this code is copied from here: https://stackoverflow.com/a/48803361/4084039
# plt.gca().fill between(K, train auc - train auc std,train auc + train auc std,alpha=
0.2, color='darkblue')
plt.plot(min samples split, train auc min samples, label='Train AUC min samples') #Plot
ting _min_samples-split curve
plt.plot(max depth, cv auc max depth, label='CV AUC')
# this code is copied from here: https://stackoverflow.com/a/48803361/4084039
# plt.gca().fill_between(K, cv_auc - cv_auc_std,cv_auc + cv_auc_std,alpha=0.2,color='da
rkorange')
plt.plot(min_samples_split, cv_auc_min_samples, label='CV AUC min_samples')
plt.scatter(max depth, train auc max depth, label='Train AUC points')
plt.scatter(max_depth, cv_auc_max_depth, label='CV AUC points')
plt.scatter(min samples split, train auc min samples, label=('Train AUC min samples poi
nts'))
plt.scatter(min_samples_split, cv_auc_min_samples, label=('CV AUC min_samples'))
plt.legend(loc='center left', bbox_to_anchor=(1, 0.5))
plt.xlabel("max depth, min samples split: hyperparameter")
plt.ylabel("AUC")
plt.title("Hyper parameter Vs AUC plot")
plt.grid()
plt.show()
results.head()
```



Out[157]:

	mean_fit_time	std_fit_time	mean_score_time	std_score_time	param_max_depth	param_n
0	5.150788	0.339405	0.112220	0.021373	1	
1	4.790684	0.067502	0.103318	0.006303	1	
2	4.782198	0.059545	0.101782	0.005550	1	
3	4.778852	0.080818	0.101776	0.008331	1	
4	22.934586	0.337322	0.107333	0.007506	5	

5 rows × 32 columns

In [158]:

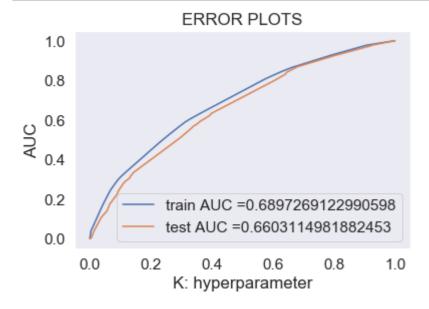
clf.best_params_

Out[158]:

{'max_depth': 5, 'min_samples_split': 5}

In [159]:

```
from sklearn.metrics import roc_curve, auc
DTC_set3 = DecisionTreeClassifier(max_depth=5, min_samples_split=5, class_weight='balan
ced')
DTC_set3.fit(x_train_avg_w2v, y_train)
y_train_pred = clf.predict_proba(x_train_avg_w2v)[:, 1]
y_test_pred = clf.predict_proba(x_test_avg_w2v)[:, 1]
train fpr, train tpr, tr thresholds = roc curve(y train, y train pred)
test_fpr, test_tpr, te_thresholds = roc_curve(y_test, y_test_pred)
plt.grid()
plt.plot(train_fpr, train_tpr, label="train AUC ="+str(auc(train_fpr, train_tpr)))
plt.plot(test_fpr, test_tpr, label="test AUC ="+str(auc(test_fpr, test_tpr)))
plt.legend()
plt.xlabel("K: hyperparameter")
plt.ylabel("AUC")
plt.title("ERROR PLOTS")
plt.show()
```



In [160]:

```
# we are writing our own function for predict, with defined thresould
# we will pick a threshold that will give the least fpr
def find_best_threshold(threshould, fpr, tpr):
    t = threshould[np.argmax(tpr*(1-fpr))]
    # (tpr*(1-fpr)) will be maximum if your fpr is very low and tpr is very high
    print("the maximum value of tpr*(1-fpr)", max(tpr*(1-fpr)), "for threshold", np.rou
nd(t,3))
    return t
def predict with best t(proba, threshould):
    predictions = []
    for i in proba:
        if i>=threshould:
            predictions.append(1)
        else:
            predictions.append(0)
    return predictions
```

In [161]:

[5677 8277]]

```
#Confusion matrix

from sklearn.metrics import confusion_matrix
best_t = find_best_threshold(tr_thresholds, train_fpr, train_tpr)
print("Train confusion matrix")
print(confusion_matrix(y_train, predict_with_best_t(y_train_pred, best_t)))
print("Test confusion matrix")
print(confusion_matrix(y_test, predict_with_best_t(y_test_pred, best_t)))

the maximum value of tpr*(1-fpr) 0.4061738217657156 for threshold 0.49
Train confusion matrix
[[ 3484    1684]
    [11262    17070]]
Test confusion matrix
[[ 1625    921]
```

In [162]:

```
import seaborn as sns

#Train Confusion matrix

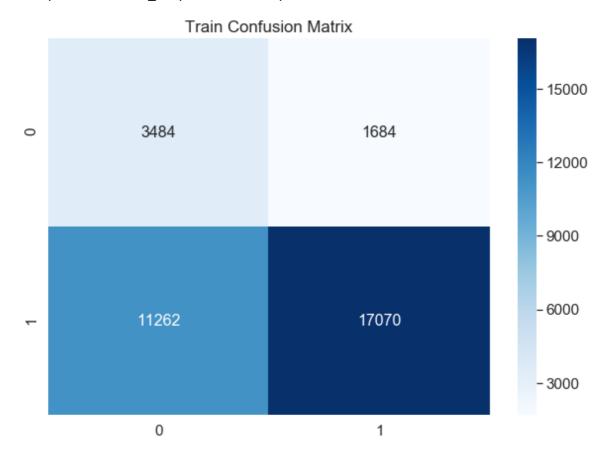
#Reference- https://www.kaggle.com/agungor2/various-confusion-matrix-plots

df_cm = pd.DataFrame(confusion_matrix(y_train, predict_with_best_t(y_train_pred, best_t)), columns=np.unique(y_train), index = np.unique(y_train))

plt.figure(figsize = (10,7))
sns.set(font_scale=1.4)#for label size
plt.title('Train Confusion Matrix')
sns.heatmap(df_cm, cmap="Blues", annot=True, fmt='g', annot_kws={"size": 16})
```

Out[162]:

<matplotlib.axes._subplots.AxesSubplot at 0x1b5242174e0>



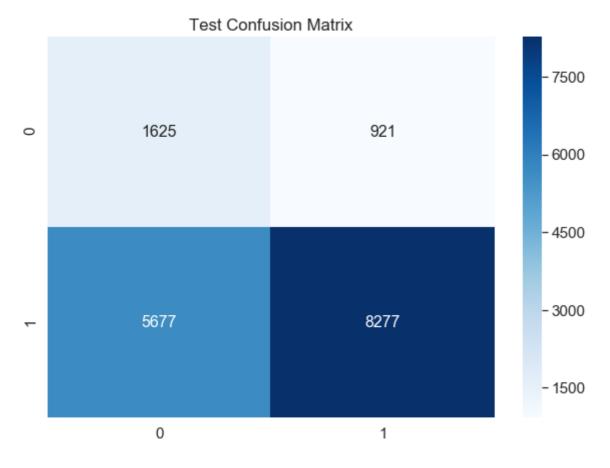
In [163]:

```
#Test Confusion matrix
df_cm = pd.DataFrame(confusion_matrix(y_test, predict_with_best_t(y_test_pred, best_t
)), columns=np.unique(y_test), index = np.unique(y_test))

plt.figure(figsize = (10,7))
sns.set(font_scale=1.4)#for label size
plt.title('Test Confusion Matrix')
sns.heatmap(df_cm, cmap="Blues", annot=True, fmt='g', annot_kws={"size": 16})
```

Out[163]:

<matplotlib.axes._subplots.AxesSubplot at 0x1b52388d7b8>



In [164]:

```
#Getting all the false positive points
```

In [165]:

```
test_fpr = []
for i in range(len(y_test)):
    if y_test[i]==0 and y_test_predicted[i]==1:
        test_fpr.append(i)
```

In [166]:

```
#Plotting the word cloud
```

In [167]:

```
values = x_test['essay'].values
essay_fpr = []
for i in range(len(values)):
    if y_test[i] == 0 and y_test_predicted[i] == 1:
        essay_fpr.append(values[i])

print(len(essay_fpr))
```

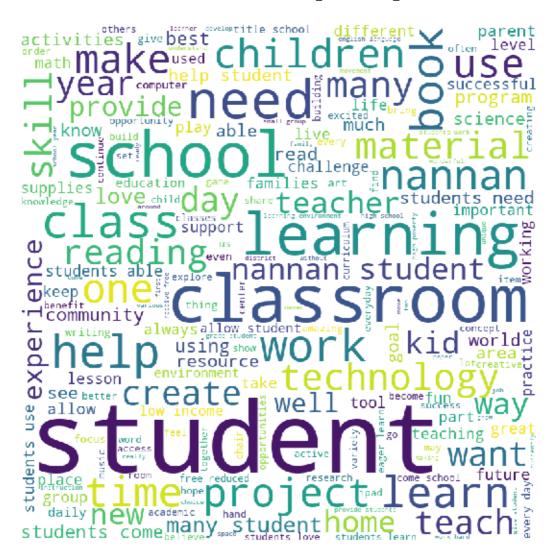
891

In [168]:

```
word=[]
for i in essay_fpr:
    word.append(i.split())
```

In [169]:

```
import nltk
from wordcloud import WordCloud, STOPWORDS
comment words = ' '
stopwords = set(STOPWORDS)
for val in essay_fpr:
    # typecaste each val to string
   val = str(val)
    # split the value
   tokens = val.split()
    # Converts each token into Lowercase
    for i in range(len(tokens)):
        tokens[i] = tokens[i].lower()
    for words in tokens:
        comment_words = comment_words + words + ' '
wordcloud = WordCloud(width = 800, height = 800,
                background_color ='white',
                stopwords = stopwords,
                min_font_size = 10).generate(comment_words)
# plot the WordCloud image
plt.figure(figsize = (8, 8), facecolor = None)
plt.imshow(wordcloud)
plt.axis("off")
plt.tight_layout(pad = 0)
plt.show()
```



In [170]:

```
#Plotting the boxplot with price

values = x_test['price'].values
price_fpr = []
for i in range(len(values)):
    if y_test[i] == 0 and y_test_predicted[i] == 1:
        price_fpr.append(values[i])

print(len(price_fpr))
```

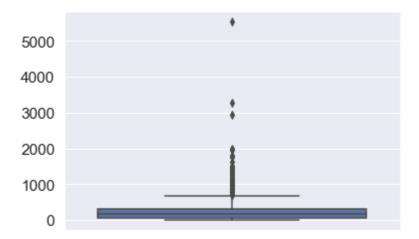
891

In [171]:

```
sns.boxplot(y=price_fpr)
```

Out[171]:

<matplotlib.axes._subplots.AxesSubplot at 0x1b5891c0e80>



In [172]:

#plotting the pdf with teacher_number_of_previously_posted_projects

In [173]:

```
values = x_test['teacher_number_of_previously_posted_projects'].values
teacher_previous_projects_fpr = []
for i in range(len(values)):
    if y_test[i]==0 and y_test_predicted[i]==1:
        teacher_previous_projects_fpr.append(values[i])

print(len(teacher_previous_projects_fpr))
```

891

In [174]:

```
#Reference https://stackoverflow.com/questions/15415455/plotting-probability-density-fu
nction-by-sample-with-matplotLib
from scipy.stats.kde import gaussian_kde
from numpy import linspace

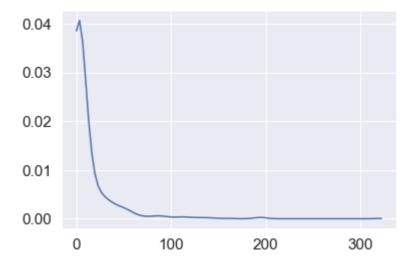
kde = gaussian_kde(teacher_previous_projects_fpr)

dist_space = linspace( min(teacher_previous_projects_fpr), max(teacher_previous_project
s_fpr), 100 )

plt.plot( dist_space, kde(dist_space) )
```

Out[174]:

[<matplotlib.lines.Line2D at 0x1b521f809b0>]



In []:

In []:

2.4.4 Applying Decision Trees on TFIDF W2V, SET 4

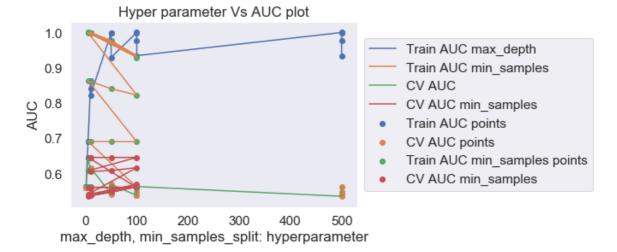
In [175]:

Please write all the code with proper documentation

In [176]:

In [177]:

```
from sklearn.model selection import GridSearchCV
from sklearn.tree import DecisionTreeClassifier
tuned parameters = {'max depth':[1, 5, 10, 50, 100, 500], 'min samples split':[5, 10, 5
0, 100]}
DTC = DecisionTreeClassifier(class weight='balanced')
clf = GridSearchCV(DTC, tuned_parameters, cv=10, scoring='roc_auc')
clf.fit(x_train_avg_w2v, y_train)
results = pd.DataFrame.from dict(clf.cv results )
max_depth_results = results.sort_values(['param_max_depth'])
min_samples_split_results = results.sort_values(['param_min_samples_split'])
#For max depth
train_auc_max_depth= results['mean_train_score']
train_auc_std_max_depth= results['std_train_score']
cv_auc_max_depth = results['mean_test_score']
cv_auc_std_max_depth= results['std_test_score']
max_depth = results['param_max_depth']
#For min_samples_split
train_auc_min_samples = results['mean_train_score']
train auc std min samples = results['std train score']
cv_auc_min_samples = results['mean_test_score']
cv_auc_std_min_samples = results['std_test_score']
min_samples_split = results['param_min_samples_split']
plt.plot(max depth, train auc max depth, label='Train AUC max depth')
# this code is copied from here: https://stackoverflow.com/a/48803361/4084039
# plt.gca().fill_between(K, train_auc - train_auc_std,train_auc + train_auc_std,alpha=
0.2, color='darkblue')
plt.plot(min samples split, train auc min samples, label='Train AUC min samples') #Plot
ting min samples-split curve
plt.plot(max_depth, cv_auc_max_depth, label='CV AUC')
# this code is copied from here: https://stackoverflow.com/a/48803361/4084039
# plt.gca().fill_between(K, cv_auc - cv_auc_std,cv_auc + cv_auc_std,alpha=0.2,color='da
rkorange')
plt.plot(min samples split, cv auc min samples, label='CV AUC min samples')
plt.scatter(max depth, train auc max depth, label='Train AUC points')
plt.scatter(max_depth, cv_auc_max_depth, label='CV AUC points')
plt.scatter(min samples split, train auc min samples, label=('Train AUC min samples poi
nts'))
plt.scatter(min samples split, cv auc min samples, label=('CV AUC min samples'))
plt.legend(loc='center left', bbox_to_anchor=(1, 0.5))
plt.xlabel("max depth, min samples split: hyperparameter")
plt.ylabel("AUC")
plt.title("Hyper parameter Vs AUC plot")
plt.grid()
plt.show()
results.head()
```



Out[177]:

	mean_fit_time	std_fit_time	mean_score_time	std_score_time	param_max_depth	param_n
0	5.004209	0.295018	0.107509	0.006885	1	
1	4.684733	0.119384	0.103218	0.009065	1	
2	5.015690	0.369614	0.112798	0.019882	1	
3	4.724663	0.039653	0.103178	0.004986	1	
4	23.571398	1.369166	0.104841	0.009355	5	

5 rows × 32 columns

In [178]:

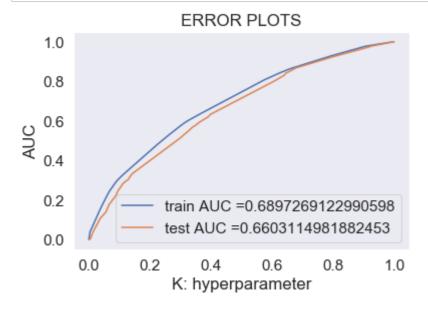
clf.best_params_

Out[178]:

{'max_depth': 5, 'min_samples_split': 5}

In [179]:

```
from sklearn.metrics import roc_curve, auc
DTC_set4 = DecisionTreeClassifier(max_depth=5, min_samples_split=5, class_weight='balan
ced')
DTC_set4.fit(x_train_avg_w2v, y_train)
y_train_pred = clf.predict_proba(x_train_avg_w2v)[:, 1]
y_test_pred = clf.predict_proba(x_test_avg_w2v)[:, 1]
train fpr, train tpr, tr thresholds = roc curve(y train, y train pred)
test_fpr, test_tpr, te_thresholds = roc_curve(y_test, y_test_pred)
plt.grid()
plt.plot(train_fpr, train_tpr, label="train AUC ="+str(auc(train_fpr, train_tpr)))
plt.plot(test_fpr, test_tpr, label="test AUC ="+str(auc(test_fpr, test_tpr)))
plt.legend()
plt.xlabel("K: hyperparameter")
plt.ylabel("AUC")
plt.title("ERROR PLOTS")
plt.show()
```



In [181]:

```
# we are writing our own function for predict, with defined thresould
# we will pick a threshold that will give the least fpr
def find_best_threshold(threshould, fpr, tpr):
    t = threshould[np.argmax(tpr*(1-fpr))]
    # (tpr*(1-fpr)) will be maximum if your fpr is very low and tpr is very high
    print("the maximum value of tpr*(1-fpr)", max(tpr*(1-fpr)), "for threshold", np.rou
nd(t,3))
    return t
def predict with best t(proba, threshould):
    predictions = []
    for i in proba:
        if i>=threshould:
            predictions.append(1)
        else:
            predictions.append(0)
    return predictions
```

In [182]:

[[1625 921] [5677 8277]]

```
#Confusion matrix

from sklearn.metrics import confusion_matrix
best_t = find_best_threshold(tr_thresholds, train_fpr, train_tpr)
print("Train confusion matrix")
print(confusion_matrix(y_train, predict_with_best_t(y_train_pred, best_t)))
print("Test confusion matrix")
print(confusion_matrix(y_test, predict_with_best_t(y_test_pred, best_t)))

the maximum value of tpr*(1-fpr) 0.4061738217657156 for threshold 0.49
Train confusion matrix
[[ 3484    1684]
    [11262    17070]]
Test confusion matrix
```

In [183]:

```
import seaborn as sns

#Train Confusion matrix

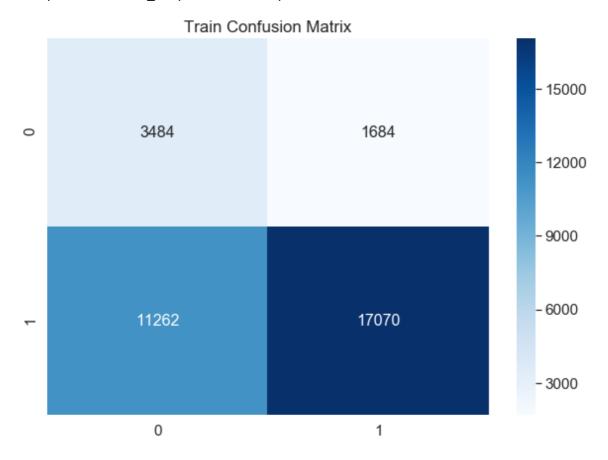
#Reference- https://www.kaggle.com/agungor2/various-confusion-matrix-plots

df_cm = pd.DataFrame(confusion_matrix(y_train, predict_with_best_t(y_train_pred, best_t)), columns=np.unique(y_train), index = np.unique(y_train))

plt.figure(figsize = (10,7))
sns.set(font_scale=1.4)#for label size
plt.title('Train Confusion Matrix')
sns.heatmap(df_cm, cmap="Blues", annot=True, fmt='g', annot_kws={"size": 16})
```

Out[183]:

<matplotlib.axes._subplots.AxesSubplot at 0x1b524c5f898>



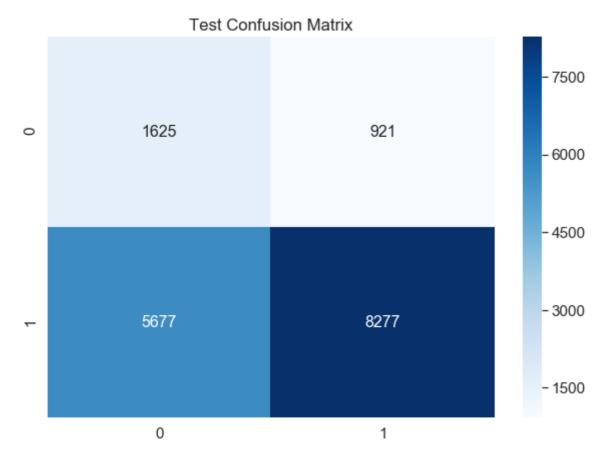
In [184]:

```
#Test Confusion matrix
df_cm = pd.DataFrame(confusion_matrix(y_test, predict_with_best_t(y_test_pred, best_t
)), columns=np.unique(y_test), index = np.unique(y_test))

plt.figure(figsize = (10,7))
sns.set(font_scale=1.4)#for Label size
plt.title('Test Confusion Matrix')
sns.heatmap(df_cm, cmap="Blues", annot=True, fmt='g', annot_kws={"size": 16})
```

Out[184]:

<matplotlib.axes._subplots.AxesSubplot at 0x1b57a208a58>



In [185]:

```
#Getting all the false positive points
```

In [186]:

```
test_fpr = []
for i in range(len(y_test)):
    if y_test[i]==0 and y_test_predicted[i]==1:
        test_fpr.append(i)
```

In [187]:

```
#Plotting the word cloud
```

In [188]:

```
values = x_test['essay'].values
essay_fpr = []
for i in range(len(values)):
    if y_test[i] == 0 and y_test_predicted[i] == 1:
        essay_fpr.append(values[i])

print(len(essay_fpr))
```

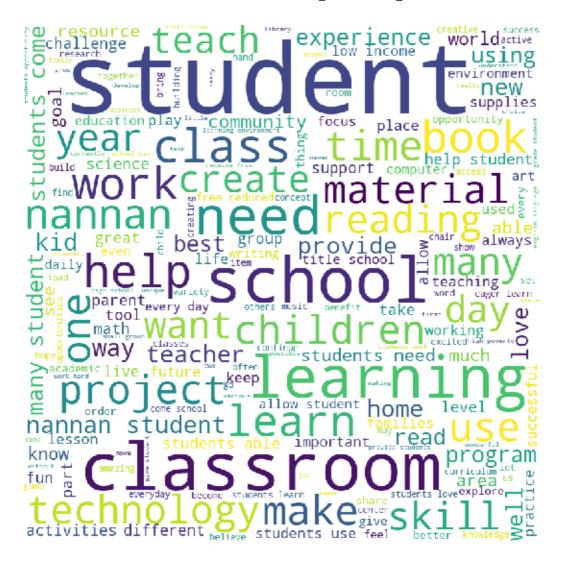
891

In [189]:

```
word=[]
for i in essay_fpr:
    word.append(i.split())
```

In [190]:

```
import nltk
from wordcloud import WordCloud, STOPWORDS
comment words = ' '
stopwords = set(STOPWORDS)
for val in essay_fpr:
    # typecaste each val to string
   val = str(val)
    # split the value
   tokens = val.split()
    # Converts each token into Lowercase
    for i in range(len(tokens)):
        tokens[i] = tokens[i].lower()
    for words in tokens:
        comment_words = comment_words + words + ' '
wordcloud = WordCloud(width = 800, height = 800,
                background_color ='white',
                stopwords = stopwords,
                min_font_size = 10).generate(comment_words)
# plot the WordCloud image
plt.figure(figsize = (8, 8), facecolor = None)
plt.imshow(wordcloud)
plt.axis("off")
plt.tight_layout(pad = 0)
plt.show()
```



In [191]:

```
#Plotting the boxplot with price

values = x_test['price'].values
price_fpr = []
for i in range(len(values)):
    if y_test[i]==0 and y_test_predicted[i]==1:
        price_fpr.append(values[i])

print(len(price_fpr))
```

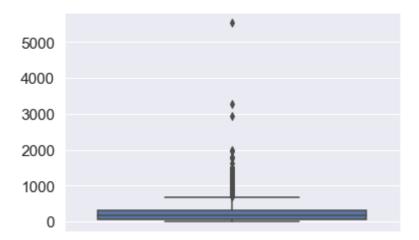
891

In [192]:

```
sns.boxplot(y=price_fpr)
```

Out[192]:

<matplotlib.axes._subplots.AxesSubplot at 0x1b526a24080>



In [193]:

#plotting the pdf with teacher_number_of_previously_posted_projects

In [194]:

```
values = x_test['teacher_number_of_previously_posted_projects'].values
teacher_previous_projects_fpr = []
for i in range(len(values)):
    if y_test[i]==0 and y_test_predicted[i]==1:
        teacher_previous_projects_fpr.append(values[i])
print(len(teacher_previous_projects_fpr))
```

891

In [195]:

```
#Reference https://stackoverflow.com/questions/15415455/plotting-probability-density-fu
nction-by-sample-with-matplotlib
from scipy.stats.kde import gaussian_kde
from numpy import linspace

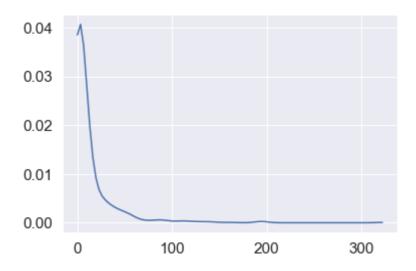
kde = gaussian_kde(teacher_previous_projects_fpr)

dist_space = linspace( min(teacher_previous_projects_fpr), max(teacher_previous_projects_fpr), 100 )

plt.plot( dist_space, kde(dist_space) )
```

Out[195]:

[<matplotlib.lines.Line2D at 0x1b5250b4c50>]



2.5 [Task-2]Getting top 5k features using `feature_importances_`

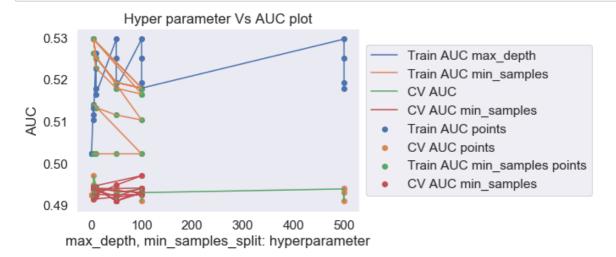
In [196]:

```
# please write all the code with proper documentation, and proper titles for each subse
ction
# 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 debugging your
code
# when you plot any graph make sure you use
# a. Title, that describes your plot, this will be very helpful to the reader
# b. Legends if needed
# c. X-axis label
# d. Y-axis label
```

```
In [197]:
x_train_set5 = DTC_set2.feature_importances_
x_train_set5 = x_train_set5[:5000]
In [198]:
x_train_set5 = x_train_set5.reshape(-1, 1)
x_train_set5.shape
Out[198]:
(5000, 1)
In [209]:
x_test_set5 = DTC_set2.feature_importances_
x_{test_set5} = x_{test_set5}[:5000]
x_test_set5 = x_test_set5.reshape(-1, 1)
x_test_set5.shape
Out[209]:
(5000, 1)
In [199]:
y_train_set5 = y_train[:5000]
y_train_set5.shape
Out[199]:
(5000,)
In [212]:
y_{\text{test\_set5}} = y_{\text{test}}[:5000]
y_test_set5.shape
Out[212]:
(5000,)
In [ ]:
```

In [229]:

```
from sklearn.model selection import GridSearchCV
from sklearn.tree import DecisionTreeClassifier
tuned parameters = {'max depth':[1, 5, 10, 50, 100, 500], 'min samples split':[5, 10, 5
0, 100]}
DTC = DecisionTreeClassifier(class_weight='balanced')
clf = GridSearchCV(DTC, tuned_parameters, cv=10, scoring='roc_auc')
clf.fit(x_train_set5, y_train_set5)
results = pd.DataFrame.from dict(clf.cv results )
max_depth_results = results.sort_values(['param_max_depth'])
min_samples_split_results = results.sort_values(['param_min_samples_split'])
#For max depth
train_auc_max_depth= results['mean_train_score']
train_auc_std_max_depth= results['std_train_score']
cv_auc_max_depth = results['mean_test_score']
cv_auc_std_max_depth= results['std_test_score']
max_depth = results['param_max_depth']
#For min_samples_split
train_auc_min_samples = results['mean_train_score']
train auc std min samples = results['std train score']
cv_auc_min_samples = results['mean_test_score']
cv_auc_std_min_samples = results['std_test_score']
min_samples_split = results['param_min_samples_split']
plt.plot(max_depth, train_auc_max_depth, label='Train AUC max_depth')
# this code is copied from here: https://stackoverflow.com/a/48803361/4084039
# plt.gca().fill_between(K, train_auc - train_auc_std,train_auc + train_auc_std,alpha=
0.2, color='darkblue')
plt.plot(min samples split, train auc min samples, label='Train AUC min samples') #Plot
ting min samples-split curve
plt.plot(max_depth, cv_auc_max_depth, label='CV AUC')
# this code is copied from here: https://stackoverflow.com/a/48803361/4084039
# plt.gca().fill_between(K, cv_auc - cv_auc_std,cv_auc + cv_auc_std,alpha=0.2,color='da
rkorange')
plt.plot(min samples split, cv auc min samples, label='CV AUC min samples')
plt.scatter(max depth, train auc max depth, label='Train AUC points')
plt.scatter(max_depth, cv_auc_max_depth, label='CV AUC points')
plt.scatter(min samples split, train auc min samples, label=('Train AUC min samples poi
nts'))
plt.scatter(min samples split, cv auc min samples, label=('CV AUC min samples'))
plt.legend(loc='center left', bbox_to_anchor=(1, 0.5))
plt.xlabel("max depth, min samples split: hyperparameter")
plt.ylabel("AUC")
plt.title("Hyper parameter Vs AUC plot")
plt.grid()
plt.show()
results.head()
```



Out[229]:

	mean_fit_time	std_fit_time	mean_score_time	std_score_time	param_max_depth	param_n
0	0.006802	0.009814	0.001601	0.001744	1	
1	0.005054	0.001665	0.002901	0.003477	1	
2	0.003520	0.005680	0.002163	0.004663	1	
3	0.000201	0.000602	0.001563	0.004688	1	
4	0.003327	0.006668	0.000150	0.000451	5	

5 rows × 32 columns

In [230]:

clf.best_params_

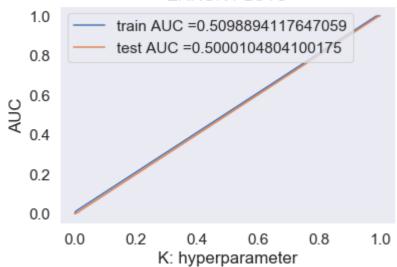
Out[230]:

{'max_depth': 5, 'min_samples_split': 100}

In [231]:

```
from sklearn.metrics import roc_curve, auc
DTC_set5 = DecisionTreeClassifier(max_depth=5, min_samples_split=100, class_weight='bal
anced')
DTC_set5.fit(x_train_set5, y_train_set5)
y_train_pred = clf.predict_proba(x_train_set5)[:, 1]
y_test_pred = clf.predict_proba(x_test_set5)[:, 1]
train fpr, train tpr, tr thresholds = roc curve(y train set5, y train pred)
test_fpr, test_tpr, te_thresholds = roc_curve(y_test_set5, y_test_pred)
plt.grid()
plt.plot(train_fpr, train_tpr, label="train AUC ="+str(auc(train_fpr, train_tpr)))
plt.plot(test_fpr, test_tpr, label="test AUC ="+str(auc(test_fpr, test_tpr)))
plt.legend()
plt.xlabel("K: hyperparameter")
plt.ylabel("AUC")
plt.title("ERROR PLOTS")
plt.show()
```

ERROR PLOTS



In [232]:

[4155

51]]

```
#Confusion matrix
from sklearn.metrics import confusion_matrix
best_t = find_best_threshold(tr_thresholds, train_fpr, train_tpr)
print("Train confusion matrix")
print(confusion_matrix(y_train_set5, predict_with_best_t(y_train_pred, best_t)))
print("Test confusion matrix")
print(confusion_matrix(y_test_set5, predict_with_best_t(y_test_pred, best_t)))
the maximum value of tpr*(1-fpr) 0.013826823529411765 for threshold 0.688
Train confusion matrix
[[ 747
          3]
 [4191
         59]]
Test confusion matrix
[[ 783
         11]
```

In [234]:

```
import seaborn as sns

#Train Confusion matrix

#Reference- https://www.kaggle.com/agungor2/various-confusion-matrix-plots

df_cm = pd.DataFrame(confusion_matrix(y_train_set5, predict_with_best_t(y_train_pred, b est_t)), columns=np.unique(y_train_set5), index = np.unique(y_train_set5))

plt.figure(figsize = (10,7))
sns.set(font_scale=1.4)#for label size
plt.title('Train Confusion Matrix')
sns.heatmap(df_cm, cmap="Blues", annot=True, fmt='g', annot_kws={"size": 16})
```

Out[234]:

<matplotlib.axes._subplots.AxesSubplot at 0x1b5264a9588>



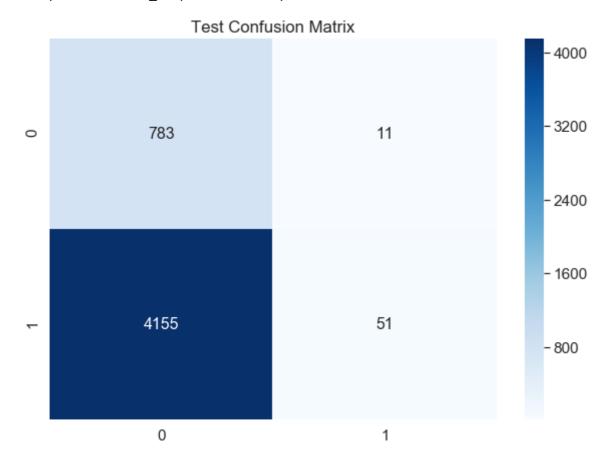
In [236]:

```
#Test Confusion matrix
df_cm = pd.DataFrame(confusion_matrix(y_test_set5, predict_with_best_t(y_test_pred, bes
t_t)), columns=np.unique(y_test_set5), index = np.unique(y_test_set5))

plt.figure(figsize = (10,7))
sns.set(font_scale=1.4)#for Label size
plt.title('Test Confusion Matrix')
sns.heatmap(df_cm, cmap="Blues", annot=True, fmt='g', annot_kws={"size": 16})
```

Out[236]:

<matplotlib.axes._subplots.AxesSubplot at 0x1b525cf87b8>



In []:

In []:

3. Conclusion

In [0]:

Please compare all your models using Prettytable library

In [237]:

```
from prettytable import PrettyTable

x = PrettyTable()

x. field_names = ['Vectorizer', 'Model', 'max_depth', 'min_samples_split', 'Test AUC']

x.add_row(['BOW', 'Decision Tree', '10', '100', '0.65'])

x.add_row(['TFIDF', 'Decision Tree', '10', '5', '0.65'])

x.add_row(['AVG W2V', 'Decision Tree', '5', '5', '0.66'])

x.add_row(['TFIDF W2V', 'Decision Tree', '5', '5', '0.66'])

x.add_row(['TFIDF', 'Decision Tree', '5', '100', '0.5'])

print(x)
```

Vectorizer	+ Model +	+ max_depth +	+ min_samples_split +	++ Test AUC ++
BOW TFIDF AVG W2V TFIDF W2V	Decision Tree Decision Tree Decision Tree Decision Tree	10 10 5 5	100 5 5	0.65 0.65 0.66 0.66
TFIDF	Decision Tree	5	100	0.5

In []: