DonorsChoose

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

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

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

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

About the DonorsChoose Data Set

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

Feature	Desc
`project_id`	A unique identifier for the proposed project. **Example:** `p0`
`project_title`	Title of the project. **Exa** • `Art Will Make You H` • `First Grad`
`project_grade_category`	Grade level of students for which the project is targeted. One of the fo enumerated v • `Grades P` • `Grade` • `Grade` • `Grades`
`project_subject_categories`	One or more (comma-separated) subject categories for the project fr following enumerated list of v • `Applied Lea` • `Care & H` • `Health & S` • `History & C` • `Literacy & Lan` • `Math & Sc` • `Music & The` • `Special` • `W` **Exar** • `Music & The` • `Literacy & Language, Math & Sc`

Feature	Desc
`school_state`	State where school is located (Two-letter U.S. posta (https://en.wikipedia.org/wiki/List_of_U.S._state_abbreviations#Postal_c **Exampl**
`project_subject_subcategories`	One or more (comma-separated) subject subcategories for the p **Exan** • Lit • Literature & Writing, Social Sci
`project_resource_summary`	An explanation of the resources needed for the project. **Exa** • My students need hands on literacy materials to ma sensory needs!<
`project_essay_1`	First application
`project_essay_2`	Second application
`project_essay_3`	Third application
`project_essay_4`	Fourth application
`project_submitted_datetime`	Datetime when project application was submitted. **Example:** 2016-0 12:43:5
`teacher_id`	A unique identifier for the teacher of the proposed project. **Exa** bdf8baa8fedef6bfeec7ae4ff1c
`teacher_prefix`	Teacher's title. One of the following enumerated v • • • • • • Tea
`teacher_number_of_previously_posted_projects`	Number of project applications previously submitted by the same te **Exam**

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

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

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

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.

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: d
etected 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]:

```
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' 's
chool_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
1	p069063	Bouncy Bands for Desks (Blue support pipes)	3	14.95

1.2 preprocessing of `project_subject_categories`

In [5]:

```
catogories = list(project_data['project_subject_categories'].values)
# remove special characters from list of strings python: https://stackoverflow.com/a/473019

# https://www.geeksforgeeks.org/removing-stop-words-nltk-python/
# https://stackoverflow.com/questions/23669024/how-to-strip-a-specific-word-from-a-string
# https://stackoverflow.com/questions/8270092/remove-all-whitespace-in-a-string-in-python
cat_list = []
for i in catogories:
    temp = ""
    # consider we have text like this "Math & Science, Warmth, Care & Hunger"
    for j in i.split(','): # it will split it in three parts ["Math & Science", "Warmth", "
        if 'The' in j.split(): # this will split each of the catogory based on space "Math
            j=j.replace('The','') # if we have the words "The" we are going to replace it w
        j = j.replace(' ','') # we are placeing all the ' '(space) with ''(empty) ex:"Math
        temp+=j.strip()+" " #" abc ".strip() will return "abc", remove the trailing spaces
        temp = temp.replace('&','_') # we are replacing the & value into
    cat_list.append(temp.strip())

project_data['clean_categories'] = cat_list
project_data.drop(['project_subject_categories'], axis=1, inplace=True)

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

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

1.3 preprocessing of `project_subject_subcategories`

In [6]:

```
sub_catogories = list(project_data['project_subject_subcategories'].values)
# remove special characters from list of strings python: https://stackoverflow.com/a/473019

# https://www.geeksforgeeks.org/removing-stop-words-nltk-python/
# https://stackoverflow.com/questions/23669024/how-to-strip-a-specific-word-from-a-string
# https://stackoverflow.com/questions/8270092/remove-all-whitespace-in-a-string-in-python

sub_cat_list = []
for i in sub_catogories:
    temp = ""
    # consider we have text like this "Math & Science, Warmth, Care & Hunger"
    for j in i.split(','): # it will split it in three parts ["Math & Science", "Warmth", "
        if 'The' in j.split(): # this will split each of the catogory based on space "Math
            j=j.replace('The','') # if we have the words "The" we are going to replace it w
        j = j.replace(' ','') # we are placeing all the ' '(space) with ''(empty) ex:"Math
        temp +=j.strip()+" "#" abc ".strip() will return "abc", remove the trailing spaces
        temp = temp.replace('&','_')
    sub_cat_list.append(temp.strip())

project_data['clean_subcategories'] = sub_cat_list
project_data.drop(['project_subject_subcategories'], axis=1, inplace=True)

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

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

1.3 Text preprocessing

In [7]:

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

In [8]:

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

Out[8]:

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

In [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 second
or third languages. We are a melting pot of refugees, immigrants, and native
-born Americans bringing the gift of language to our school. \r\n\r\n We hav
e over 24 languages represented in our English Learner program with students
at every level of mastery.  We also have over 40 countries represented with
the families within our school.  Each student brings a wealth of knowledge a
nd experiences to us that open our eyes to new cultures, beliefs, and respec
t.\"The limits of your language are the limits of your world.\"-Ludwig Wittg
enstein  Our English learner's have a strong support system at home that beg
s for more resources.  Many times our parents are learning to read and speak
English along side of their children.  Sometimes this creates barriers for p
arents to be able to help their child learn phonetics, letter recognition, a
nd other reading skills.\r\n\r\nBy providing these dvd's and players, studen
ts are able to continue their mastery of the English language even if no one
at home is able to assist.  All families with students within the Level 1 pr
oficiency status, will be a offered to be a part of this program.  These edu
cational 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 deve
lop early reading skills.\r\n\r\nParents that do not have access to a dvd pl
ayer will have the opportunity to check out a dvd player to use for the yea
r.  The plan is to use these videos and educational dvd's for the years to c
ome for other EL students.\r\nnannan
==================================================
The 51 fifth grade students that will cycle through my classroom this year a
ll love learning, at least most of the time. At our school, 97.3% of the stu
dents receive free or reduced price lunch. Of the 560 students, 97.3% are mi
nority students. \r\nThe school has a vibrant community that loves to get to
gether and celebrate. Around Halloween there is a whole school parade to sho
w 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 games. At the e
nd of the year the school hosts a carnival to celebrate the hard work put in
during the school year, with a dunk tank being the most popular activity.My
students will use these five brightly colored Hokki stools in place of regul
ar, stationary, 4-legged chairs. As I will only have a total of ten in the c
lassroom and not enough for each student to have an individual one, they wil
l 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 th
em in place of chairs at my small group tables during math and reading time
s. The rest of the day they will be used by the students who need the highes
t amount of movement in their life in order to stay focused on school.\r\n\r
\nWhenever asked what the classroom is missing, my students always say more
Hokki Stools. They can't get their fill of the 5 stools we already have. Whe
n the students are sitting in group with me on the Hokki Stools, they are al
ways moving, but at the same time doing their work. Anytime the students get
to pick where they can sit, the Hokki Stools are the first to be taken. Ther
e are always students who head over to the kidney table to get one of the st
ools 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 comprom
```

ise that allow my students to do desk work and move at the same time. These stools will help students to meet their 60 minutes a day of movement by allo wing them to activate their core muscles for balance while they sit. For man y of my students, these chairs will take away the barrier that exists in sch ools for a child who can't sit still.nannan

==========

How do you remember your days of school? Was it in a sterile environment wit h plain walls, rows of desks, and a teacher in front of the room? A typical day in our room is nothing like that. I work hard to create a warm inviting themed room for my students look forward to coming to each day.\r\n\r\nMy cl ass is made up of 28 wonderfully unique boys and girls of mixed races in Ark ansas.\r\nThey attend a Title I school, which means there is a high enough p ercentage of free and reduced-price lunch to qualify. Our school is an \"ope n 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; th ey are like sponges, absorbing all the information and experiences and keep on wanting more.With these resources such as the comfy red throw pillows and the whimsical nautical hanging decor and the blue fish nets, I will be able to help create the mood in our classroom setting to be one of a themed nauti cal environment. Creating a classroom environment is very important in the s uccess in each and every child's education. The 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 pictures of each child with them, hav e them developed, and then hung in our classroom ready for their first day o f 4th grade. This kind gesture will set the tone before even the first day of school! The nautical thank you cards will be used throughout the year by the students as they create thank you cards to their team groups.\r\n\r\nYou r generous donations will help me to help make our classroom a fun, invitin g, learning environment from day one.\r\n\r\nIt costs lost of money out of m y own pocket on resources to get our classroom ready. Please consider helpin g 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 la nguage delays, cognitive delays, gross/fine motor delays, to autism. They ar e eager beavers and always strive to work their hardest working past their l imitations. \r\n\r\nThe materials we have are the ones I seek out for my stu dents. I teach in a Title I school where most of the students receive free o r reduced price lunch. Despite their disabilities and limitations, my stude nts 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 yo u were in a meeting? This is how my kids feel all the time. The want to be a ble to move as they learn or so they say.Wobble chairs are the answer and I love then because they develop their core, which enhances gross motor and in Turn fine motor skills. \r\nThey also want to learn through games, my kids d on't want to sit and do worksheets. They want to learn to count by jumping a nd playing. Physical engagement is the key to our success. The number toss a nd color and shape mats can make that happen. My students will forget they a re 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"\'ll", " will", phrase)
    phrase = re.sub(r"\'t", " not", phrase)
    phrase = re.sub(r"\'ve", " have", phrase)
    phrase = re.sub(r"\'m", " am", phrase)
    return phrase
```

In [12]:

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

```
My kindergarten students have varied disabilities ranging from speech and la
nguage delays, cognitive delays, gross/fine motor delays, to autism. They ar
e eager beavers and always strive to work their hardest working past their l
imitations. \r\n\r\nThe materials we have are the ones I seek out for my stu
dents. I teach in a Title I school where most of the students receive free o
r reduced price lunch.  Despite their disabilities and limitations, my stude
nts 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 yo
u were in a meeting? This is how my kids feel all the time. The want to be a
ble to move as they learn or so they say.Wobble chairs are the answer and I
love then because they develop their core, which enhances gross motor and in
Turn fine motor skills. \r\nThey also want to learn through games, my kids d
o not want to sit and do worksheets. They want to learn to count by jumping
and playing. Physical engagement is the key to our success. The number toss
and color and shape mats can make that happen. My students will forget they
are doing work and just have the fun a 6 year old deserves.nannan
==================================================
```

In [13]:

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

```
My kindergarten students have varied disabilities ranging from speech and la
nguage delays, cognitive delays, gross/fine motor delays, to autism. They ar
e eager beavers and always strive to work their hardest working past their l
imitations.     The materials we have are the ones I seek out for my student
s. I teach in a Title I school where most of the students receive free or re
duced price lunch.  Despite their disabilities and limitations, 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 groove and move as you wer
e in a meeting? This is how my kids feel all the time. The want to be able t
o move as they learn or so they say.Wobble chairs are the answer and I love
then because they develop their core, which enhances gross motor and in Turn
fine motor skills.   They also want to learn through games, my kids do not w
ant to sit and do worksheets. They want to learn to count by jumping and pla
ying. Physical engagement is the key to our success. The number toss and col
or and shape mats can make that happen. My students will forget they are doi
ng work and just have the fun a 6 year old deserves.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 la
nguage delays cognitive delays gross fine motor delays to autism They are ea
ger beavers and always strive to work their hardest working past their limit
ations The materials we have are the ones I seek out for my students I teach
in a Title I school where most of the students receive free or reduced price
lunch Despite their disabilities and limitations my students love coming to
school and come eager to learn and explore Have you ever felt like you had a
nts 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 le
arn or so they say Wobble chairs are the answer and I love then because they
develop their core which enhances gross motor and in Turn fine motor skills
They also want to learn through games my kids do not want to sit and do work
sheets They want to learn to count by jumping and playing Physical engagemen
t is the key to our success The number toss and color and shape mats can mak
e 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're", "
            "you'll", "you'd", 'your', 'yours', 'yourself', 'yourselves', 'he', 'him', 'his
            'she', "she's", 'her', 'hers', 'herself', 'it', "it's", 'its', 'itself', 'they'
            'theirs', 'themselves', 'what', 'which', 'who', 'whom', 'this', 'that', "that'l
            'am', 'is', 'are', 'was', 'were', 'be', 'been', 'being', 'have', 'has', 'had',
            'did', 'doing', 'a', 'an', 'the', 'and', 'but', 'if', 'or', 'because', 'as', 'u
            'at', 'by', 'for', 'with', 'about', 'against', 'between', 'into', 'through', 'd
            'above', 'below', 'to', 'from', 'up', 'down', 'in', 'out', 'on', 'off', 'over',
            'then', 'once', 'here', 'there', 'when', 'where', 'why', 'how', 'all', 'any',
            'most', 'other', 'some', 'such', 'only', 'own', 'same', 'so', 'than', 'too', 'v
            's', 't', 'can', 'will', 'just', 'don', "don't", 'should', "should've", 'now',
            've', 'y', 'ain', 'aren', "aren't", 'couldn', "couldn't", 'didn', "didn't", 'do
            "hadn't", 'hasn', "hasn't", 'haven', "haven't", 'isn', "isn't", 'ma', 'mightn',
            "mustn't", 'needn', "needn't", 'shan', "shan't", 'shouldn', "shouldn't", 'wasn'
            '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('\\"', ' ')
    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 [00:49<00:00, 1009.94it/s]
```

In [17]:

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

Out[17]:

```
'kindergarten students varied disabilities ranging speech language delays co
gnitive delays gross fine motor delays autism eager beavers always strive wo
rk hardest working past limitations materials ones seek students teach title
school students receive free reduced price lunch despite disabilities limita
tions students love coming school come eager learn explore ever felt like an
ts 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 j
umping 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('\\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_titles.append(sent.lower().strip())
```

```
100%|██████████████████████████████████████████████████████████████████████████████
██| 50000/50000 [00:02<00:00, 21702.23it/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-replace-

project_data['project_grade_category'] = project_data['project_grade_category'].str.replace
project_data['project_grade_category'] = project_data['project_grade_category'].str.replace
```

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_titl
e',
       '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
- price : numerical

1.5.1 Vectorizing Categorical data

- https://www.appliedaicourse.com/course/applied-ai-course-online/lessons/handling-categorical-and-numerical-features/ (https://www.appliedaicourse.com/course/applied-ai-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, bina
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=False,
sub_categories_one_hot = vectorizer.fit_transform(project_data['clean_subcategories'].value
print(vectorizer.get_feature_names())
print("Shape of matrix after one hot encodig ",sub_categories_one_hot.shape)
```

```
['Economics', 'CommunityService', 'FinancialLiteracy', 'ParentInvolvement',
'Extracurricular', 'Civics_Government', 'ForeignLanguages', 'NutritionEducat
ion', 'Warmth', 'Care_Hunger', 'SocialSciences', 'PerformingArts', 'Characte
rEducation', 'TeamSports', 'Other', 'College_CareerPrep', 'Music', 'History_
Geography', 'Health_LifeScience', 'EarlyDevelopment', 'ESL', 'Gym_Fitness',
'EnvironmentalScience', 'VisualArts', 'Health_Wellness', 'AppliedSciences',
'SpecialNeeds', 'Literature_Writing', 'Mathematics', 'Literacy']
Shape of matrix after one hot encodig  (50000, 30)
```

In [26]:

```
# you can do the similar thing with state, teacher_prefix and project_grade_category also
```

In [27]:

```
#school state
#Using CountVectorizer to convert values into one hot encoded
vectorizer = CountVectorizer(lowercase=False , binary=True)
vectorizer.fit(project_data['school_state'].values)
print(vectorizer.get_feature_names())

school_state_one_hot = vectorizer.transform(project_data['school_state'].values)
print('Shape of matrix after one hot encoding', school_state_one_hot.shape)
```

```
['AK', 'AL', 'AR', 'AZ', 'CA', 'CO', 'CT', 'DC', 'DE', 'FL', 'GA', 'HI', 'I
A', 'ID', 'IL', 'IN', 'KS', 'KY', 'LA', 'MA', 'MD', 'ME', 'MI', 'MN', 'MO',
'MS', 'MT', 'NC', 'ND', 'NE', 'NH', 'NJ', 'NM', 'NV', 'NY', 'OH', 'OK', 'O
R', 'PA', 'RI', 'SC', 'SD', 'TN', 'TX', 'UT', 'VA', 'VT', 'WA', 'WI', 'WV',
'WY']
Shape of matrix after one hot encoding (50000, 51)
```

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 unicode
#I fixed it by using stackoverflow.com
#https://stackoverflow.com/questions/39303912/tfidfvectorizer-in-scikit-learn-valueerror-np
print(vectorizer.get_feature_names())


teacher_prefix_one_hot = vectorizer.transform(project_data['teacher_prefix'].values.astype(
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_categ
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 project
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-pickl

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/38230349/4
```

```
084039\ndef (https://stackoverflow.com/a/38230349/4084039\ndef) loadGloveMod
el(gloveFile):\n    print ("Loading Glove Model")\n    f = open(gloveFil
e,\'r\', encoding="utf8")\n    model = {}\n    for line in tqdm(f):\n
splitLine = line.split()\n        word = splitLine[0]\n        embedding = n
p.array([float(val) for val in splitLine[1:]])\n        model[word] = embedd
ing\n    print ("Done.",len(model)," words loaded!")\n    return model\nmode
l = loadGloveModel(\'glove.42B.300d.txt\')\n\n# ============================
\nOutput:\n    \nLoading Glove Model\n1917495it [06:32, 4879.69it/s]\nDone.
 1917495  words loaded!\n\n# ============================\n\nwords = []\nfor
i in preproced_texts:\n    words.extend(i.split(\' \'))\n\nfor i in preproce
d_titles:\n    words.extend(i.split(\' \'))\nprint("all the words in the cou
pus", len(words))\nwords = set(words)\nprint("the unique words in the coupu
s", len(words))\n\ninter_words = set(model.keys()).intersection(words)\nprin
t("The number of words that are present in both glove vectors and our coupu
s",       len(inter_words),"(",np.round(len(inter_words)/len(words)*100,
3),"%)")\n\nwords_courpus = {}\nwords_glove = set(model.keys())\nfor i in wo
rds:\n    if i in words_glove:\n        words_courpus[i] = model[i]\nprint
("word 2 vec length", len(words_courpus))\n\n\n# stronging variables into pi
ckle files python: http://www.jessicayung.com/how-to-use-pickle-to-save-and-
load-variables-in-python/\n\nimport (http://www.jessicayung.com/how-to-use-p
ickle-to-save-and-load-variables-in-python/\n\nimport) pickle\nwith open(\'g
love_vectors\', \'wb\') as f:\n    pickle.dump(words_courpus, f)\n\n\n'
```

In [36]:

```
# stronging variables into pickle files python: http://www.jessicayung.com/how-to-use-pickl
# 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))
print(len(avg_w2v_vectors[0]))
```

```
100%|████████████████████████████████████████████████████████████████
███| 50000/50000 [00:24<00:00, 2069.16it/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_essays = []; # the avg-w2v for each sentence/review is stored in this lis
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((sentenc
            tf_idf = dictionary[word]*(sentence.count(word)/len(sentence.split())) # gettir
            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_essays.append(vector)

print(len(tfidf_w2v_vectors_essays))
print(len(tfidf_w2v_vectors_essays[0]))
```

```
100%|████████████████████████████████████████████████████████████████
████| 50000/50000 [02:57<00:00, 281.96it/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 lis
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((sentenc
            tf_idf = dictionary[word]*(sentence.count(word)/len(sentence.split())) # gettin
            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, 19246.10it/s]

50000
300
```

In [ ]:

```

```

1.5.3 Vectorizing Numerical features

In [42]:

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

In [43]:

```
# check this one: https://www.youtube.com/watch?v=0HOqOcln3Z4&t=530s
# standardization sklearn: https://scikit-learn.org/stable/modules/generated/sklearn.prepro
from sklearn.preprocessing import StandardScaler

# price_standardized = standardScalar.fit(project_data['price'].values)
# this will rise the error
# ValueError: Expected 2D array, got 1D array instead: array=[725.05 213.03 329.   ... 399.
# 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 standar
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],
       [-0.00088225],
       [ 0.57512161],
       ...,
       [-0.65382764],
       [-0.52109689],
       [ 0.54492668]])
```

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]:

```
# please write all the code with proper documentation, and proper titles for each subsectio
# 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
```

__ 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 stu
for learning my students learn in many different ways using all of our senses and multiple
of techniques to help all my students succeed students in my class come from a variety of d
for wonderful sharing of experiences and cultures including native americans our school is
learners which can be seen through collaborative student project based learning in and out
in my class love to work with hands on materials and have many different opportunities to p
mastered having the social skills to work cooperatively with friends is a crucial aspect of
montana is the perfect place to learn about agriculture and nutrition my students love to r
in the early childhood classroom i have had several kids ask me can we try cooking with rea
and create common core cooking lessons where we learn important math and writing concepts w
food for snack time my students will have a grounded appreciation for the work that went in
of where the ingredients came from as well as how it is healthy for their bodies this proje
nutrition and agricultural cooking recipes by having us peel our own apples to make homemad
and mix up healthy plants from our classroom garden in the spring we will also create our o
shared with families students will gain math and literature skills as well as a life long e
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
```

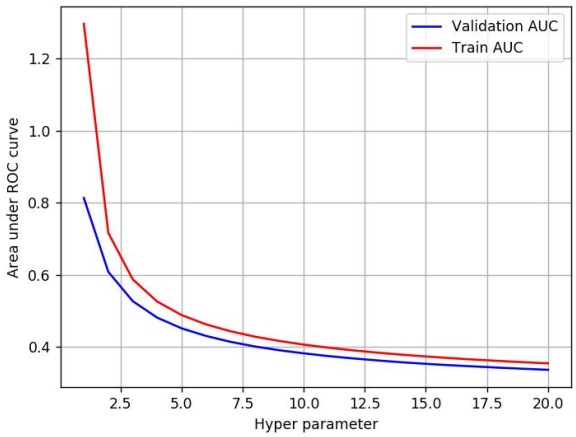
```
neg: 0.01, neu: 0.745, pos: 0.245, compound: 0.9975,
```

Assignment 7: SVM

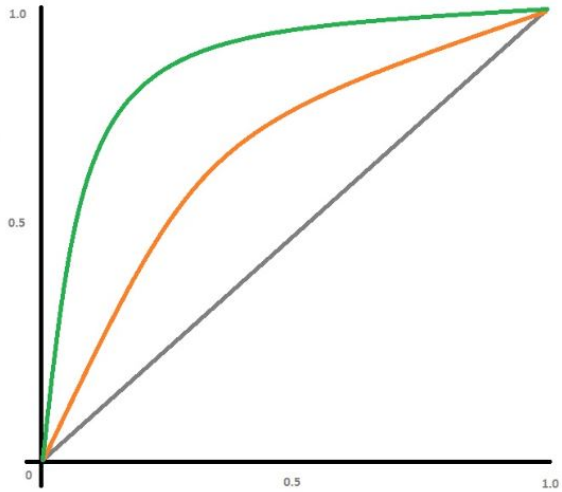
1. **[Task-1] Apply Support Vector Machines(SGDClassifier with hinge loss: Linear SVM) on these feature sets**
    - Set 1: categorical, numerical features + project_title(BOW) + preprocessed_eassay (BOW)
    - Set 2: categorical, numerical features + project_title(TFIDF)+ preprocessed_eassay (TFIDF)
    - 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. **The hyper paramter tuning (best alpha in range [10^-4 to 10^4], and the best penalty among 'l1', 'l2')**
    - Find the best hyper parameter which will give the maximum AUC (https://www.appliedaicourse.com/course/applied-ai-course-online/lessons/receiver-operating-characteristic-curve-roc-curve-and-auc-1/) value
    - Find the best hyper paramter using k-fold cross validation or simple cross validation data
    - Use gridsearch cv or randomsearch cv or you can also write your own for loops to do this task of hyperparameter tuning

3. **Representation of results**

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



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



- Along with plotting ROC curve, you need to print the [confusion matrix (https://www.appliedaicourse.com/course/applied-ai-course-online/lessons/confusion-matrix-tpr-fpr-fnr-tnr-1/)](https://www.appliedaicourse.com/course/applied-ai-course-online/lessons/confusion-matrix-tpr-fpr-fnr-tnr-1/) with predicted and original labels of test data points. Please visualize your confusion matrices using [seaborn heatmaps.](https://seaborn.pydata.org/generated/seaborn.heatmap.html)

	Predicted: NO	Predicted: YES
Actual: NO	TN = ??	FP = ??
Actual: YES	FN = ??	TP = ??

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

4. **[Task-2] Apply the Support Vector Machines on these features by finding the best hyper paramter as suggested in step 2 and step 3**

- Consider these set of features Set 5 :
  - **school_state** : categorical data
  - **clean_categories** : categorical data
  - **clean_subcategories** : categorical data
  - **project_grade_category** :categorical data

- **teacher_prefix** : categorical data
- **quantity** : numerical data
- **teacher_number_of_previously_posted_projects** : numerical data
- **price** : numerical data
- **sentiment score's of each of the essay** : numerical data
- **number of words in the title** : numerical data
- **number of words in the combine essays** : numerical data
- **Apply TruncatedSVD (http://scikit-learn.org/stable/modules/generated/sklearn.decomposition.TruncatedSVD.html) on TfidfVectorizer (https://scikit-learn.org/stable/modules/generated/sklearn.feature_extraction.text.TfidfVectorizer.html) of essay text, choose the number of components (`n_components`) using elbow method (https://www.appliedaicourse.com/course/applied-ai-course-online/lessons/pca-code-example-using-non-visualization/)** : numerical data

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

Vectorizer	Model	Hyper parameter	AUC
BOW	Brute	7	0.78
TFIDF	Brute	12	0.79
W2V	Brute	10	0.78
TFIDFW2V	Brute	6	0.78

Note: Data Leakage

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

2. Support Vector Machines

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 subsectio
# 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 cod
# 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_tr, x_test, y_tr, y_test = train_test_split(x, y, test_size=0.33, stratify=y)
x_train, x_cv, y_train, y_cv = train_test_split(x_tr, y_tr, test_size=0.33, stratify=y_tr)
```

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 subsectio
# 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 cod
# 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]:

```
#encoding numerical features


#Price
from sklearn.preprocessing import Normalizer
normalizer = Normalizer()
normalizer.fit(x_train['price'].values.reshape(1, -1))

x_train_price_norm = normalizer.transform(x_train['price'].values.reshape(1, -1))
x_test_price_norm = normalizer.transform(x_test['price'].values.reshape(1, -1))
x_cv_price_norm = normalizer.transform(x_cv['price'].values.reshape(1, -1))

print('AFter vectorizations:')
print(x_train_price_norm.shape, y_train.shape)
print(x_test_price_norm.shape, y_test.shape)
print(x_cv_price_norm.shape, y_cv.shape)
```

```
AFter vectorizations:
(1, 22445) (22445,)
(1, 16500) (16500,)
(1, 11055) (11055,)
```

In [54]:

```
#Teacher_number_of_previously_posted_projects

normalizer.fit(x_train['teacher_number_of_previously_posted_projects'].values.reshape(1, -1

x_train_previous_projects_norm = normalizer.transform(x_train['teacher_number_of_previously
x_test_previous_projects_norm = normalizer.transform(x_test['teacher_number_of_previously_p
x_cv_previous_projects_norm = normalizer.transform(x_cv['teacher_number_of_previously_poste

print('After Vectorizations:')
print(x_train_previous_projects_norm.shape, y_train.shape)
print(x_test_previous_projects_norm.shape, y_test.shape)
print(x_cv_previous_projects_norm.shape, y_cv.shape)
```

```
After Vectorizations:
(1, 22445) (22445,)
(1, 16500) (16500,)
(1, 11055) (11055,)
```

In [55]:

```
#Quantity

normalizer.fit(x_train['quantity'].values.reshape(1, -1))

x_train_quantity_norm = normalizer.transform(x_train['quantity'].values.reshape(1, -1))
x_test_quantity_norm = normalizer.transform(x_test['quantity'].values.reshape(1, -1))
x_cv_quantity_norm = normalizer.transform(x_cv['quantity'].values.reshape(1, -1))

print('After Vectorizations:')
print(x_train_quantity_norm.shape, y_train.shape)
print(x_test_quantity_norm.shape, y_test.shape)
print(x_cv_quantity_norm.shape, y_cv.shape)
```

```
After Vectorizations:
(1, 22445) (22445,)
(1, 16500) (16500,)
(1, 11055) (11055,)
```

In [56]:

```
#Encoding Categorical Features

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

x_train_state_one_hot = vectorizer.transform(x_train['school_state'].values)
x_test_state_one_hot = vectorizer.transform(x_test['school_state'].values)
x_cv_state_one_hot = vectorizer.transform(x_cv['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(x_cv_state_one_hot.shape, y_cv.shape)
```

```
After Vectorizations:
(22445, 51) (22445,)
(16500, 51) (16500,)
(11055, 51) (11055,)
```

In [57]:

```
#teacher_prefix

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

x_train_teacher_one_hot = vectorizer.transform(x_train['teacher_prefix'].values.astype('U'))
x_test_teacher_one_hot = vectorizer.transform(x_test['teacher_prefix'].values.astype('U'))
x_cv_teacher_one_hot = vectorizer.transform(x_cv['teacher_prefix'].values.astype('U'))

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

```
After Vectorizations:
(22445, 5) (22445,)
(16500, 5) (16500,)
(11055, 5) (11055,)
```

In [58]:

```
#Project_grade_category

vectorizer = CountVectorizer()
vectorizer.fit(x_train['project_grade_category'].values)

x_train_grade_one_hot = vectorizer.transform(x_train['project_grade_category'].values)
x_test_grade_one_hot = vectorizer.transform(x_test['project_grade_category'].values)
x_cv_grade_one_hot = vectorizer.transform(x_cv['project_grade_category'].values)

print("After Vectorizations:")
print(x_train_grade_one_hot.shape, y_train.shape)
print(x_test_grade_one_hot.shape, y_test.shape)
print(x_cv_grade_one_hot.shape, y_cv.shape)
```

```
After Vectorizations:
(22445, 4) (22445,)
(16500, 4) (16500,)
(11055, 4) (11055,)
```

In [59]:

```
#project_subject_categories


vectorizer = CountVectorizer()
vectorizer.fit(x_train['clean_categories'].values)

x_train_categories_one_hot = vectorizer.transform(x_train['clean_categories'].values)
x_test_categories_one_hot = vectorizer.transform(x_test['clean_categories'].values)
x_cv_categories_one_hot = vectorizer.transform(x_cv['clean_categories'].values)

print("After Vectorizations:")
print(x_train_categories_one_hot.shape, y_train.shape)
print(x_test_categories_one_hot.shape, y_test.shape)
print(x_cv_categories_one_hot.shape, y_cv.shape)
```

```
After Vectorizations:
(22445, 9) (22445,)
(16500, 9) (16500,)
(11055, 9) (11055,)
```

In [60]:

```
#project_subject_subcategories


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

x_train_subcategories_one_hot = vectorizer.transform(x_train['clean_subcategories'].values)
x_test_subcategories_one_hot = vectorizer.transform(x_test['clean_subcategories'].values)
x_cv_subcategories_one_hot = vectorizer.transform(x_cv['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)
print(x_cv_subcategories_one_hot.shape, y_cv.shape)
```

```
After Vectorizations:
(22445, 30) (22445,)
(16500, 30) (16500,)
(11055, 30) (11055,)
```

2.3 Make Data Model Ready: encoding eassay, and project_title

In [61]:

```
# please write all the code with proper documentation, and proper titles for each subsectio
# 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 cod
# 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
```

BOW: Project_title, essay

In [62]:

```
#Project_title

vectorizer = CountVectorizer(min_df=10, ngram_range=(2, 2), max_features=5000)
vectorizer.fit(x_train['project_title'].values)

x_train_title_bow = vectorizer.transform(x_train['project_title'].values)
x_test_title_bow = vectorizer.transform(x_test['project_title'].values)
x_cv_title_bow = vectorizer.transform(x_cv['project_title'].values)

print("After Vectorizations:")
print(x_train_title_bow.shape, y_train.shape)
print(x_test_title_bow.shape, y_test.shape)
print(x_cv_title_bow.shape, y_cv.shape)
```

```
After Vectorizations:
(22445, 450) (22445,)
(16500, 450) (16500,)
(11055, 450) (11055,)
```

In [63]:

```
#Essay

vectorizer = CountVectorizer(min_df=10, ngram_range=(2, 2), max_features=5000)
vectorizer.fit(x_train['essay'].values)

x_train_essay_bow = vectorizer.transform(x_train['essay'].values)
x_test_essay_bow = vectorizer.transform(x_test['essay'].values)
x_cv_essay_bow = vectorizer.transform(x_cv['essay'].values)

print("After Vectorizations:")
print(x_train_essay_bow.shape, y_train.shape)
print(x_test_essay_bow.shape, y_test.shape)
print(x_cv_essay_bow.shape, y_cv.shape)
```

```
After Vectorizations:
(22445, 5000) (22445,)
(16500, 5000) (16500,)
(11055, 5000) (11055,)
```

TFIDF: Project_title, essay

In [64]:

```
#project_title

vectorizer = TfidfVectorizer(min_df=10, ngram_range=(2, 2), max_features=5000)
vectorizer.fit(x_train['project_title'].values)

x_train_title_tfidf = vectorizer.transform(x_train['project_title'].values)
x_test_title_tfidf = vectorizer.transform(x_test['project_title'].values)
x_cv_title_tfidf = vectorizer.transform(x_cv['project_title'].values)

print("After Vectorizations:")
print(x_train_title_tfidf.shape, y_train.shape)
print(x_test_title_tfidf.shape, y_test.shape)
print(x_cv_title_tfidf.shape, y_cv.shape)
```

```
After Vectorizations:
(22445, 450) (22445,)
(16500, 450) (16500,)
(11055, 450) (11055,)
```

In [65]:

```
#essay

from sklearn.feature_extraction.text import TfidfVectorizer
vectorizer = TfidfVectorizer(min_df=10, ngram_range=(2, 2), max_features=5000)
vectorizer.fit(x_train['essay'].values)

x_train_essay_tfidf = vectorizer.transform(x_train['essay'].values)
x_test_essay_tfidf = vectorizer.transform(x_test['essay'].values)
x_cv_essay_tfidf = vectorizer.transform(x_cv['essay'].values)

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

```
After Vectorizations:
(22445, 5000) (22445,)
(16500, 5000) (16500,)
(11055, 5000) (11055,)
```

AVG-W2V: Project_title, essay

In [66]:

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

Project_title

In [67]:

```
x_train_title_avg_w2v = []; # the avg-w2v for each sentence/review is stored in this list
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%|████████████████████████████████████████████████████████████████
██| 22445/22445 [00:00<00:00, 34991.07it/s]

22445
300
```

In [68]:

```
x_test_title_avg_w2v = []; # the avg-w2v for each sentence/review is stored in this list
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, 37252.98it/s]

16500
300
```

In [69]:

```
x_cv_title_avg_w2v = []; # the avg-w2v for each sentence/review is stored in this list
for sentence in tqdm(x_cv['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_cv_title_avg_w2v.append(vector)
```

```
100%|████████████████████████████████████████████████████████████████████
██| 11055/11055 [00:00<00:00, 39874.58it/s]
```

Essay

In [70]:

```
# average Word2Vec
# compute average word2vec for each review.
x_train_essay_avg_w2v = []; # the avg-w2v for each sentence/review is stored in this list
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%|████████████████████████████████████████████████████████████████████
███| 22445/22445 [00:11<00:00, 2029.06it/s]

22445
300
```

In [71]:

```
x_test_essay_avg_w2v = []; # the avg-w2v for each sentence/review is stored in this list
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 [00:08<00:00, 1984.40it/s]
```

In [72]:

```
x_cv_essay_avg_w2v = []; # the avg-w2v for each sentence/review is stored in this list
for sentence in tqdm(x_cv['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_cv_essay_avg_w2v.append(vector)
```

```
100%|████████████████████████████████████████████████████████████████
███| 11055/11055 [00:05<00:00, 2008.19it/s]
```

TFIDF-W2V: Project_title, 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((sentenc
            tf_idf = dictionary[word]*(sentence.count(word)/len(sentence.split())) # gettin
            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%|████████████████████████████████████████████████████████████████
██| 22445/22445 [00:01<00:00, 18273.37it/s]

22445
```

In [75]:

```
x_test_title_tfidf_w2v = []; # the avg-w2v for each sentence/review is stored in this list
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((sentenc
            tf_idf = dictionary[word]*(sentence.count(word)/len(sentence.split())) # gettin
            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, 18361.50it/s]
```

In [76]:

```
x_cv_title_tfidf_w2v = []; # the avg-w2v for each sentence/review is stored in this list
for sentence in tqdm(x_cv['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((sentenc
            tf_idf = dictionary[word]*(sentence.count(word)/len(sentence.split())) # gettir
            vector += (vec * tf_idf) # calculating tfidf weighted w2v
            tf_idf_weight += tf_idf
    if tf_idf_weight != 0:
        vector /= tf_idf_weight
    x_cv_title_tfidf_w2v.append(vector)
```

```
100%|████████████████████████████████████████████████████████████████████
██| 11055/11055 [00:00<00:00, 16712.54it/s]
```

Essay

In [77]:

```
#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 [78]:

```
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((sentenc
            tf_idf = dictionary[word]*(sentence.count(word)/len(sentence.split())) # gettir
            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%|████████████████████████████████████████████████████████████████████
████| 22445/22445 [01:16<00:00, 295.27it/s]

22445
```

In [79]:

```
x_test_essay_tfidf_w2v = []; # the avg-w2v for each sentence/review is stored in this list
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((sentenc
            tf_idf = dictionary[word]*(sentence.count(word)/len(sentence.split())) # gettin
            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 [00:57<00:00, 286.25it/s]
```

In [80]:

```
x_cv_essay_tfidf_w2v = []; # the avg-w2v for each sentence/review is stored in this list
for sentence in tqdm(x_cv['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((sentenc
            tf_idf = dictionary[word]*(sentence.count(word)/len(sentence.split())) # gettin
            vector += (vec * tf_idf) # calculating tfidf weighted w2v
            tf_idf_weight += tf_idf
    if tf_idf_weight != 0:
        vector /= tf_idf_weight
    x_cv_essay_tfidf_w2v.append(vector)
```

```
100%|████████████████████████████████████████████████████████████████████████████
████| 11055/11055 [00:39<00:00, 280.79it/s]
```

2.4 Appling Support Vector Machines on different kind of featurization as mentioned in the instructions

Apply Support Vector Machines 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 [81]:

```
# please write all the code with proper documentation, and proper titles for each subsectio
# 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 cod
# 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 [82]:

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

x_test_price_norm = x_test_price_norm.reshape(-1,1)
x_test_previous_projects_norm = x_test_previous_projects_norm.reshape(-1,1)
x_test_quantity_norm = x_test_quantity_norm.reshape(-1, 1)

x_cv_price_norm = x_cv_price_norm.reshape(-1,1)
x_cv_previous_projects_norm = x_cv_previous_projects_norm.reshape(-1,1)
x_cv_quantity_norm = x_cv_quantity_norm.reshape(-1, 1)
```

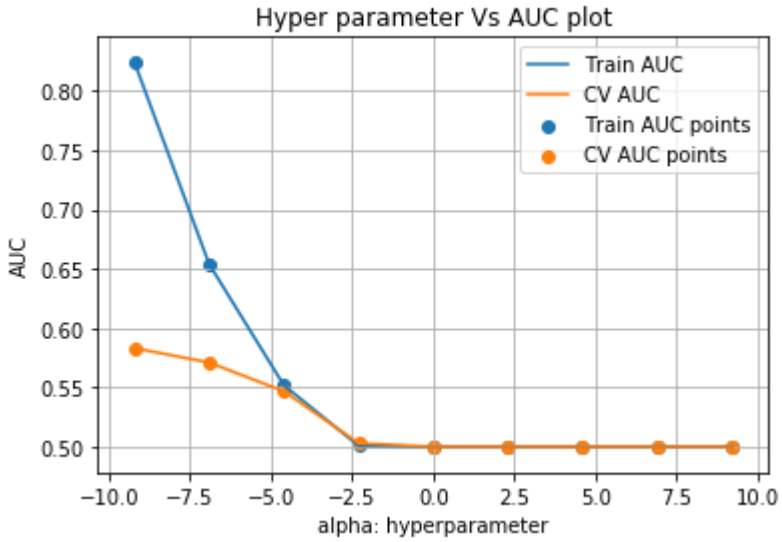
Set-1

In [83]:

```
#merging features

from scipy.sparse import hstack
x_train_bow = hstack((x_train_price_norm, x_train_previous_projects_norm, x_train_state_one
                      x_train_grade_one_hot, x_train_categories_one_hot, x_train_subcategori
                      x_train_title_bow)).tocsr()

x_test_bow = hstack((x_test_price_norm, x_test_previous_projects_norm, x_test_state_one_hot
                      x_test_grade_one_hot, x_test_categories_one_hot, x_test_subcategories_
                      x_test_title_bow)).tocsr()

x_cv_bow = hstack((x_cv_price_norm, x_cv_previous_projects_norm, x_cv_state_one_hot, x_cv_t
                      x_cv_grade_one_hot, x_cv_categories_one_hot, x_cv_subcategories_one_ho
                      x_cv_title_bow)).tocsr()
```

In [84]:

```
#https://www.appliedaicourse.com/lecture/11/applied-machine-learning-online-course/3061/ass
#https://stackoverflow.com/questions/52982340/using-gridsearchcv-gives-an-error-with-calibr

from sklearn.model_selection import train_test_split
from sklearn.model_selection import GridSearchCV
from sklearn.linear_model import SGDClassifier
from sklearn.calibration import CalibratedClassifierCV
from sklearn.datasets import *

tuned_parameters = {'alpha':[10**-4, 10**-3, 10**-2, 10**-1, 10**0, 10**1, 10**2, 10**3, 10
sgd = SGDClassifier(loss='hinge', class_weight='balanced')
clf = GridSearchCV(sgd, tuned_parameters, cv=10, refit=True, scoring='roc_auc')
clf.fit(x_train_bow, y_train)

calibrated_clf = CalibratedClassifierCV(clf, cv='prefit')
calibrated_clf.fit(x_train_bow, y_train)

results = pd.DataFrame.from_dict(calibrated_clf.base_estimator.cv_results_)
results = results.sort_values(['param_alpha'])

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']
alpha =  results['param_alpha']

for i in range(len(alpha)):
    alpha[i] = np.log(alpha[i])
plt.plot(alpha, 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,c

plt.plot(alpha, 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='darkor

plt.scatter(alpha, train_auc, label='Train AUC points')
plt.scatter(alpha, cv_auc, label='CV AUC points')


plt.legend()
plt.xlabel("alpha: 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_alpha	param_penalty
0	0.152552	0.016611	0.003518	0.004189	-9.21034	l1
1	0.136772	0.013299	0.002607	0.000200	-6.90776	l1
2	0.101227	0.010900	0.002804	0.000510	-4.60517	l1
3	0.091694	0.011087	0.002206	0.000460	-2.30259	l1
4	0.093950	0.010742	0.002105	0.000200	0	l1

5 rows × 32 columns

In [85]:

```
calibrated_clf.base_estimator.best_params_
```

Out[85]:

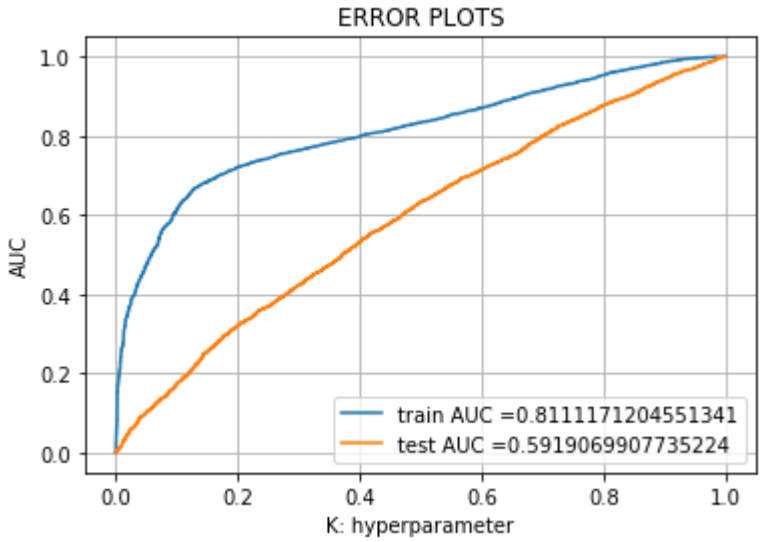
```
{'alpha': 0.0001, 'penalty': 'l1'}
```

In [87]:

```
#Best alpha
best_alpha = 0.0001
```

In [89]:

```
#Hyper Parameter Tuning with Best-alpha

from sklearn.metrics import roc_curve, auc

sgd = SGDClassifier(loss='hinge', penalty='l1', alpha=best_alpha, class_weight='balanced')
sgd.fit(x_train_bow, y_train)
calibrated_clf = CalibratedClassifierCV(sgd, cv='prefit')
calibrated_clf.fit(x_train_bow, y_train)

y_train_pred = calibrated_clf.predict_proba(x_train_bow) [:, 1]
y_test_pred = calibrated_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 [90]:

```
# 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.round(t
    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 [91]:

```
#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)))
```

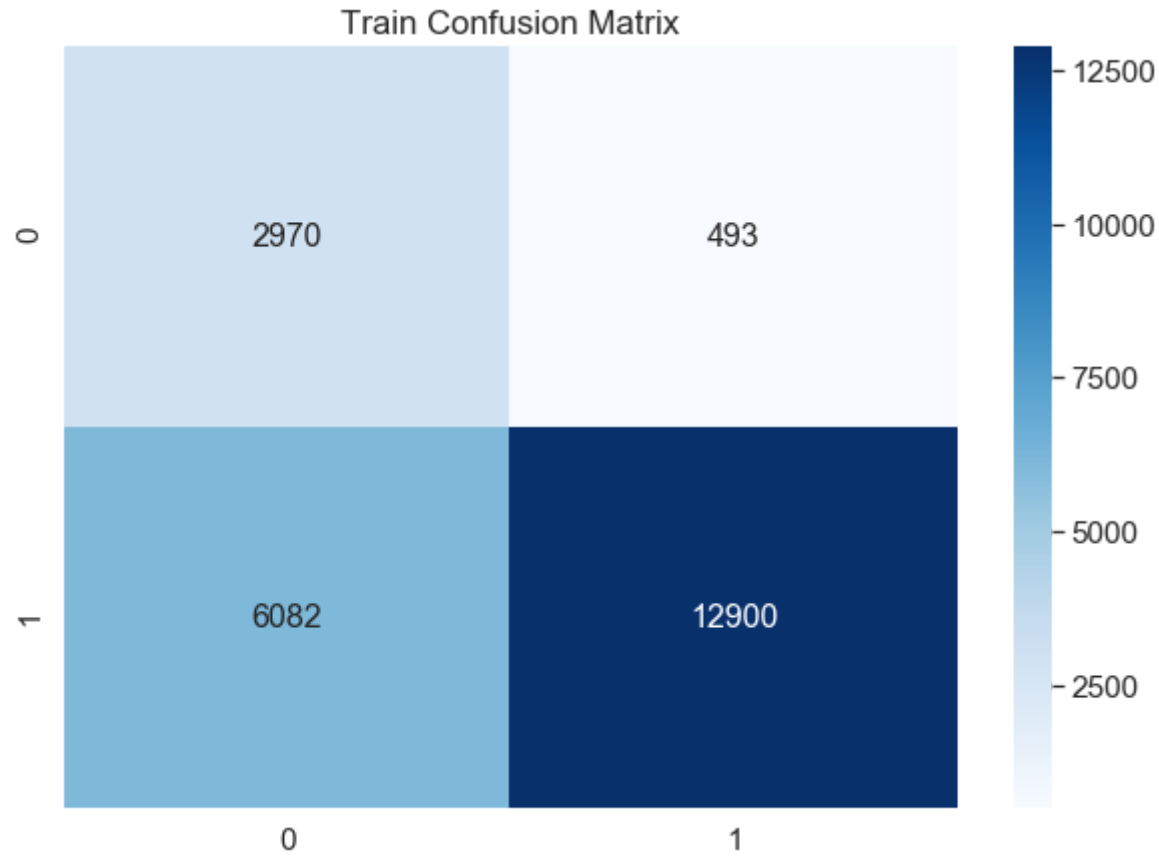
```
the maximum value of tpr*(1-fpr) 0.5828431531088939 for threshold 0.855
Train confusion matrix
[[ 2970   493]
 [ 6082 12900]]
Test confusion matrix
[[1273 1273]
 [5133 8821]]
```

In [92]:

```
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)),

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[92]:

```
<matplotlib.axes._subplots.AxesSubplot at 0x184b9fd6780>
```

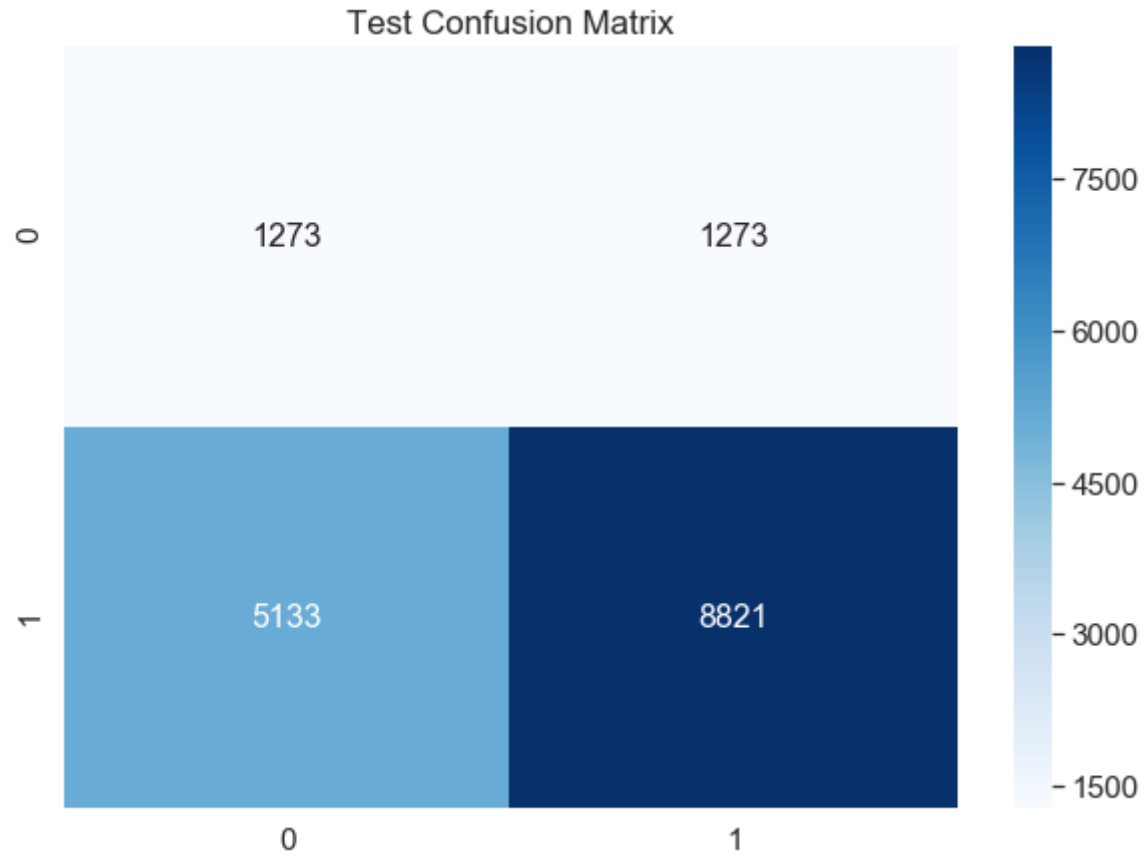


In [93]:

```
#Test Confusion matrix
df_cm = pd.DataFrame(confusion_matrix(y_test, predict_with_best_t(y_test_pred, best_t)), cc

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[93]:

```
<matplotlib.axes._subplots.AxesSubplot at 0x184bb94ea20>
```



In [ ]:

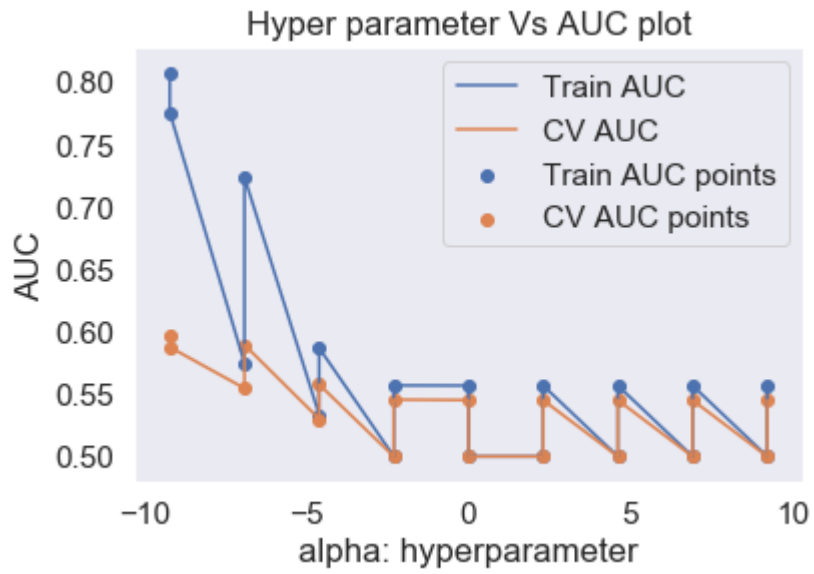
Set-2

In [94]:

```
#Merging Features

x_train_tfidf = hstack((x_train_price_norm, x_train_previous_projects_norm, x_train_state_c
                    x_train_grade_one_hot, x_train_categories_one_hot, x_train_subcategori
                    x_train_title_tfidf)).tocsr()

x_test_tfidf = hstack((x_test_price_norm, x_test_previous_projects_norm, x_test_state_one_h
                    x_test_grade_one_hot, x_test_categories_one_hot, x_test_subcategories_
                    x_test_title_tfidf)).tocsr()

x_cv_tfidf = hstack((x_cv_price_norm, x_cv_previous_projects_norm, x_cv_state_one_hot, x_cv
                    x_cv_grade_one_hot, x_cv_categories_one_hot, x_cv_subcategories_one_ho
                    x_cv_title_tfidf)).tocsr()
```

In [96]:

```
#Hyperparameter Tuning

#https://www.appliedaicourse.com/lecture/11/applied-machine-learning-online-course/3061/ass
#https://stackoverflow.com/questions/52982340/using-gridsearchcv-gives-an-error-with-calibr

from sklearn.model_selection import train_test_split
from sklearn.model_selection import GridSearchCV
from sklearn.linear_model import SGDClassifier
from sklearn.calibration import CalibratedClassifierCV
from sklearn.datasets import *

tuned_parameters = {'alpha':[10**-4, 10**-3, 10**-2, 10**-1, 10**0, 10**1, 10**2, 10**3, 10
sgd = SGDClassifier(loss='hinge', class_weight='balanced')
clf = GridSearchCV(sgd, tuned_parameters, cv=3, refit=True, scoring='roc_auc')
clf.fit(x_train_tfidf, y_train)

calibrated_clf = CalibratedClassifierCV(clf, cv='prefit')
calibrated_clf.fit(x_train_tfidf, y_train)

results = pd.DataFrame.from_dict(calibrated_clf.base_estimator.cv_results_)
results = results.sort_values(['param_alpha'])

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']
alpha =  results['param_alpha']

for i in range(len(alpha)):
    alpha[i] = np.log(alpha[i])
plt.plot(alpha, 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,c

plt.plot(alpha, 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='darkor

plt.scatter(alpha, train_auc, label='Train AUC points')
plt.scatter(alpha, cv_auc, label='CV AUC points')


plt.legend()
plt.xlabel("alpha: hyperparameter")
plt.ylabel("AUC")
plt.title("Hyper parameter Vs AUC plot")
plt.grid()
plt.show()

results.head()
```



Out[96]:

	mean_fit_time	std_fit_time	mean_score_time	std_score_time	param_alpha	param_penalty
0	0.119654	0.003434	0.005682	0.000236	-9.21034	l1
1	0.084225	0.023857	0.007854	0.002328	-9.21034	l2
2	0.075212	0.001802	0.005001	0.000017	-6.90776	l1
3	0.049633	0.000819	0.005692	0.000639	-6.90776	l2
4	0.075407	0.004396	0.005490	0.000017	-4.60517	l1

In [97]:

```
#Best-Alpha
calibrated_clf.base_estimator.best_params_
```

Out[97]:

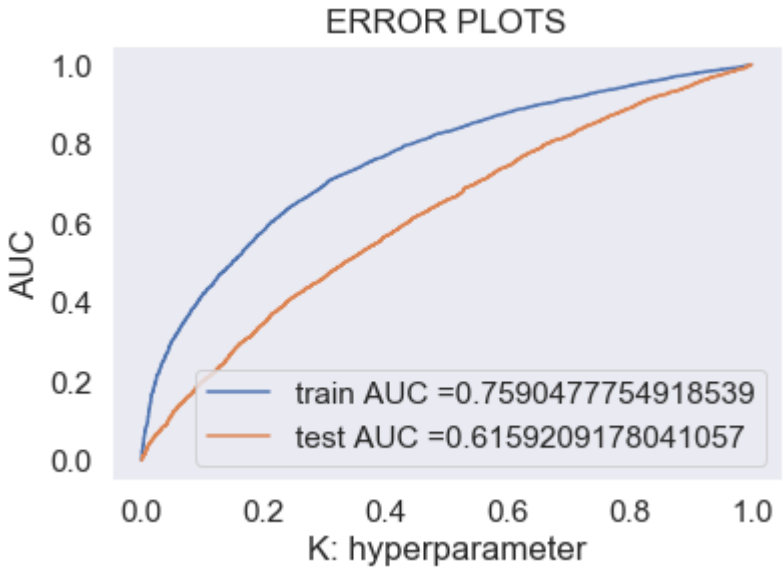
```
{'alpha': 0.0001, 'penalty': 'l1'}
```

In [98]:

```
best_alpha = 0.0001
```

In [99]:

```
#Hyper Parameter Tuning with Best-alpha

from sklearn.metrics import roc_curve, auc

sgd = SGDClassifier(loss='hinge', penalty='l1', alpha=best_alpha, class_weight='balanced')
sgd.fit(x_train_tfidf, y_train)
calibrated_clf = CalibratedClassifierCV(sgd, cv='prefit')
calibrated_clf.fit(x_train_tfidf, y_train)

y_train_pred = calibrated_clf.predict_proba(x_train_tfidf) [:, 1]
y_test_pred = calibrated_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 [100]:

```
# 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.round(t
    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 [101]:

```
#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)))
```

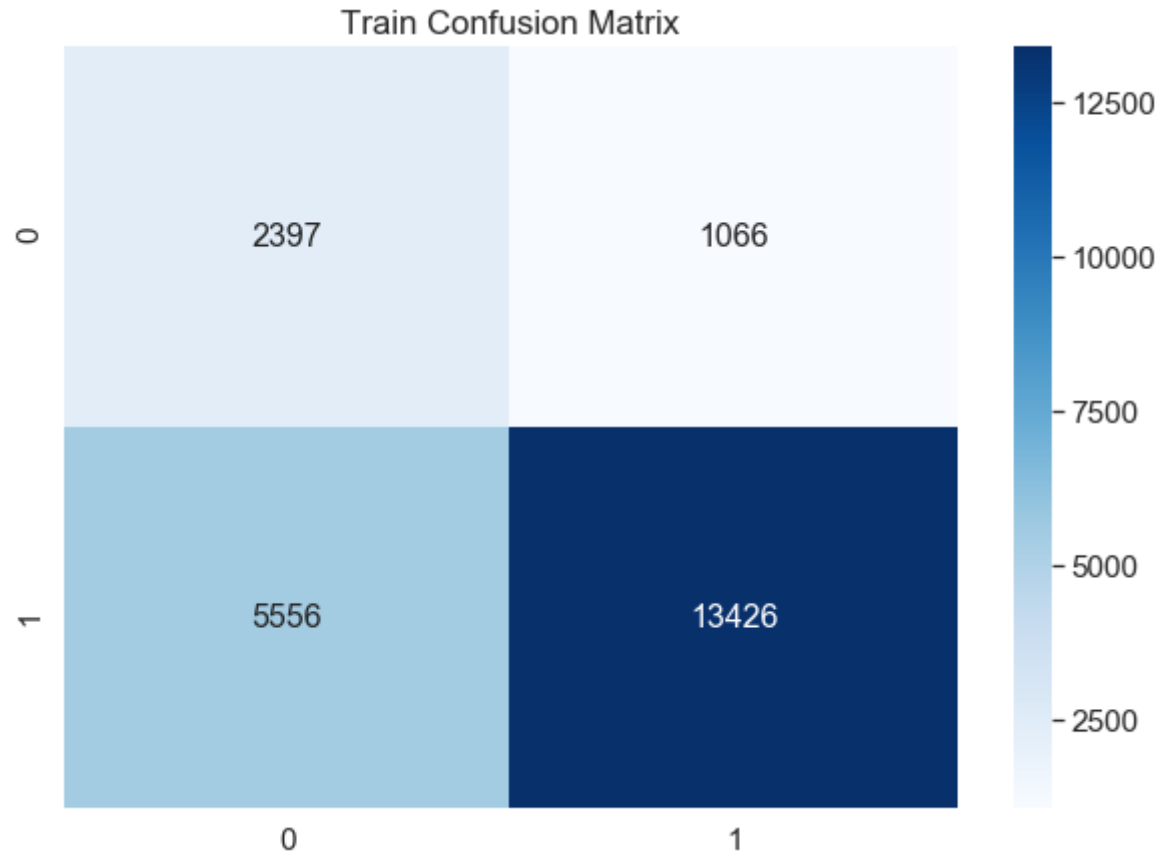
```
the maximum value of tpr*(1-fpr) 0.4895761088981574 for threshold 0.839
Train confusion matrix
[[ 2397  1066]
 [ 5556 13426]]
Test confusion matrix
[[1213 1333]
 [4530 9424]]
```

In [102]:

```
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)),

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[102]:

```
<matplotlib.axes._subplots.AxesSubplot at 0x184ba0a6358>
```

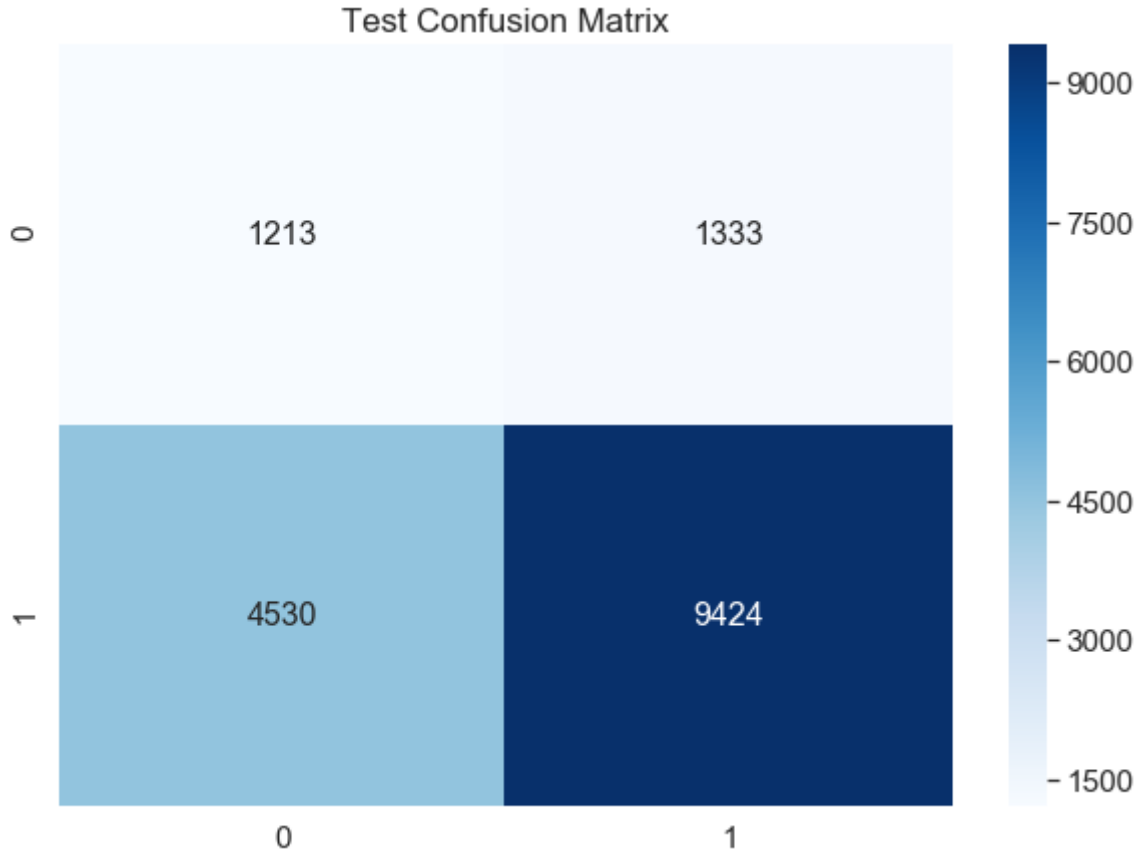


In [103]:

```
#Test Confusion matrix
df_cm = pd.DataFrame(confusion_matrix(y_test, predict_with_best_t(y_test_pred, best_t)), cc

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[103]:

```
<matplotlib.axes._subplots.AxesSubplot at 0x184ba13b7f0>
```



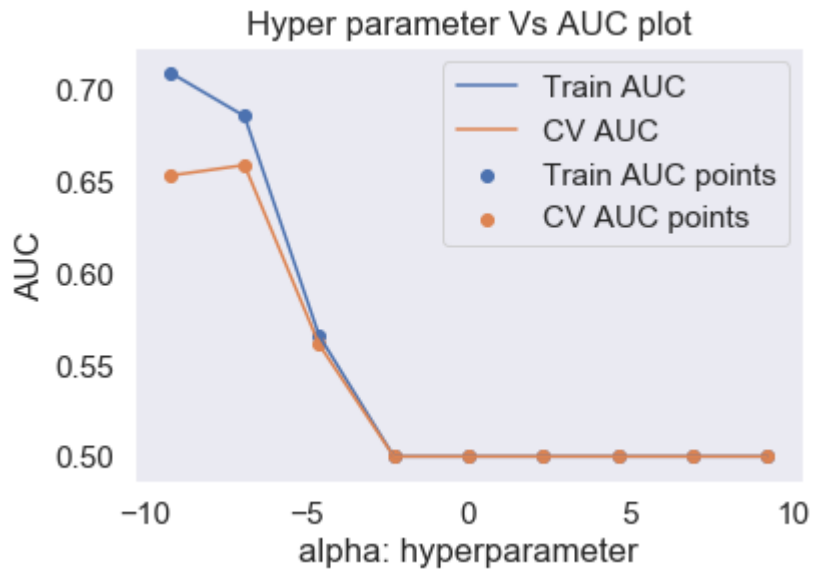
Set-3

In [104]:

```
#Merging Features

x_train_avg_w2v = hstack((x_train_price_norm, x_train_previous_projects_norm, x_train_state
                    x_train_grade_one_hot, x_train_categories_one_hot, x_train_subcategori
                    x_train_title_avg_w2v)).tocsr()

x_test_avg_w2v = hstack((x_test_price_norm, x_test_previous_projects_norm, x_test_state_one
                    x_test_grade_one_hot, x_test_categories_one_hot, x_test_subcategories_
                    x_test_title_avg_w2v)).tocsr()

x_cv_avg_w2v = hstack((x_cv_price_norm, x_cv_previous_projects_norm, x_cv_state_one_hot, x_
                    x_cv_grade_one_hot, x_cv_categories_one_hot, x_cv_subcategories_one_hc
                    x_cv_title_avg_w2v)).tocsr()
```

In [105]:

```
#Hyperparameter Tuning

from sklearn.model_selection import train_test_split
from sklearn.model_selection import GridSearchCV
from sklearn.linear_model import SGDClassifier
from sklearn.calibration import CalibratedClassifierCV
from sklearn.datasets import *

tuned_parameters = {'alpha':[10**-4, 10**-3, 10**-2, 10**-1, 10**0, 10**1, 10**2, 10**3, 10
sgd = SGDClassifier(loss='hinge', class_weight='balanced')
clf = GridSearchCV(sgd, tuned_parameters, cv=10, refit=True, scoring='roc_auc')
clf.fit(x_train_avg_w2v, y_train)

calibrated_clf = CalibratedClassifierCV(clf, cv='prefit')
calibrated_clf.fit(x_train_avg_w2v, y_train)

results = pd.DataFrame.from_dict(calibrated_clf.base_estimator.cv_results_)
results = results.sort_values(['param_alpha'])

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']
alpha =  results['param_alpha']

for i in range(len(alpha)):
    alpha[i] = np.log(alpha[i])
plt.plot(alpha, 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,c

plt.plot(alpha, 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='darkor

plt.scatter(alpha, train_auc, label='Train AUC points')
plt.scatter(alpha, cv_auc, label='CV AUC points')


plt.legend()
plt.xlabel("alpha: hyperparameter")
plt.ylabel("AUC")
plt.title("Hyper parameter Vs AUC plot")
plt.grid()
plt.show()

results.head()
```



Out[105]:

	mean_fit_time	std_fit_time	mean_score_time	std_score_time	param_alpha	param_penalty
0	1.000378	0.047500	0.009000	0.006310	-9.21034	l1
1	1.249034	0.045182	0.008193	0.004302	-6.90776	l1
2	1.274753	0.011197	0.007484	0.004762	-4.60517	l1
3	1.285261	0.006086	0.007737	0.004407	-2.30259	l1
4	1.321795	0.075063	0.010139	0.009210	0	l1

5 rows × 32 columns

In [106]:

```
calibrated_clf.base_estimator.best_params_
```

Out[106]:

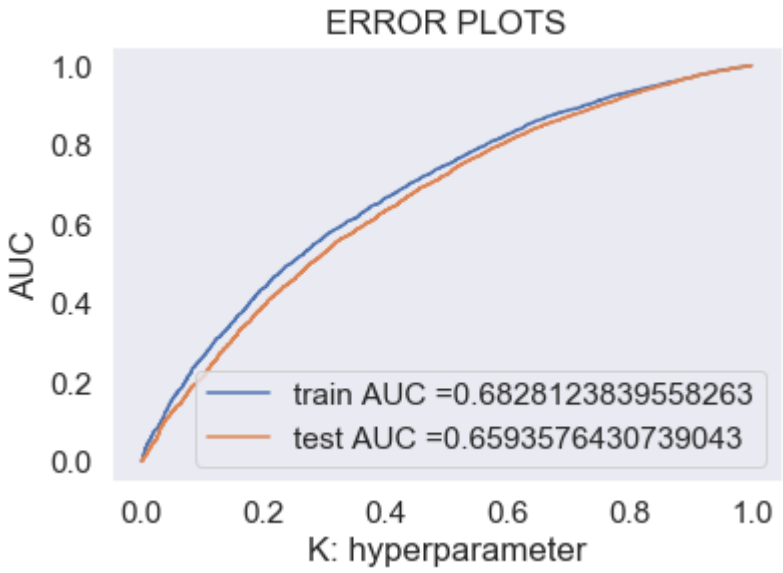
```
{'alpha': 0.001, 'penalty': 'l1'}
```

In [108]:

```
#Best-alpha
best_alpha = 0.001
```

In [109]:

```
#Hyperparameter Tuning with Best-alpha

from sklearn.metrics import roc_curve, auc

sgd = SGDClassifier(loss='hinge', penalty='l1', alpha=best_alpha, class_weight='balanced')
sgd.fit(x_train_avg_w2v, y_train)
calibrated_clf = CalibratedClassifierCV(sgd, cv='prefit')
calibrated_clf.fit(x_train_avg_w2v, y_train)

y_train_pred = calibrated_clf.predict_proba(x_train_avg_w2v) [:, 1]
y_test_pred = calibrated_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 [110]:

```
# 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.round(t
    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 [111]:

```
#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)))
```

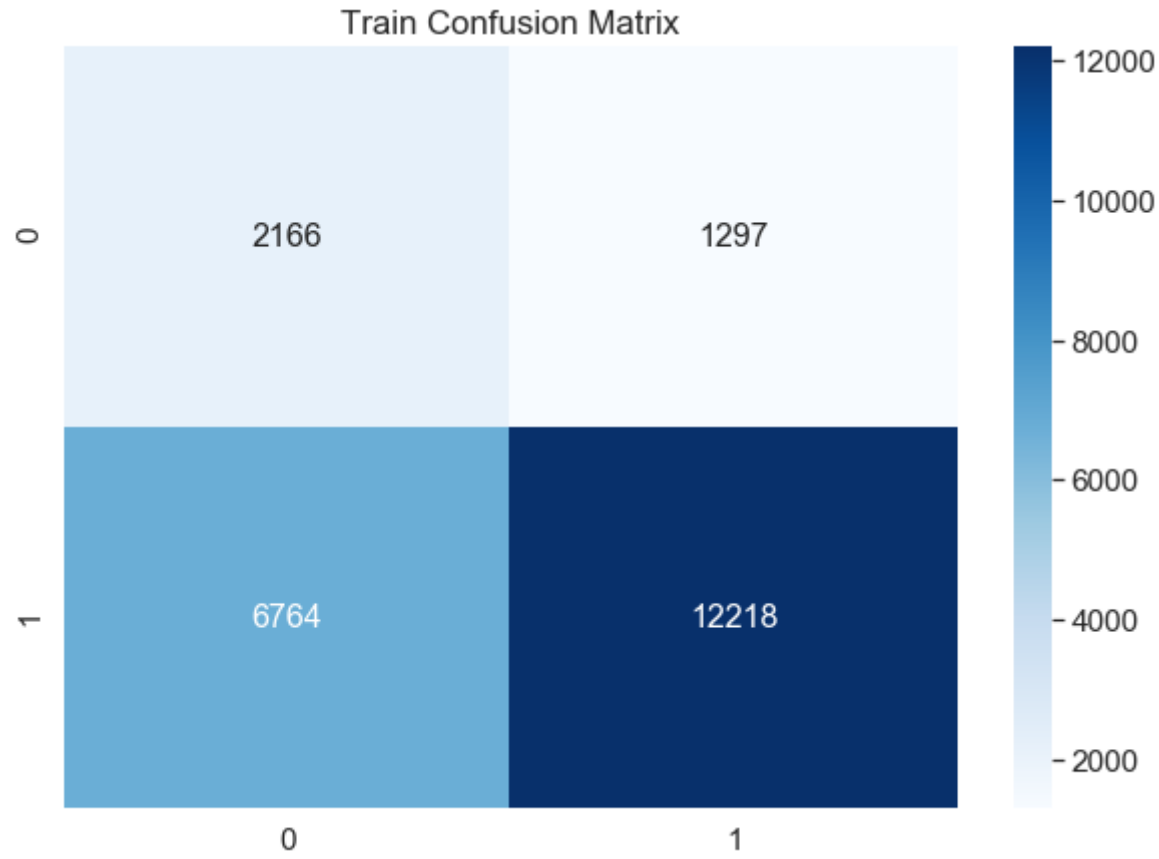
```
the maximum value of tpr*(1-fpr) 0.40259104686102765 for threshold 0.841
Train confusion matrix
[[ 2166  1297]
 [ 6764 12218]]
Test confusion matrix
[[1521 1025]
 [5079 8875]]
```

In [112]:

```
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)),

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[112]:

```
<matplotlib.axes._subplots.AxesSubplot at 0x184ba30f2b0>
```

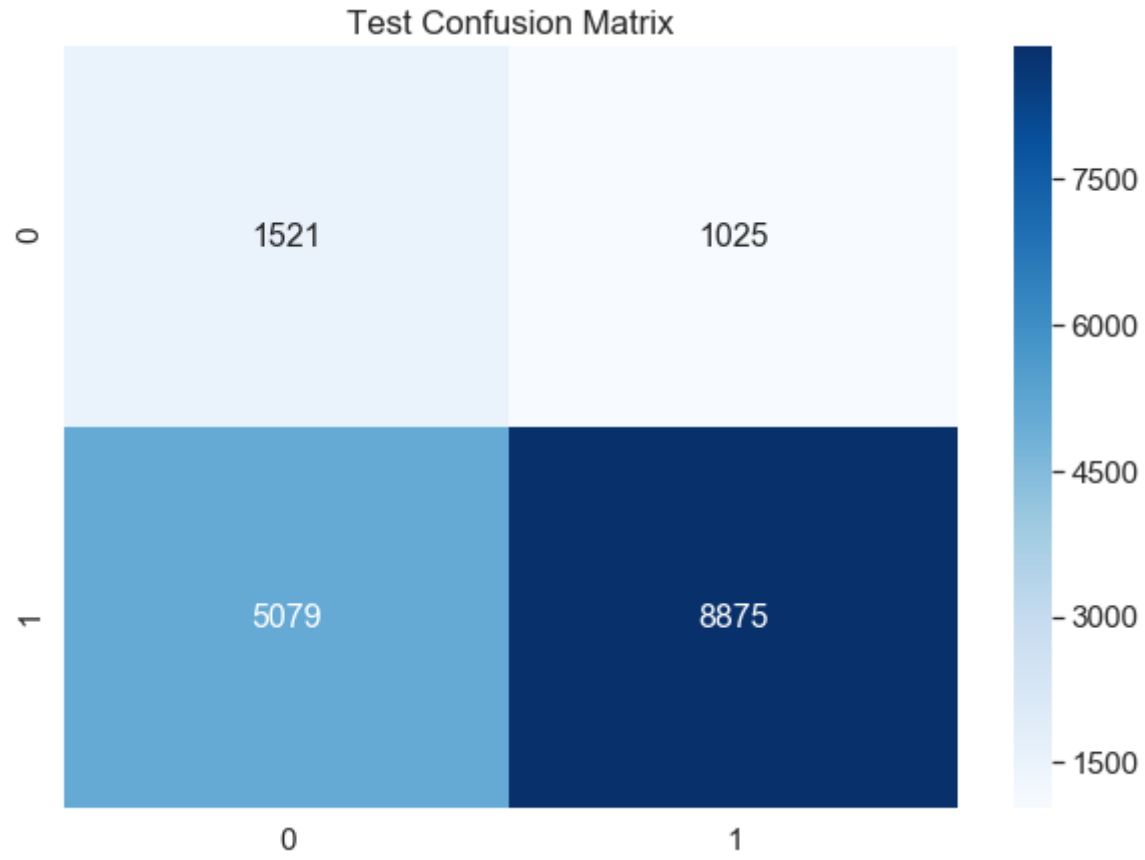


In [113]:

```
#Test Confusion matrix
df_cm = pd.DataFrame(confusion_matrix(y_test, predict_with_best_t(y_test_pred, best_t)), cc

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[113]:

```
<matplotlib.axes._subplots.AxesSubplot at 0x184bba13160>
```



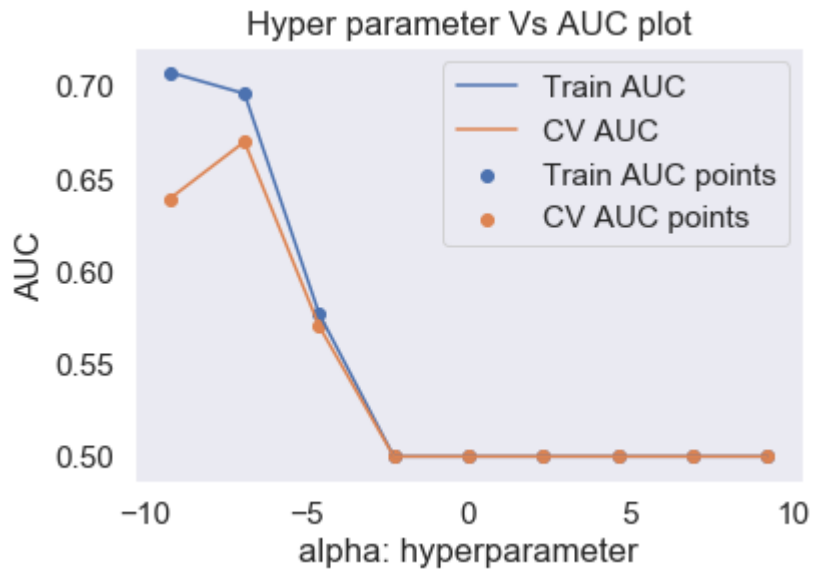
Set-4

In [114]:

```
#Merging Features

x_train_tfidf_w2v = hstack((x_train_price_norm, x_train_previous_projects_norm, x_train_sta
                    x_train_grade_one_hot, x_train_categories_one_hot, x_train_subcategori
                    x_train_title_tfidf_w2v)).tocsr()

x_test_tfidf_w2v = hstack((x_test_price_norm, x_test_previous_projects_norm, x_test_state_c
                    x_test_grade_one_hot, x_test_categories_one_hot, x_test_subcategories_
                    x_test_title_tfidf_w2v)).tocsr()

x_cv_tfidf_w2v = hstack((x_cv_price_norm, x_cv_previous_projects_norm, x_cv_state_one_hot,
                    x_cv_grade_one_hot, x_cv_categories_one_hot, x_cv_subcategories_one_hc
                    x_cv_title_tfidf_w2v)).tocsr()
```

In [115]:

```
#Hyperparameter Tuning

from sklearn.model_selection import train_test_split
from sklearn.model_selection import GridSearchCV
from sklearn.linear_model import SGDClassifier
from sklearn.calibration import CalibratedClassifierCV
from sklearn.datasets import *

tuned_parameters = {'alpha':[10**-4, 10**-3, 10**-2, 10**-1, 10**0, 10**1, 10**2, 10**3, 10
sgd = SGDClassifier(loss='hinge', class_weight='balanced')
clf = GridSearchCV(sgd, tuned_parameters, cv=10, refit=True, scoring='roc_auc')
clf.fit(x_train_tfidf_w2v, y_train)

calibrated_clf = CalibratedClassifierCV(clf, cv='prefit')
calibrated_clf.fit(x_train_tfidf_w2v, y_train)

results = pd.DataFrame.from_dict(calibrated_clf.base_estimator.cv_results_)
results = results.sort_values(['param_alpha'])

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']
alpha =  results['param_alpha']

for i in range(len(alpha)):
    alpha[i] = np.log(alpha[i])
plt.plot(alpha, 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,c

plt.plot(alpha, 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='darkor

plt.scatter(alpha, train_auc, label='Train AUC points')
plt.scatter(alpha, cv_auc, label='CV AUC points')


plt.legend()
plt.xlabel("alpha: hyperparameter")
plt.ylabel("AUC")
plt.title("Hyper parameter Vs AUC plot")
plt.grid()
plt.show()

results.head()
```



Out[115]:

	mean_fit_time	std_fit_time	mean_score_time	std_score_time	param_alpha	param_penalty
0	1.048197	0.108390	0.008529	0.005532	-9.21034	l1
1	1.299528	0.067493	0.007025	0.004206	-6.90776	l1
2	1.297707	0.023057	0.008835	0.003402	-4.60517	l1
3	1.309446	0.010214	0.005727	0.004556	-2.30259	l1
4	1.329276	0.030665	0.004461	0.002995	0	l1

5 rows × 32 columns

In [116]:

```
calibrated_clf.base_estimator.best_params_
```

Out[116]:

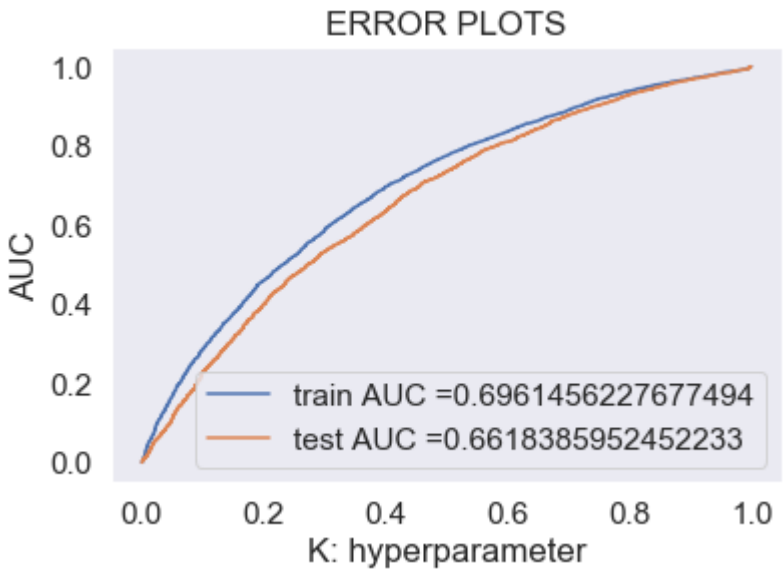
```
{'alpha': 0.001, 'penalty': 'l1'}
```

In [117]:

```
#Best-alpha
best_alpha = 0.001
```

In [118]:

```
#Hyperparameter Tuning with Best-alpha

from sklearn.metrics import roc_curve, auc

sgd = SGDClassifier(loss='hinge', penalty='l1', alpha=best_alpha, class_weight='balanced')
sgd.fit(x_train_tfidf_w2v, y_train)
calibrated_clf = CalibratedClassifierCV(sgd, cv='prefit')
calibrated_clf.fit(x_train_tfidf_w2v, y_train)

y_train_pred = calibrated_clf.predict_proba(x_train_tfidf_w2v) [:, 1]
y_test_pred = calibrated_clf.predict_proba(x_test_tfidf_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 [119]:

```
# 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.round(t
    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 [120]:

```
#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)))
```

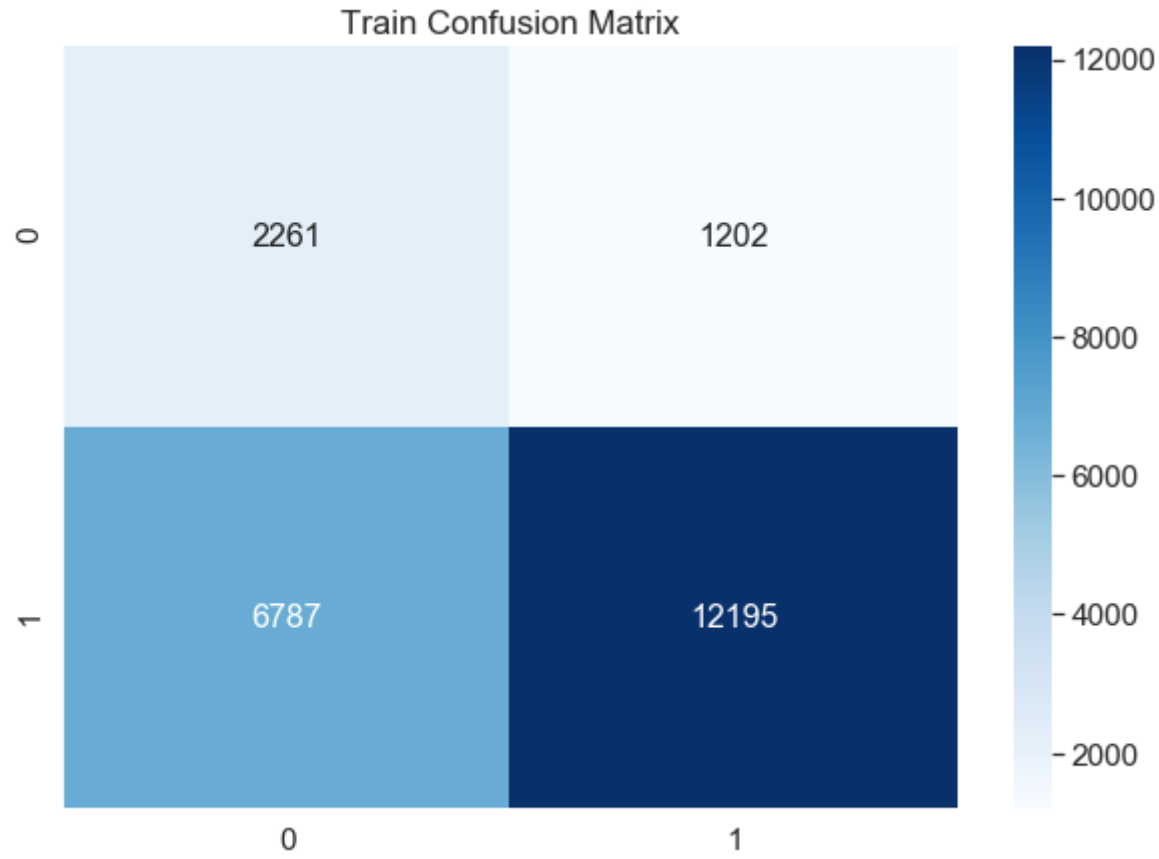
```
the maximum value of tpr*(1-fpr) 0.41945744426540477 for threshold 0.84
Train confusion matrix
[[ 2261  1202]
 [ 6787 12195]]
Test confusion matrix
[[1518 1028]
 [5033 8921]]
```

In [121]:

```
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)),

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[121]:

```
<matplotlib.axes._subplots.AxesSubplot at 0x184bb9f75c0>
```

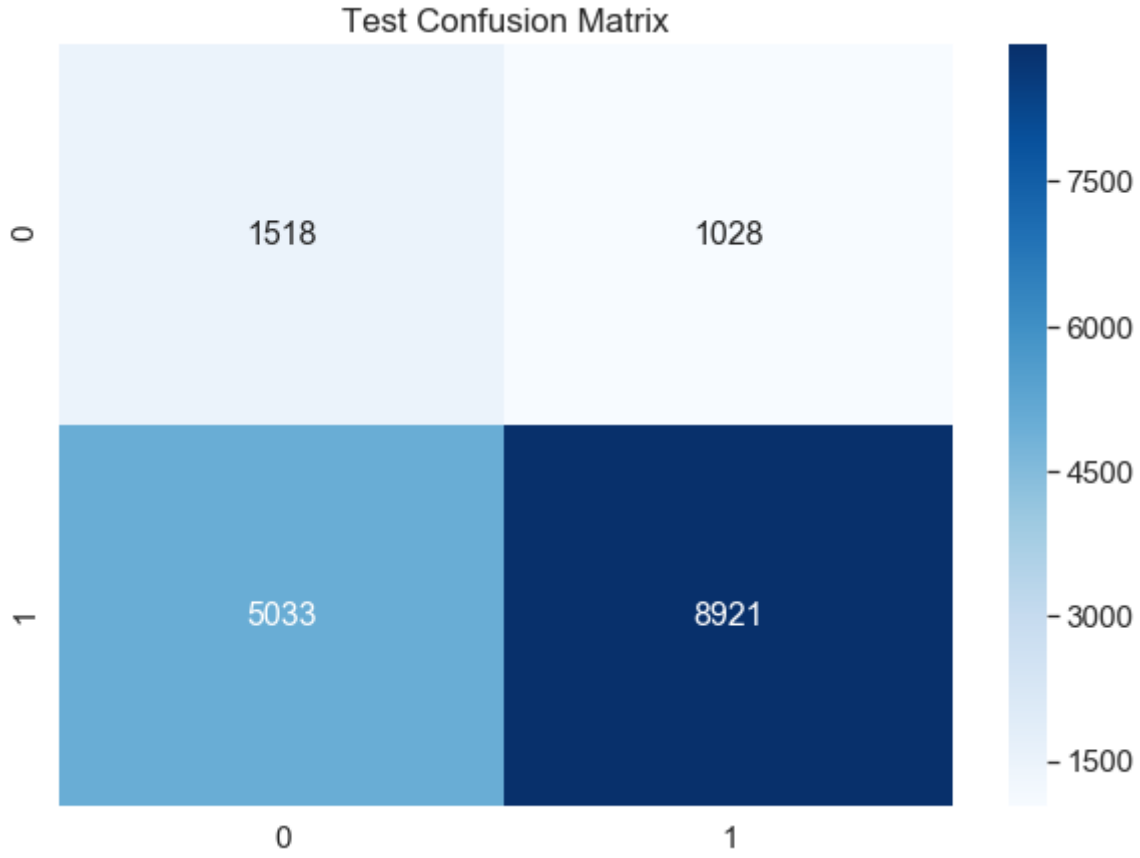


In [122]:

```
#Test Confusion matrix
df_cm = pd.DataFrame(confusion_matrix(y_test, predict_with_best_t(y_test_pred, best_t)), cc

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[122]:

```
<matplotlib.axes._subplots.AxesSubplot at 0x184ba2a0c50>
```



2.5 Support Vector Machines with added Features `Set 5`

In [123]:

```
# please write all the code with proper documentation, and proper titles for each subsectio
# 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 cod
# 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 [124]:

```
#Sentiment Score
```

In [125]:

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

sentiment_neg=[]
sentiment_neu=[]
sentiment_pos=[]
sentiment_comp=[]

sid = SIA()
for sentence in tqdm(project_data['essay'].values):
    ss = sid.polarity_scores(sentence)
    sentiment_neg.append(ss['neg'])
    sentiment_neu.append(ss['neu'])
    sentiment_pos.append(ss['pos'])
    sentiment_comp.append(ss['compound'])
```

```
100%|████████████████████████████████████████████████████████████████████
████| 50000/50000 [02:30<00:00, 331.58it/s]
```

In [126]:

```
#Number of words in Essay and Title
```

In [127]:

```
#Train Data

#number of words in the title
words_in_title_train = []
for i in x_train['project_title']:
    a = len(i.split())
    words_in_title_train.append(a)

#number of words in the essay
words_in_essay_train = []
for i in x_train['essay']:
    a = len(i.split())
    words_in_essay_train.append(a)


print(np.shape(words_in_title_train))
print(np.shape(words_in_essay_train))
```

```
(22445,)
(22445,)
```

In [128]:

```
#Test Data

#number of words in the title
words_in_title_test = []
for i in x_test['project_title']:
    a = len(i.split())
    words_in_title_test.append(a)

#number of words in the essay
words_in_essay_test = []
for i in x_test['essay']:
    a = len(i.split())
    words_in_essay_test.append(a)


print(np.shape(words_in_title_test))
print(np.shape(words_in_essay_test))
```

```
(16500,)
(16500,)
```

In [129]:

```
#CV data

#number of words in the title
words_in_title_cv = []
for i in x_cv['project_title']:
    a = len(i.split())
    words_in_title_cv.append(a)

#number of words in the essay
words_in_essay_cv = []
for i in x_cv['essay']:
    a = len(i.split())
    words_in_essay_cv.append(a)


print(np.shape(words_in_title_cv))
print(np.shape(words_in_essay_cv))
```

```
(11055,)
(11055,)
```

In [130]:

```
#Reshaping

#Train
words_in_essay_train = np.array(words_in_essay_train)
words_in_title_train = np.array(words_in_title_train)

words_in_essay_train = words_in_essay_train.reshape(-1, 1)
words_in_title_train = words_in_title_train.reshape(-1, 1)

print(words_in_essay_train.shape)
print(words_in_title_train.shape)
print('*'*20)


#Test
words_in_title_test = np.array(words_in_title_test)
words_in_essay_test = np.array(words_in_essay_test)

words_in_title_test = words_in_title_test.reshape(-1, 1)
words_in_essay_test = words_in_essay_test.reshape(-1, 1)

print(words_in_title_test.shape)
print(words_in_essay_test.shape)
print('*'*20)


#CV
words_in_title_cv = np.array(words_in_title_cv)
words_in_essay_cv = np.array(words_in_essay_cv)

words_in_title_cv = words_in_title_cv.reshape(-1, 1)
words_in_essay_cv = words_in_essay_cv.reshape(-1, 1)

print(words_in_title_cv.shape)
print(words_in_essay_cv.shape)
```

```
(22445, 1)
(22445, 1)
********************
(16500, 1)
(16500, 1)
********************
(11055, 1)
(11055, 1)
```

In [131]:

```
#Reshaping the lists we got in sentiment analysis and slicing lists into train, test, cv

#For Train
#Sentiment_neg
sentiment_neg_train = np.array(sentiment_neg)
sentiment_neg_train = sentiment_neg_train.reshape(-1, 1) [:22445]
print(sentiment_neg_train.shape)

#Sentiment_pos
sentiment_pos_train = np.array(sentiment_pos)
sentiment_pos_train = sentiment_pos_train.reshape(-1, 1) [:22445]
print(sentiment_pos_train.shape)

#Sentiment_neu
sentiment_neu_train = np.array(sentiment_neu)
sentiment_neu_train = sentiment_neu_train.reshape(-1, 1) [:22445]
print(sentiment_neu_train.shape)

#Sentiment_comp
sentiment_comp_train = np.array(sentiment_comp)
sentiment_comp_train = sentiment_comp_train.reshape(-1, 1) [:22445]
print(sentiment_comp_train.shape)
print('*'*50)

#For Test
#Sentiment_neg
sentiment_neg_test = np.array(sentiment_neg)
sentiment_neg_test = sentiment_neg_test.reshape(-1, 1) [22445:38945]
print(sentiment_neg_test.shape)

#Sentiment_pos
sentiment_pos_test = np.array(sentiment_pos)
sentiment_pos_test = sentiment_pos_test.reshape(-1, 1) [22445:38945]
print(sentiment_pos_test.shape)

#Sentiment_neu
sentiment_neu_test = np.array(sentiment_neu)
sentiment_neu_test = sentiment_neu_test.reshape(-1, 1) [22445:38945]
print(sentiment_neu_test.shape)

#Sentiment_comp
sentiment_comp_test = np.array(sentiment_comp)
sentiment_comp_test = sentiment_comp_test.reshape(-1, 1) [22445:38945]
print(sentiment_comp_test.shape)
print('*'*50)

#For CV
#Sentiment_neg
sentiment_neg_cv = np.array(sentiment_neg)
sentiment_neg_cv = sentiment_neg_cv.reshape(-1, 1) [38945:50000]
print(sentiment_neg_cv.shape)

#Sentiment_pos
sentiment_pos_cv = np.array(sentiment_pos)
sentiment_pos_cv = sentiment_pos_cv.reshape(-1, 1) [38945:50000]
print(sentiment_pos_cv.shape)

#Sentiment_neu
sentiment_neu_cv = np.array(sentiment_neu)
```

```
sentiment_neu_cv = sentiment_neu_cv.reshape(-1, 1) [38945:50000]
print(sentiment_neu_cv.shape)

#Sentiment_comp
sentiment_comp_cv = np.array(sentiment_comp)
sentiment_comp_cv = sentiment_comp_cv.reshape(-1, 1) [38945:50000]
print(sentiment_comp_cv.shape)
```

```
(22445, 1)
(22445, 1)
(22445, 1)
(22445, 1)
**************************************************
(16500, 1)
(16500, 1)
(16500, 1)
(16500, 1)
**************************************************
(11055, 1)
(11055, 1)
(11055, 1)
(11055, 1)
```

In [132]:

```
#Apply TruncatedSVD on TfidfVectorizer of essay text, choose the number of components (`n_c
```

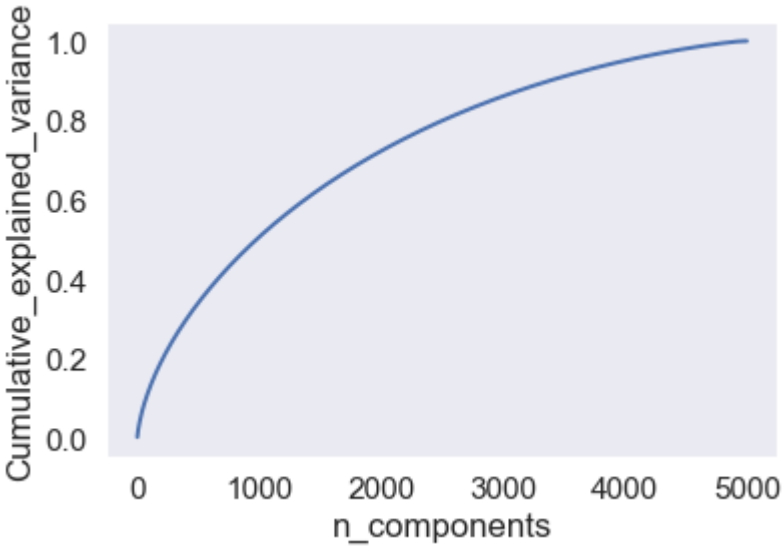
In [133]:

```
#Checking the shapes of  TfidfVectorizer Essay Train, test, cv
print(x_train_essay_tfidf.shape)
print(x_test_essay_tfidf.shape)
print(x_cv_essay_tfidf.shape)
```

```
(22445, 5000)
(16500, 5000)
(11055, 5000)
```

In [134]:

```
#Elbow Method to choose n_components for TruncatedSVD
from sklearn import decomposition
pca = decomposition.PCA()


#For Train
x_train_essay_tfidf = x_train_essay_tfidf.toarray()

pca.n_components = 5000
pca_data = pca.fit_transform(x_train_essay_tfidf)

percentage_var_explained = pca.explained_variance_ / np.sum(pca.explained_variance_);

cum_var_explained = np.cumsum(percentage_var_explained)

# Plot the PCA spectrum
plt.figure(1, figsize=(6, 4))

plt.clf()
plt.plot(cum_var_explained, linewidth=2)
plt.axis('tight')
plt.grid()
plt.xlabel('n_components')
plt.ylabel('Cumulative_explained_variance')
plt.show()
```



In [ ]:

```

```

In [135]:

```
#I want to preserve 70% of data. So I'm choosing n_components=1800

from sklearn.decomposition import TruncatedSVD
svd = TruncatedSVD(n_components=1800, n_iter=100)

train_essay_svd = svd.fit_transform(x_train_essay_tfidf)
print(train_essay_svd.shape)
```

```
(22445, 1800)
```

In [136]:

```
test_essay_svd = svd.transform(x_test_essay_tfidf)
cv_essay_svd = svd.transform(x_cv_essay_tfidf)
print(test_essay_svd.shape)
print(cv_essay_svd.shape)
```

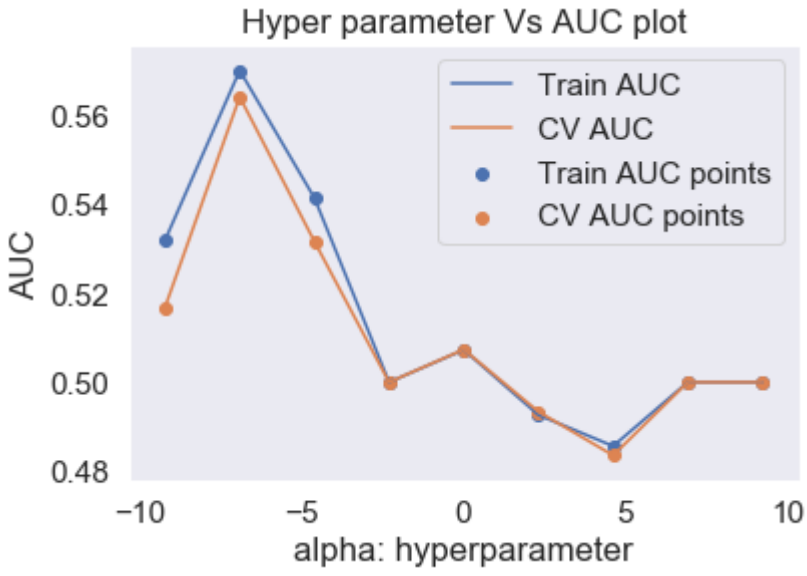
```
(16500, 1800)
(11055, 1800)
```

In [137]:

```
from scipy.sparse import hstack
x_train_set_5 = hstack((x_train_price_norm, x_train_previous_projects_norm, x_train_state_c
                        x_train_grade_one_hot, x_train_categories_one_hot, x_train_subcateg
                        words_in_essay_train, words_in_title_train, \
                       sentiment_neg_train, sentiment_pos_train, sentiment_neu_train, sentim

x_test_set_5 = hstack((x_test_price_norm, x_test_previous_projects_norm, x_test_state_one_h
                      x_test_grade_one_hot, x_test_categories_one_hot, x_test_subcategories_
                        words_in_essay_test, words_in_title_test,\
                       sentiment_neg_test, sentiment_pos_test, sentiment_neu_test, sentiment

x_cv_set_5 = hstack((x_cv_price_norm, x_cv_previous_projects_norm, x_cv_state_one_hot, x_cv
                      x_cv_grade_one_hot, x_cv_categories_one_hot, x_cv_subcategories_one_ho
                      words_in_essay_cv, words_in_title_cv,\
                     sentiment_neg_cv, sentiment_pos_cv, sentiment_neu_cv, sentiment_comp_cv
```

In [138]:

```
#Hyperparameter Tuning

tuned_parameters = {'alpha':[10**-4, 10**-3, 10**-2, 10**-1, 10**0, 10**1, 10**2, 10**3, 10
sgd = SGDClassifier(loss='hinge', class_weight='balanced')
clf = GridSearchCV(sgd, tuned_parameters, cv=10, refit=True, scoring='roc_auc')
clf.fit(x_train_set_5, y_train)

calibrated_clf = CalibratedClassifierCV(clf, cv='prefit')
calibrated_clf.fit(x_train_set_5, y_train)

results = pd.DataFrame.from_dict(calibrated_clf.base_estimator.cv_results_)
results = results.sort_values(['param_alpha'])

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']
alpha =  results['param_alpha']

for i in range(len(alpha)):
    alpha[i] = np.log(alpha[i])
plt.plot(alpha, 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,c

plt.plot(alpha, 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='darkor

plt.scatter(alpha, train_auc, label='Train AUC points')
plt.scatter(alpha, cv_auc, label='CV AUC points')


plt.legend()
plt.xlabel("alpha: hyperparameter")
plt.ylabel("AUC")
plt.title("Hyper parameter Vs AUC plot")
plt.grid()
plt.show()

results.head()
```



Out[138]:

	mean_fit_time	std_fit_time	mean_score_time	std_score_time	param_alpha	param_penalty
0	3.874661	0.459224	0.082872	0.141049	-9.21034	l1
1	2.736110	0.035132	0.020776	0.008024	-6.90776	l1
2	2.777774	0.115531	0.043571	0.034854	-4.60517	l1
3	3.675699	0.229914	0.021675	0.014871	-2.30259	l1
4	3.899208	0.189958	0.023768	0.020504	0	l1

5 rows × 32 columns

In [139]:

```
calibrated_clf.base_estimator.best_params_
```

Out[139]:

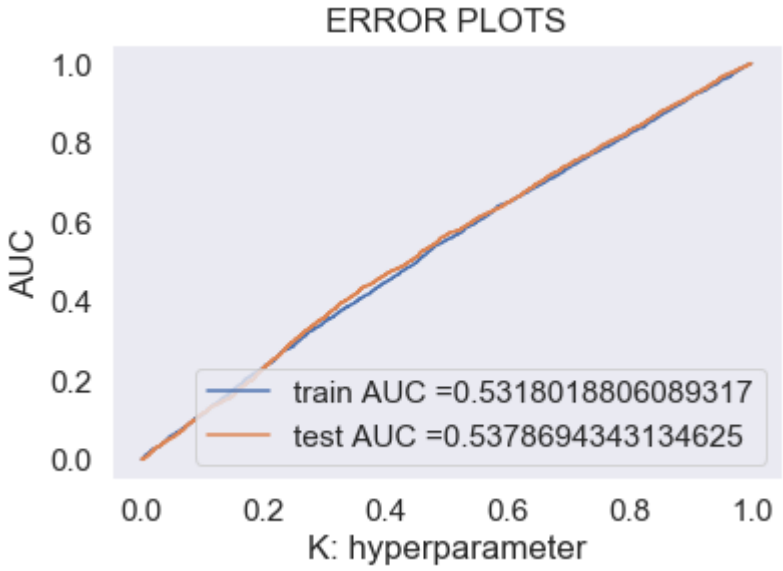
```
{'alpha': 0.001, 'penalty': 'l1'}
```

In [140]:

```
#Best-alpha
best_alpha = 0.001
```

In [141]:

```
#Hyperparameter Tuning with Best-alpha

from sklearn.metrics import roc_curve, auc

sgd = SGDClassifier(loss='hinge', penalty='l1', alpha=best_alpha, class_weight='balanced')
sgd.fit(x_train_set_5, y_train)
calibrated_clf = CalibratedClassifierCV(sgd, cv='prefit')
calibrated_clf.fit(x_train_set_5, y_train)

y_train_pred = calibrated_clf.predict_proba(x_train_set_5) [:, 1]
y_test_pred = calibrated_clf.predict_proba(x_test_set_5) [:, 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 [142]:

```
# 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.round(t
    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 [143]:

```
#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)))
```

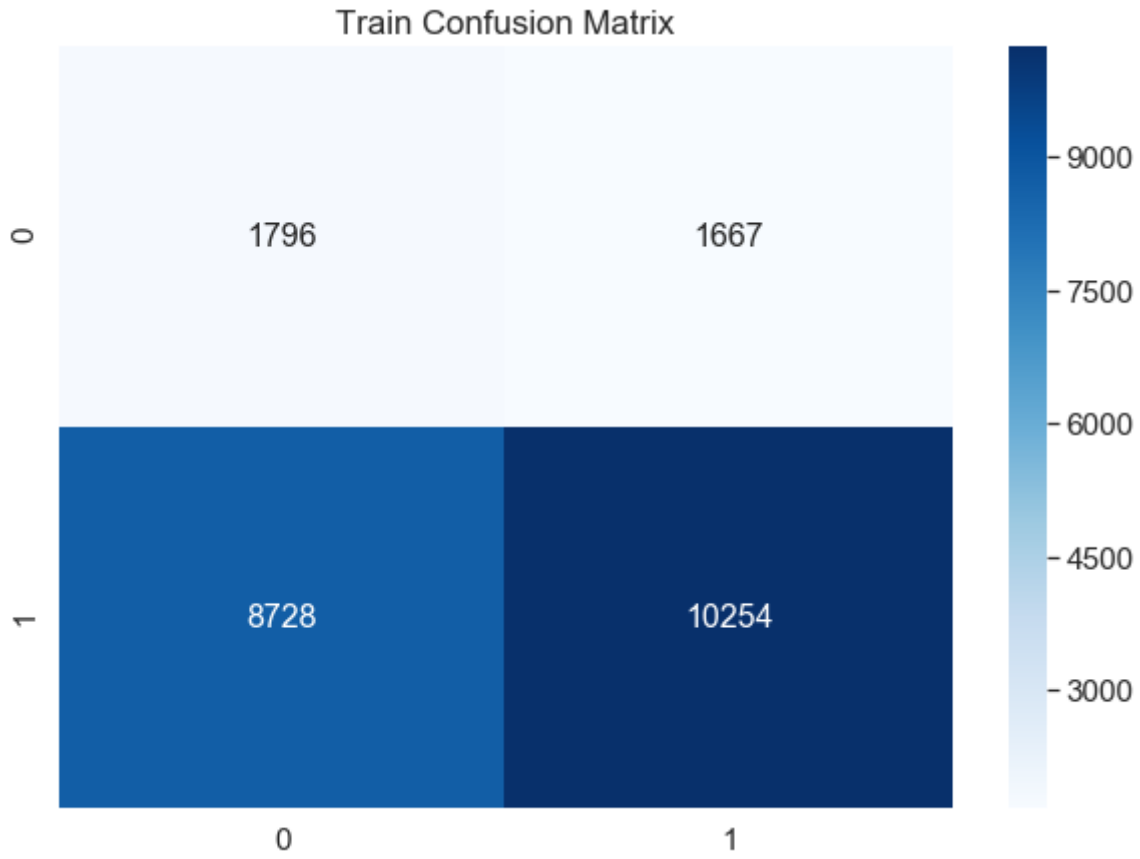
```
the maximum value of tpr*(1-fpr) 0.28015939108901833 for threshold 0.844
Train confusion matrix
[[ 1796  1667]
 [ 8728 10254]]
Test confusion matrix
[[1331 1215]
 [6406 7548]]
```

In [144]:

```
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)),

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[144]:

```
<matplotlib.axes._subplots.AxesSubplot at 0x184ba001e80>
```

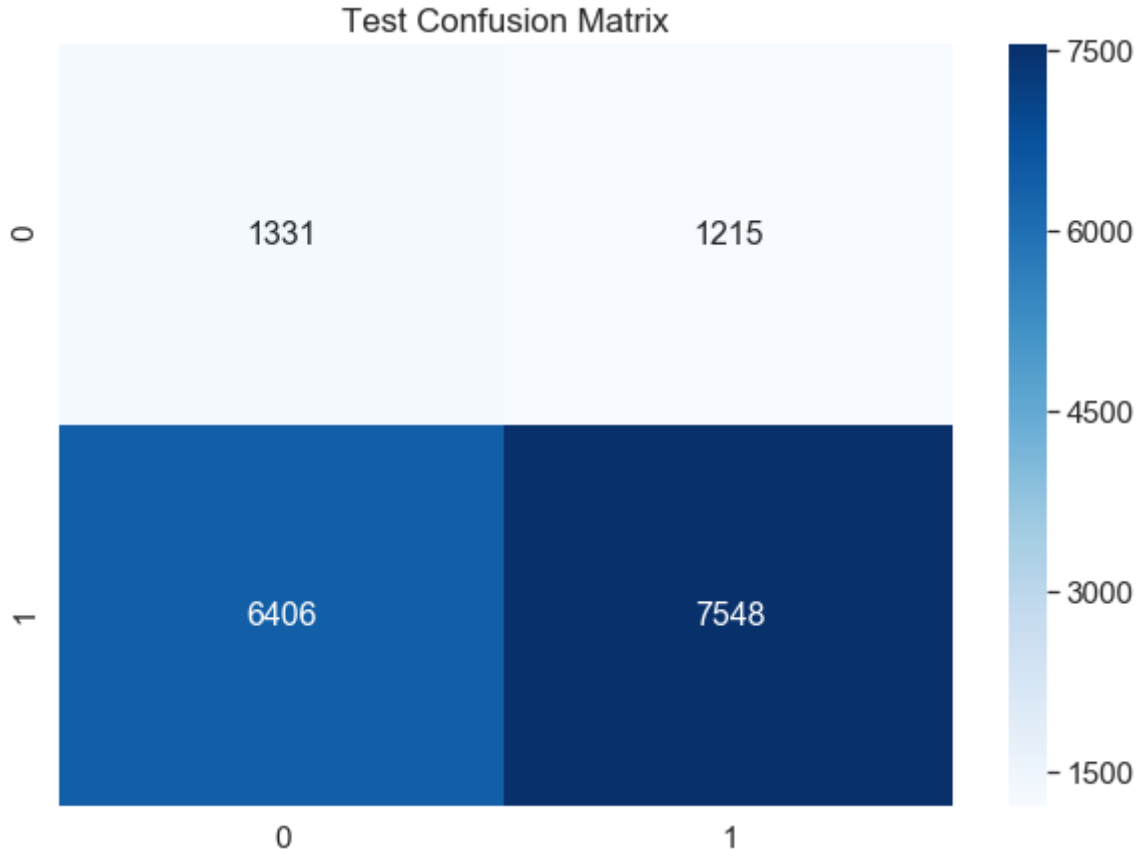


In [145]:

```
#Test Confusion matrix
df_cm = pd.DataFrame(confusion_matrix(y_test, predict_with_best_t(y_test_pred, best_t)), cc

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[145]:

```
<matplotlib.axes._subplots.AxesSubplot at 0x184bbe9f3c8>
```



3. Conclusion

In [143]:

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

In [146]:

```
from prettytable import PrettyTable

x = PrettyTable()

x. field_names = ['Vectorizer', 'Model', 'HyperParameter', 'AUC']

x.add_row(['Set-1', 'Brute', '0.0001', '0.59'])
x.add_row(['Set-2', 'Brute', '0.0001', '0.61'])
x.add_row(['Set-3', 'Brute', '0.001', '0.65'])
x.add_row(['Set-4', 'Brute', '0.001', '0.66'])
x.add_row(['Set-5', 'Brute', '0.001', '0.53'])

print(x)
```

```
+------------+-------+----------------+------+
| Vectorizer | Model | HyperParameter | AUC  |
+------------+-------+----------------+------+
|   Set-1    | Brute |     0.0001     | 0.59 |
|   Set-2    | Brute |     0.0001     | 0.61 |
|   Set-3    | Brute |     0.001      | 0.65 |
|   Set-4    | Brute |     0.001      | 0.66 |
|   Set-5    | Brute |     0.001      | 0.53 |
+------------+-------+----------------+------+
```

In [ ]: