Assignment 7: SVM

Data splitting and pre-processing

```
In [2]:
```

```
%matplotlib inline
import warnings
warnings.filterwarnings("ignore")
import sqlite3
import pandas as pd
import numpy as np
import nltk
import string
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.feature extraction.text import TfidfTransformer
from sklearn.feature extraction.text import TfidfVectorizer
from sklearn.model_selection import train test split
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
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
```

```
In [3]:
```

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

In [4]:

```
project_data.isnull().sum()
```

Out[4]:

```
Unnamed: 0
                                                        0
                                                        0
teacher id
                                                        0
teacher_prefix
                                                        3
school state
project_submitted_datetime
                                                        0
                                                        0
project_grade_category
project_subject_categories
                                                        0
project subject subcategories
                                                        0
project title
                                                        0
project essay 1
                                                        0
                                                        0
project essay 2
```

```
project_essay_3
                                                 105490
                                                 105490
{\tt project\_essay\_4}
project_resource_summary
                                                      0
teacher_number_of_previously_posted_projects
                                                      0
project is approved
dtype: int64
In [5]:
#filling 3 null teacher prefix values with Teacher
project_data["teacher_prefix"].fillna("Teacher",inplace = True)
project_data.isnull().sum()
Out[5]:
Unnamed: 0
id
                                                       0
                                                       0
teacher_id
teacher_prefix
school state
                                                      0
project submitted datetime
                                                       0
                                                       0
project grade category
                                                      0
project_subject_categories
                                                       0
project_subject_subcategories
project title
                                                       0
project essay 1
                                                      0
project_essay_2
project_essay_3
                                                 105490
                                                 105490
project_essay_4
project_resource_summary
                                                      0
teacher_number_of_previously_posted_projects
                                                      0
                                                       0
project_is_approved
dtype: int64
In [6]:
# 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 [7]:
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 [8]:
project_data.info()
<class 'pandas.core.frame.DataFrame'>
Int64Index: 109248 entries, 0 to 109247
Data columns (total 20 columns):
Unnamed: 0
                                                 109248 non-null int64
id
                                                 109248 non-null object
teacher id
                                                 109248 non-null object
teacher prefix
                                                 109248 non-null object
school state
                                                 109248 non-null object
                                                 109248 non-null object
project_submitted_datetime
project grade category
                                                 109248 non-null object
project subject categories
                                                 109248 non-null object
{\tt project\_subject\_subcategories}
                                                109248 non-null object
project title
                                                 109248 non-null object
project_essay_1
                                                 109248 non-null object
project_essay_2
                                                 109248 non-null object
project_essay_3
                                                 3758 non-null object
                                                 3758 non-null object
project_essay_4
                                                 109248 non-null object
project resource summary
teacher_number_of_previously_posted_projects
                                                 109248 non-null int64
                                                 109248 non-null int64
project_is_approved
```

```
109248 non-null object
essay
                                                 109248 non-null float64
price
quantity
                                                 109248 non-null int64
dtypes: float64(1), int64(4), object(15)
memory usage: 17.5+ MB
In [9]:
#splitting data as 30% to test
y = project_data["project_is_approved"]
X = project_data.drop("project_is_approved",axis = 1)
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.30, random_state=42)
In [10]:
print(X train.shape," ",y train.shape)
print(X_test.shape," ",y_test.shape)
(76473, 19)
              (76473,)
(32775, 19)
              (32775,)
```

Preprocessing categorical Features

1. project subject categories

In [11]:

```
#using code from assignment
# project subject categories
catogories = list(X_train['project_subject_categories'].values)
cat list = []
for i in catogories:
   temp = ""
    for j in i.split(','):
        if 'The' in j.split():
            j=j.replace('The','')
        j = j.replace(' ','')
        temp+=j.strip()+" "
        temp = temp.replace('&','_')
    cat_list.append(temp.strip())
X train['clean categories'] = cat list
X_train.drop(['project_subject_categories'], axis=1, inplace=True)
from collections import Counter
my counter = Counter()
for word in X train['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]))
# project subject categories for test data
catogories = list(X_test['project_subject_categories'].values)
cat list = []
for i in catogories:
    temp = ""
    for j in i.split(','):
       if 'The' in j.split():
           j=j.replace('The','')
        j = j.replace(' ','')
        temp+=j.strip()+" "
        temp = temp.replace('&','_')
    cat_list.append(temp.strip())
X_test['clean_categories'] = cat_list
X test.drop(['project subject categories']. axis=1. inplace=True)
```

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1. project subject sub categories

```
In [12]:
```

```
sub catogories = list(X train['project subject subcategories'].values)
sub cat list = []
for i in sub_catogories:
   temp = ""
    for j in i.split(','):
        if 'The' in j.split(): # this will split each of the catogory based on space "Math & Scienc
e"=> "Math","&", "Science"
            j=j.replace('The','') # if we have the words "The" we are going to replace it with ''(i
.e removing 'The')
       j = j.replace(' ','')
        temp +=j.strip()+" "
        temp = temp.replace('&',' ')
    sub_cat_list.append(temp.strip())
X_train['clean_subcategories'] = sub_cat_list
X_train.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 X train['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]))
sub catogories = list(X test['project subject subcategories'].values)
sub_cat_list = []
for i in sub_catogories:
    temp = ""
    for j in i.split(','):
       if 'The' in j.split(): # this will split each of the catogory based on space "Math & Scienc"
e"=> "Math", "&", "Science"
            j=j.replace('The','') # if we have the words "The" we are going to replace it with ''(i
.e removing 'The')
       j = j.replace(' ','')
        temp +=j.strip()+" "
        temp = temp.replace('&',' ')
    sub cat list.append(temp.strip())
X test['clean subcategories'] = sub cat list
X test.drop(['project subject subcategories'], axis=1, inplace=True)
4
                                                                                                 I
```

1. Teacher Prefix

In [13]:

```
#preprocessing teacher prefix
prefix = list(X_train['teacher_prefix'].values)
prefix list = []
for i in prefix:
    temp = ""
    if "." in i:
           i=i.replace('.','')
    temp+=i.strip()+" "
    prefix list.append(temp.strip())
X train['clean prefix'] = prefix list
my_counter = Counter()
for word in X train['clean prefix'].values:
 my_counter.update(word.split())
prefix dict = dict(my counter)
sorted_prefix_dict = dict(sorted(prefix_dict.items(), key=lambda kv: kv[1]))
print(sorted prefix dict)
```

```
#preprocessing teacher prefix for test data
prefix = list(X_test['teacher_prefix'].values)
prefix_list = []
for i in prefix:
    temp = ""
    if "." in i:
        i = i.replace('.','')
    temp+=i.strip()+" "
    prefix_list.append(temp.strip())

X_test['clean_prefix'] = prefix_list
```

{'Dr': 9, 'Teacher': 1647, 'Mr': 7457, 'Ms': 27320, 'Mrs': 40040}

1. Project Grade Category

In [14]:

```
# preprocessing of grade category for train data
grade = list(X_train['project_grade_category'].values)
grade_list = []
for i in grade:
   temp = ""
   if "Grades" in i:
     i = i.replace("Grades","")
   if "6-8" in i:
     i = i.replace("6-8", "six eight")
    if "3-5" in i:
     i = i.replace("3-5","three five")
    if "9-12" in i:
     i = i.replace("9-12", "nine twelve")
    if "PreK-2" in i:
     i = i.replace("PreK-2","prek_two")
    temp+=i.strip()+" "
   grade list.append(temp.strip())
X train['clean_grade'] = grade_list
my counter = Counter()
for word in X train['clean grade'].values:
 my counter.update(word.split())
grade_dict = dict(my_counter)
sorted grade dict = dict(sorted(grade dict.items(), key=lambda kv: kv[1]))
print(sorted grade dict)
# preprocessing of grade category for test data
grade = list(X test['project grade category'].values)
grade_list = []
for i in grade:
   temp = ""
   if "Grades" in i:
     i = i.replace("Grades","")
    if "6-8" in i:
     i = i.replace("6-8","six eight")
    if "3-5" in i:
     i = i.replace("3-5","three five")
    if "9-12" in i:
     i = i.replace("9-12", "nine twelve")
    if "PreK-2" in i:
     i = i.replace("PreK-2", "prek two")
    temp+=i.strip()+" "
    grade_list.append(temp.strip())
X_test['clean_grade'] = grade_list
```

{'nine_twelve': 7670, 'six_eight': 11843, 'three_five': 25958, 'prek_two': 31002}

```
In [15]:
```

```
#no need of preprocessing on school state

state = X_train["school_state"].value_counts()
sorted_state = dict(state)
sorted_state_dict = dict(sorted(sorted_state.items(), key=lambda kv: kv[1]))
X_train["clean_state"] = X_train["school_state"]

#similarly for X_test
X_test["clean_state"] = X_test["school_state"]
```

Preprocessing Numerical Feature

1. Standardizing price

In [16]:

```
from sklearn.preprocessing import StandardScaler

price_scalar = StandardScaler()
price_scalar.fit(project_data['price'].values.reshape(-1,1))
print(f"Mean : {price_scalar.mean_[0]}, Standard deviation : {np.sqrt(price_scalar.var_[0])}")

#train data price standardization
price_standardized = price_scalar.transform(X_train['price'].values.reshape(-1, 1))

#test data price standardization. Fit method applied on X_train
test_price_standardized = price_scalar.transform(X_test['price'].values.reshape(-1, 1))
```

Mean : 298.1193425966608, Standard deviation : 367.49634838483496

1. Standardizing quantity

In [17]:

```
price_scalar = StandardScaler()
price_scalar.fit(X_train["quantity"].values.reshape(-1, 1))
print(f"Mean of Quantity : {price_scalar.mean_[0]}, Standard deviation of Quantity :
{np.sqrt(price_scalar.var_[0])}")

#train data quantity standardization
quantity_standardized = price_scalar.transform(X_train["quantity"].values.reshape(-1, 1))

#test data quantity stanardization. Fit method applied on X_train
test_quantity_standardized = price_scalar.transform(X_test["quantity"].values.reshape(-1, 1))

C:\Users\rdbz3b\AppData\Local\Continuum\anaconda3\lib\site-
packages\sklearn\utils\validation.py:595: DataConversionWarning:

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

Mean of Quantity: 16.951629987054254, Standard deviation of Quantity: 25.894395919389655

```
C:\Users\rdbz3b\AppData\Local\Continuum\anaconda3\lib\site-
packages\sklearn\utils\validation.py:595: DataConversionWarning:

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

C:\Users\rdbz3b\AppData\Local\Continuum\anaconda3\lib\site-
packages\sklearn\utils\validation.py:595: DataConversionWarning:

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

1. Standardizing number of ppp

```
In [18]:
price scalar = StandardScaler()
price scalar.fit(X train['teacher number of previously posted projects'].values.reshape(-1,1))
print(f"Mean : {price_scalar.mean_[0]}, Standard deviation : {np.sqrt(price_scalar.var_[0])}")
#train data ppp standardization
number_ppp_standardized =
price scalar.transform(X train['teacher number of previously posted projects'].values.reshape(-1,
#test data price stanardization. Fit method applied on X train
test number ppp standardized =
price_scalar.transform(X_test['teacher_number_of_previously_posted_projects'].values.reshape(-1, 1)
C:\Users\rdbz3b\AppData\Local\Continuum\anaconda3\lib\site-
packages\sklearn\utils\validation.py:595: DataConversionWarning:
Data with input dtype int64 was converted to float64 by StandardScaler.
Mean: 11.09995684751481, Standard deviation: 27.56154092526415
C:\Users\rdbz3b\AppData\Local\Continuum\anaconda3\lib\site-
packages\sklearn\utils\validation.py:595: DataConversionWarning:
Data with input dtype int64 was converted to float64 by StandardScaler.
C:\Users\rdbz3b\AppData\Local\Continuum\anaconda3\lib\site-
packages\sklearn\utils\validation.py:595: DataConversionWarning:
Data with input dtype int64 was converted to float64 by StandardScaler.
```

Preprocessing of Text Feature for both test and train data

In [19]:

```
#using function and stopwords form assignemnt
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
# we are removing the words from the stop words list: 'no', 'nor', 'not'
stopwords= ['i', 'me', 'my', 'myself', 'we', 'our', 'ours', 'ourselves', 'you', "you're", "you've",
            "you'll", "you'd", 'your', 'yours', 'yourself', 'yourselves', 'he', 'him', 'his',
'himself', \
            'she', "she's", 'her', 'hers', 'herself', 'it', "it's", 'its', 'itself', 'they', 'them',
'their',\
            'theirs', 'themselves', 'what', 'which', 'whoo', 'whom', 'this', 'that', "that'll",
'these', 'those', \
```

```
'am', 'is', 'are', 'was', 'were', 'be', 'been', 'being', 'have', 'has', 'had', 'having',
'do', 'does', \
            'did', 'doing', 'a', 'an', 'the', 'and', 'but', 'if', 'or', 'because', 'as', 'until', '
while', 'of', \
            'at', 'by', 'for', 'with', 'about', 'against', 'between', 'into', 'through', 'during',
'before', 'after',\
            'above', 'below', 'to', 'from', 'up', 'down', 'in', 'out', 'on', 'off', 'over', 'under'
, 'again', 'further',\
            'then', 'once', 'here', 'there', 'when', 'where', 'why', 'how', 'all', 'any', 'both', '&
ach', 'few', 'more',\
            'most', 'other', 'some', 'such', 'only', 'own', 'same', 'so', 'than', 'too', 'very', \
            's', 't', 'can', 'will', 'just', 'don', "don't", 'should', "should've", 'now', 'd', 'll'
, 'm', 'o', 're', \
            've', 'y', 'ain', 'aren', "aren't", 'couldn', "couldn't", 'didn', "didn't", 'doesn', "do
esn't", 'hadn',\
            "hadn't", 'hasn', "hasn't", 'haven', "haven't", 'isn', "isn't", 'ma', 'mightn',
"mightn't", 'mustn',\
            "mustn't", 'needn', "needn't", 'shan', "shan't", 'shouldn', "shouldn't", 'wasn',
"wasn't", 'weren', "weren't", \
            'won', "won't", 'wouldn', "wouldn't"]
                                                                                                 •
```

1. preprocessing of project essay

In [20]:

```
from tqdm import tqdm
#for train data
preprocessed essays = []
# tqdm is for printing the status bar
for sentance in tqdm(X_train['essay'].values):
   sent = decontracted(sentance)
    sent = sent.replace('\\r', ' ')
   sent = sent.replace('\\"', ' ')
   sent = sent.replace('\\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 not in stopwords)
    preprocessed_essays.append(sent.lower().strip())
test preprocessed essays = []
# tqdm is for printing the status bar
for sentance in tqdm(X test['essay'].values):
    sent = decontracted(sentance)
    sent = sent.replace('\\r', ' ')
   sent = sent.replace('\\"', ' ')
   sent = sent.replace('\\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 not in stopwords)
    test preprocessed essays.append(sent.lower().strip())
100%|
                                                                            1 76473/76473
[00:45<00:00, 1678.67it/s]
100%|
[00:20<00:00, 1634.39it/s]
```

1. preprocessing of project title

In [21]:

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

```
# for test data
test preprocessed title = []
# tqdm is for printing the status bar
for sentance in tqdm(X_test['project_title'].values):
   sent = decontracted(sentance)
   sent = sent.replace('\\r', ' ')
   sent = sent.replace('\\"', ' ')
   sent = sent.replace('\\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 not in stopwords)
    test_preprocessed_title.append(sent.lower().strip())
100%|
[00:02<00:00, 36634.63it/s]
                                                                      32775/32775
100%|
[00:00<00:00, 35620.39it/s]
```

Vectorizing of Categorical data

1. Vectorizing project categories and subcategories

```
In [22]:
```

```
vectorizer = CountVectorizer(vocabulary=list(sorted cat dict.keys()), lowercase=False, binary=True
# fitting on train data
vectorizer.fit(X train['clean categories'].values)
print(vectorizer.get_feature_names())
categories feature = vectorizer.get feature names()
# for train data
categories one hot = vectorizer.transform(X train['clean categories'].values)
print("Shape of matrix after one hot encodig ", categories one hot.shape)
# for test data
test categories one hot = vectorizer.transform(X test['clean categories'].values)
```

['Warmth', 'Care_Hunger', 'History_Civics', 'Music_Arts', 'AppliedLearning', 'SpecialNeeds', 'Health_Sports', 'Math_Science', 'Literacy_Language'] Shape of matrix after one hot encodig (76473, 9)

1. Vectorizing project subcategories

```
In [23]:
vectorizer = CountVectorizer(vocabulary=list(sorted sub cat dict.keys()), lowercase=False, binary=
True)
# fitting on train data
vectorizer.fit(X_train['clean_subcategories'].values)
print(vectorizer.get feature names())
subcategories feature = vectorizer.get feature names()
# for train data
sub categories one hot = vectorizer.transform(X train['clean subcategories'].values)
print("Shape of matrix after one hot encodig ", sub categories one hot.shape)
# for test data
test_sub_categories_one_hot = vectorizer.transform(X_test['clean_subcategories'].values)
['Economics', 'CommunityService', 'FinancialLiteracy', 'ParentInvolvement', 'Extracurricular',
'Civics_Government', 'ForeignLanguages', 'NutritionEducation', 'Warmth', 'Care_Hunger',
'SocialSciences', 'PerformingArts', 'CharacterEducation', 'TeamSports', 'Other',
'College_CareerPrep', 'Music', 'History_Geography', 'Health_LifeScience', 'EarlyDevelopment', 'Gym
_Fitness', 'ESL', 'EnvironmentalScience', 'VisualArts', 'Health_Wellness', 'AppliedSciences',
```

```
'SpecialNeeds', 'Literature Writing', 'Mathematics', 'Literacy']
Shape of matrix after one hot encodig (76473, 30)
```

1. vectorizing teacher prefix

```
In [24]:
```

```
vectorizer = CountVectorizer(vocabulary=list(prefix_dict.keys()), lowercase=False, binary=True)
# fitting on train data
vectorizer.fit(X train['clean prefix'].values)
print(vectorizer.get feature names())
prefix feature = vectorizer.get feature names()
# for train data
prefix one hot = vectorizer.transform(X train['clean prefix'].values)
print("Shape of matrix after one hot encodig ",prefix one hot.shape)
# for test data
test prefix one hot = vectorizer.transform(X test['clean prefix'].values)
['Ms', 'Mrs', 'Mr', 'Teacher', 'Dr']
Shape of matrix after one hot encodig (76473, 5)
```

1. Vectorizing school state and grade

```
In [25]:
```

```
vectorizer = CountVectorizer(vocabulary=list(grade dict.keys()), lowercase=False, binary=True)
# fitting on train data
vectorizer.fit(X train['clean grade'].values)
print(vectorizer.get feature names())
grade feature = vectorizer.get feature names()
# for train data
grade_one_hot = vectorizer.transform(X_train['clean_grade'].values)
print("Shape of matrix after one hot encodig ",grade one hot.shape)
# for test data
test grade one hot = vectorizer.transform(X test['clean grade'].values)
vectorizer = CountVectorizer(vocabulary=list(sorted state dict.keys()), lowercase=False, binary=Tr
vectorizer.fit(X train['clean state'].values)
print(vectorizer.get feature names())
state one hot = vectorizer.transform(X train['clean state'].values)
state feature = vectorizer.get feature names()
test state one hot = vectorizer.transform(X test['clean state'].values)
['prek_two', 'nine_twelve', 'three_five', 'six_eight']
Shape of matrix after one hot encodig (76473, 4)
['VT', 'WY', 'ND', 'MT', 'RI', 'NE', 'SD', 'DE', 'AK', 'NH', 'WV', 'ME', 'HI', 'DC', 'NM', 'KS', 'I
A', 'ID', 'AR', 'CO', 'MN', 'KY', 'OR', 'MS', 'NV', 'MD', 'CT', 'TN', 'UT', 'AL', 'WI', 'VA', 'AZ',
'OK', 'NJ', 'WA', 'MA', 'LA', 'OH', 'MO', 'IN', 'PA', 'MI', 'SC', 'GA', 'IL', 'NC', 'FL', 'NY', 'TX
', 'CA']
```

Vectorizing Text Feature

1. BOW

```
In [26]:
```

```
vectorizer = CountVectorizer(min_df=10,ngram_range=(2,2),max_features=5000)
#fit using train data
vectorizer.fit (preprocessed essays)
essay_feature = vectorizer.get_feature_names()
```

```
# for train data
text_bow = vectorizer.transform(preprocessed_essays)
print("Shape of train matrix : ",text_bow.shape)
# for test data
test_text_bow = vectorizer.transform(test_preprocessed_essays)
print("Shape of test matrix : ",test_text_bow.shape)

# for title
vectorizer.fit(preprocessed_title)
title_feature = vectorizer.get_feature_names()

# for train data
title_bow = vectorizer.transform(preprocessed_title)
print("Shape of train matrix : ",title_bow.shape)
# for test data
test_title_bow = vectorizer.transform(test_preprocessed_title)
print("Shape of test matrix : ",test_title_bow.shape)
```

Shape of train matrix : (76473, 5000) Shape of test matrix : (32775, 5000) Shape of train matrix : (76473, 2843) Shape of test matrix : (32775, 2843)

1. TFIDF

In [27]:

```
vectorizer = TfidfVectorizer(min df=10,ngram range=(2,2),max features=5000)
#fit using train data
vectorizer.fit (preprocessed essays)
essay feature tfidf = vectorizer.get feature names()
# for train data
text tfidf = vectorizer.transform(preprocessed essays)
print("Shape of train matrix : ",text tfidf.shape)
# for test data
test text tfidf = vectorizer.transform(test preprocessed essays)
print("Shape of test matrix : ",test_text_tfidf.shape)
# for title
vectorizer.fit(preprocessed_title)
title feature tfidf = vectorizer.get feature names()
# for train data
title tfidf = vectorizer.transform(preprocessed title)
print("Shape of train matrix : ",title_tfidf.shape)
# for test data
test title tfidf = vectorizer.transform(test preprocessed title)
print("Shape of test matrix : ",test title tfidf.shape)
```

Shape of train matrix : (76473, 5000) Shape of test matrix : (32775, 5000) Shape of train matrix : (76473, 2843) Shape of test matrix : (32775, 2843)

1. Avg W2v

In [28]:

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

In [29]:

```
# for train data
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]))
# for test data
test avg w2v vectors = [] # the avg-w2v for each sentence/review is stored in this list
for sentence in tqdm(test_preprocessed_essays): # for each review/sentence
   vector = np.zeros(300) # as word vectors are of zero length
    cnt words =0; # num of words with a valid vector in the sentence/review
    for word in sentence.split(): # for each word in a review/sentence
        if word in glove words:
            vector += model[word]
           cnt words += 1
    if cnt words != 0:
       vector /= cnt_words
    test avg w2v vectors.append(vector)
print(len(test_avg_w2v_vectors))
print(len(test avg w2v vectors[0]))
                                                                              | 76473/76473
100%1
[00:25<00:00, 3032.42it/s]
76473
300
100%|
                                                                              32775/32775
[00:11<00:00, 2824.59it/s]
```

32775 300

In [30]:

```
title avg w2v vectors = []
for sentence in tqdm(preprocessed title):
   vector = np.zeros(300)
   cnt words =0;
    for word in sentence.split():
        if word in glove_words:
            vector += model[word]
            cnt words += 1
    if cnt words != 0:
        vector /= cnt words
    title_avg_w2v_vectors.append(vector)
print(len(title avg w2v vectors))
print(len(title_avg_w2v_vectors[0]))
# for test data
test_title_avg_w2v_vectors = []
for sentence in tqdm(test preprocessed title):
   vector = np.zeros(300)
    cnt words =0;
    for word in sentence.split():
        if word in glove_words:
            vector += model[word]
            cnt words += 1
    if cnt_words != 0:
       vector /= cnt words
    test_title_avg_w2v_vectors.append(vector)
print(len(test title avg w2v vectors))
print(len(test title avg w2v vectors[0]))
100%1
                                                                             1 76473/76473
```

```
[00:01<00:00, 56320.76it/s]
```

```
100%| 32775/32775
[00:00<00:00, 61539.59it/s]
```

1. TFIDF avgw2v

In [31]:

300

300

```
test tfidf model = TfidfVectorizer()
test_tfidf_model.fit(preprocessed_essays)
# we are converting a dictionary with word as a key, and the idf as a value
dictionary = dict(zip(test tfidf model.get feature names(), list(test tfidf model.idf )))
tfidf words = set(test tfidf model.get feature names())
test tfidf w2v vectors = []; # the avg-w2v for each sentence/review is stored in this list
for sentence in tqdm(test_preprocessed_essays): # for each review/sentence
    vector = np.zeros(300) # as word vectors are of zero length
    tf idf weight =0; # num of words with a valid vector in the sentence/review
    for word in sentence.split(): # for each word in a review/sentence
        if (word in glove_words) and (word in tfidf_words):
            vec = model[word] # getting the vector for each word
            # here we are multiplying idf value(dictionary[word]) and the tf
value((sentence.count(word)/len(sentence.split())))
            tf idf = dictionary[word]*(sentence.count(word)/len(sentence.split())) # getting the tf
idf value for each word
            vector += (vec * tf idf) # calculating tfidf weighted w2v
            tf idf weight += tf idf
    if tf idf weight != 0:
       vector /= tf idf weight
    test tfidf w2v vectors.append(vector)
print(len(test tfidf w2v vectors))
print(len(test tfidf w2v vectors[0]))
# for title
test tfidf model.fit(preprocessed title)
dictionary = dict(zip(test tfidf model.get feature names(), list(test tfidf model.idf )))
tfidf_words = set(test_tfidf_model.get_feature_names())
\texttt{test\_title\_tfidf\_w2v\_vectors} = \texttt{[]; \# the avg-w2v for each sentence/review is stored in this list}
for sentence in tqdm (test preprocessed 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((sentence.count(word)/len(sentence.split())))
           tf idf = dictionary[word]*(sentence.count(word)/len(sentence.split())) # getting the tf
idf value for each word
            vector += (vec * tf idf) # calculating tfidf weighted w2v
            tf idf weight += tf idf
    if tf idf weight != 0:
       vector /= tf idf weight
    test title tfidf w2v vectors.append(vector)
print(len(test title tfidf w2v vectors))
                                                                                32775/32775 [01:
10<00:00, 465.12it/s]
```

```
100%| 32775/32775 [00:01<00:00, 32773.61it/s]
```

32775

In [32]:

```
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())
tfidf w2v vectors = []; # the avg-w2v for each sentence/review is stored in this list
for sentence in tqdm(preprocessed essays): # for each review/sentence
    vector = np.zeros(300) # as word vectors are of zero length
    tf idf weight =0; # num of words with a valid vector in the sentence/review
    for word in sentence.split(): # for each word in a review/sentence
        if (word in glove words) and (word in tfidf words):
            vec = model[word] # getting the vector for each word
            # here we are multiplying idf value(dictionary[word]) and the tf
value((sentence.count(word)/len(sentence.split())))
            tf idf = dictionary[word]*(sentence.count(word)/len(sentence.split())) # getting the tf
idf value for each word
            vector += (vec * tf idf) # calculating tfidf weighted w2v
            tf idf weight += tf idf
    if tf idf weight != 0:
       vector /= tf idf weight
    tfidf_w2v_vectors.append(vector)
print(len(tfidf w2v vectors))
print(len(tfidf w2v vectors[0]))
# for title
tfidf model.fit(preprocessed title)
dictionary = dict(zip(tfidf_model.get_feature_names(), list(tfidf_model.idf_)))
tfidf words = set(tfidf model.get feature names())
title tfidf w2v vectors = []; # the avg-w2v for each sentence/review is stored in this list
for sentence in tqdm(preprocessed 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((sentence.count(word)/len(sentence.split())))
           tf idf = dictionary[word]*(sentence.count(word)/len(sentence.split())) # getting the tf
idf value for each word
            vector += (vec * tf idf) # calculating tfidf weighted w2v
            tf idf weight += tf idf
    if tf idf weight != 0:
       vector /= tf idf_weight
    title tfidf w2v vectors.append(vector)
print(len(title tfidf w2v vectors))
                                                                      76473/76473 [02:
100%|
48<00:00, 452.88it/s]
```

76473 300

```
100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%|
```

Printing all

```
In [33]:
```

```
print("*"*70)
print("Categorical Features that are considered :- ")
print("Subject Categories :- ", categories one hot.shape)
print("Subject Sub-Categories :- ", sub categories one hot.shape)
print("Sudent Grade :- ", grade one hot.shape)
print("School State :- ", state one hot.shape)
print("Teacher Prefix :- ",prefix one hot.shape)
print("*"*70)
******************
Categorical Features that are considered :-
Subject Categories :- (76473, 9)
Subject Sub-Categories :- (76473, 30)
Sudent Grade :- (76473, 4)
School State :- (76473, 51)
Teacher Prefix :- (76473, 5)
In [34]:
print("Text Features that are considered :- ")
print("Project Essay BOW:- ",text bow.shape)
print("Project Essay TFIDF:- ",text tfidf.shape)
print("*"*70)
print("Project Title BOW:- ",title_bow.shape)
print("Project Title TFIDF:- ", title tfidf.shape)
print("*"*70)
Text Features that are considered :-
******************
Project Essay BOW:- (76473, 5000)
Project Essay TFIDF:- (76473, 5000)
Project Title BOW:- (76473, 2843)
Project Title TFIDF:- (76473, 2843)
```

sets

In [35]:

```
#combining all feature into one
from scipy.sparse import hstack
hstack((categories_one_hot,sub_categories_one_hot,prefix_one_hot,grade_one_hot,state_one_hot,price_
standardized, quantity standardized, number ppp standardized, text bow, title bow))
hstack((test_categories_one_hot,test_sub_categories_one_hot,test_prefix_one_hot,test_grade_one_hot
, test state one hot, test price standardized, test quantity standardized, test number ppp standardized
,test_text_bow,test_title_bow))
hstack((categories_one_hot,sub_categories_one_hot,prefix_one_hot,state_one_hot,grade_one_hot,text_t
fidf,title tfidf,price standardized,quantity standardized,number ppp standardized))
set2 t =
hstack((test_categories_one_hot,test_sub_categories_one_hot,test_prefix_one_hot,test_state_one_hot
,test grade one hot,test text tfidf,test title tfidf,test price standardized,test quantity standard
ized,test_number_ppp_standardized))
hstack((categories_one_hot,sub_categories_one_hot,prefix_one_hot,state_one_hot,grade_one_hot,price_
standardized, \verb"quantity_standardized", \verb"number_ppp_standardized", \verb"avg_w2v_vectors", \verb"title_avg_w2v_vectors")) \\
set3 t =
```

```
nstack((test categories one hot,test sub categories one hot,test prefix one hot,test state one hot
,test_grade_one_hot,test_price_standardized,test_quantity_standardized,test_number_ppp_standardized
,test avg w2v vectors,test title avg w2v vectors))
hstack((categories one hot, sub categories one hot, prefix one hot, state one hot, grade one hot, price
standardized, quantity\_standardized, number\_ppp\_standardized, tfidf\_w2v\_vectors, title\_tfidf\_w2v\_vectors, title\_tfidf\_w
set4 t =
hstack((test_categories_one_hot,test_sub_categories_one_hot,test_prefix_one_hot,test_state_one_hot
,test grade one hot,test price standardized,test quantity standardized,test number ppp standardized
,test tfidf w2v vectors,test title tfidf w2v vectors))
print(set1.shape, "\t", set1 t.shape)
print(set2.shape,"\t",set2 t.shape)
print(set3.shape,"\t",set3 t.shape)
print(set4.shape,"\t",set4 t.shape)
4
(76473, 7945)
(76473, 7945)
                                (32775, 7945)
                                 (32775, 7945)
(76473, 702)
                              (32775, 702)
(76473, 702)
                            (32775, 702)
In [36]:
set feature = categories feature + subcategories feature + prefix feature + grade feature + state f
eature + essay_feature_tfidf + title feature tfidf
set feature.append("price")
set_feature.append("quantity")
set_feature.append("number")
print(len(set_feature))
7945
In [37]:
par_grid = dict(penalty = ["11","12"],alpha=[0.00001,0.0001,0.001,0.1,1,10,100,1000,10000])
alpha=[0.00001,0.0001,0.001,0.1,1,10,100,1000,10000]
In [38]:
from sklearn.calibration import CalibratedClassifierCV
from sklearn.model selection import GridSearchCV
from sklearn.metrics import confusion matrix
from sklearn.metrics import roc curve
from sklearn.linear_model import SGDClassifier
SET1 (BOW)
In [39]:
sgd = SGDClassifier(class weight="balanced")
 #using balanced class weight = "balanced" as result using it was much better than "none"
grid = GridSearchCV(sgd,par_grid,scoring="roc_auc",n_jobs=-1,cv=10)
In [40]:
grid.fit(set1,y train)
print(grid.best_estimator_)
print(grid.best_index_)
print(grid.best params )
print(grid.best score )
```

SGDClassifier(alpha=0.001, average=False, class weight='balanced',

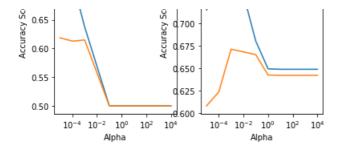
```
early stopping=False, epsilon=0.1, eta0=0.0, fit intercept=True,
       11_ratio=0.15, learning_rate='optimal', loss='hinge', max_iter=None,
       n_iter=None, n_iter_no_change=5, n_jobs=None, penalty='12',
       power t=0.5, random state=None, shuffle=True, tol=None,
       validation_fraction=0.1, verbose=0, warm start=False)
{'alpha': 0.001, 'penalty': '12'}
0.6713670013694405
In [41]:
sgd = SGDClassifier(alpha=0.001,class_weight="balanced")
ccv = CalibratedClassifierCV(sgd,cv=10)
ccv.fit(set1,y_train)
print(ccv.get_params)
<bound method BaseEstimator.get params of</pre>
CalibratedClassifierCV(base estimator=SGDClassifier(alpha=0.001, average=False,
class weight='balanced',
       early stopping=False, epsilon=0.1, eta0=0.0, fit intercept=True,
       11_ratio=0.15, learning_rate='optimal', loss='hinge', max_iter=None,
       n_iter=None, n_iter_no_change=5, n_jobs=None, penalty='12',
       power t=0.5, random state=None, shuffle=True, tol=None,
       validation_fraction=0.1, verbose=0, warm_start=False),
            cv=10, method='sigmoid')>
In [42]:
#converting results to dataframe
df = pd.DataFrame(data = grid.cv results )
# getting into list
11 train score = []
11 test score = []
12_train_score = []
12_test_score = []
for i in range(len(df)):
  if df.iloc[i]["param_penalty"] =="11":
    11_test_score.append(df.iloc[i]["mean_test_score"])
    11 train score.append(df.iloc[i]["mean train score"])
  if df.iloc[i]["param penalty"] =="12":
    12 test score.append(df.iloc[i]["mean test score"])
    12 train score.append(df.iloc[i]["mean train score"])
In [43]:
plt.figure()
plt.subplot(121)
plt.plot(alpha, 11_train_score)
plt.plot(alpha, 11 test score)
plt.xscale("log")
plt.xlabel("Alpha")
plt.ylabel("Accuracy Score")
plt.title("L1 train/test AUC plot")
plt.subplot(122)
plt.plot(alpha, 12 train score)
plt.plot(alpha, 12 test score)
plt.xscale("log")
plt.xlabel("Alpha")
plt.ylabel("Accuracy Score")
plt.title("L2 train/test AUC plot")
plt.show()
       L1 train/test AUC plot
                              L2 train/test AUC plot
                        d 775
```

0.75

0.70

d 750

일 d 725



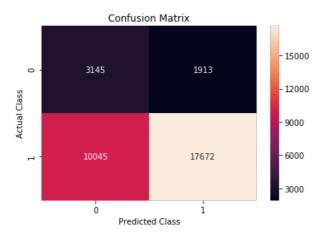
Confusion Matrix

In [44]:

```
y1_predict = grid.predict(set1_t)
cm1 = confusion_matrix(y_test,y1_predict)
# https://seaborn.pydata.org/generated/seaborn.heatmap.html
sns.heatmap(cm1, annot=True, fmt="d")
plt.ylabel("Actual Class")
plt.xlabel("Predicted Class")
plt.title("Confusion Matrix")
```

Out[44]:

Text(0.5, 1.0, 'Confusion Matrix')



AUC plotting

In [45]:

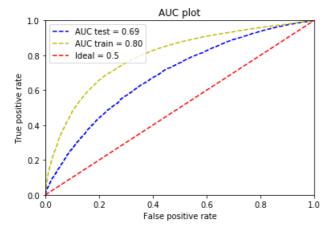
```
# probabilities calcultion
y1_predict_prob = ccv.predict_proba(set1_t)[:,1]
y1_predict_prob_train = ccv.predict_proba(set1)[:,1]

# took referance from https://stackoverflow.com/questions/25009284/how-to-plot-roc-curve-in-python
#fpr,tpr
fpr,tpr
fpr,tpr,thre = roc_curve(y_test,y1_predict_prob)

# am i doing it right here.....?
fpr_train,tpr_train,thre_train = roc_curve(y_train,y1_predict_prob_train)
```

In [46]:

```
plt.xlabel("False positive rate")
plt.ylabel("True positive rate")
plt.plot([0, 1], [0, 1], 'r--', label = "Ideal = 0.5")
plt.xlim([0,1])
plt.ylim([0,1])
plt.legend(loc = "upper left")
plt.show()
```



Set2 (TFIDF)

```
In [47]:
```

```
grid.fit(set2,y train)
print(grid.best_estimator_)
print(grid.best_index_)
print(grid.best params )
print(grid.best score )
SGDClassifier(alpha=0.0001, average=False, class weight='balanced',
       early stopping=False, epsilon=0.1, eta0=0.0, fit intercept=True,
       11_ratio=0.15, learning_rate='optimal', loss='hinge', max_iter=None,
       n iter=None, n iter no change=5, n jobs=None, penalty='11',
       power t=0.5, random state=None, shuffle=True, tol=None,
       validation fraction=0.1, verbose=0, warm start=False)
{'alpha': 0.0001, 'penalty': 'l1'}
0.6643859989303158
In [48]:
sad =
SGDClassifier(alpha=grid.best params ["alpha"], class weight="balanced", penalty=grid.best params ["
penalty"])
ccv = CalibratedClassifierCV(sqd,cv=10)
ccv.fit(set2,y train)
print(ccv.get_params)
<bound method BaseEstimator.get params of</pre>
CalibratedClassifierCV(base_estimator=SGDClassifier(alpha=0.0001, average=False,
class_weight='balanced',
       early stopping=False, epsilon=0.1, eta0=0.0, fit intercept=True,
       11_ratio=0.15, learning_rate='optimal', loss='hinge', max_iter=None,
       n_iter=None, n_iter_no_change=5, n_jobs=None, penalty='11',
       power t=0.5, random state=None, shuffle=True, tol=None,
       validation_fraction=0.1, verbose=0, warm_start=False),
            cv=10, method='sigmoid')>
```

In [49]:

```
#converting results to dataframe
df = pd.DataFrame(data = grid.cv results )
# getting into list
```

```
11_test_score = []
12_train_score = []
12_train_score = []
12_test_score = []

for i in range(len(df)):
    if df.iloc[i]["param_penalty"] =="ll":
        l1_test_score.append(df.iloc[i]["mean_test_score"])
        l1_train_score.append(df.iloc[i]["mean_train_score"])

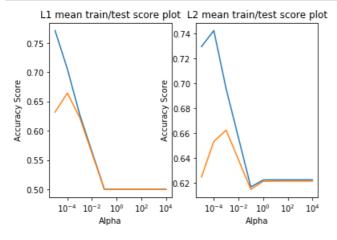
if df.iloc[i]["param_penalty"] =="l2":
        l2_test_score.append(df.iloc[i]["mean_test_score"])
        l2_train_score.append(df.iloc[i]["mean_train_score"])

print(l1_train_score)
print(l1_test_score)
print(l2_train_score)
print(l2_test_score)
```

```
[0.7711956820598286, 0.7056481370322893, 0.6278956430004208, 0.5, 0.5, 0.5, 0.5, 0.5, 0.5] [0.6319330965945366, 0.6643859989303158, 0.6226427908628587, 0.5, 0.5, 0.5, 0.5, 0.5, 0.5] [0.729627638378256, 0.7424725006884214, 0.6956403582683837, 0.616897971263666, 0.6225206342773695, 0.6226272610502457, 0.6226273492253667, 0.6226273492253667, 0.6226273492253667] [0.6250498058608464, 0.6534576279197091, 0.6625108823454687, 0.6150018560995887, 0.6213947934698029, 0.6215429509403063, 0.6215430714363522, 0.6215430714363522]
```

In [50]:

```
plt.figure()
plt.subplot(121)
plt.plot(alpha, 11 train score)
plt.plot(alpha, 11_test_score)
plt.xscale("log")
plt.xlabel("Alpha")
plt.ylabel("Accuracy Score")
plt.title("L1 mean train/test score plot")
plt.subplot(122)
plt.plot(alpha, 12 train score)
plt.plot(alpha,12 test score)
plt.xscale("log")
plt.xlabel("Alpha")
plt.ylabel("Accuracy Score")
plt.title("L2 mean train/test score plot")
plt.show()
```



Confusion matrix

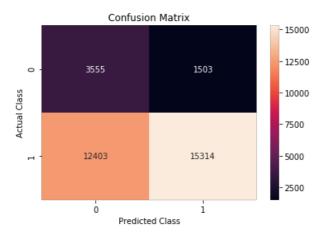
In [51]:

```
y2_predict = grid.predict(set2_t)
cm2 = confusion_matrix(y_test,y2_predict)
# https://seaborn.pydata.org/generated/seaborn.heatmap.html
sns.heatmap(cm2, annot=True, fmt="d")
plt.ylabel("Actual Class")
plt.xlabel("Predicted Class")
```

plt.title("Confusion Matrix")

Out[51]:

Text(0.5, 1.0, 'Confusion Matrix')



AUC

In [52]:

```
# probabilities calcultion
y2_predict_prob = ccv.predict_proba(set2_t)[:,1]
y2_predict_prob_train = ccv.predict_proba(set2)[:,1]

# took referance from https://stackoverflow.com/questions/25009284/how-to-plot-roc-curve-in-python
#fpr,tpr
fpr,tpr,thre = roc_curve(y_test,y2_predict_prob)

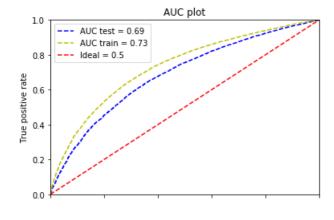
# am i doing it right here.....?
fpr_train,tpr_train,thre_train = roc_curve(y_train,y2_predict_prob_train)
```

In [53]:

```
# auc calculation for test data
roc_auc2 = metrics.auc(fpr,tpr)

# auc calculation for train data
roc_auc_train2 = metrics.auc(fpr_train,tpr_train)

# took referance from https://stackoverflow.com/questions/25009284/how-to-plot-roc-curve-in-python
plt.plot(fpr,tpr,"b--",label = 'AUC test = %0.2f'%roc_auc2)
plt.plot(fpr_train,tpr_train,"y--",label = 'AUC train = %0.2f'%roc_auc_train2)
plt.title("AUC plot")
plt.xlabel("Talse positive rate")
plt.ylabel("True positive rate")
plt.plot([0, 1], [0, 1],'r--',label = "Ideal = 0.5")
plt.xlim([0,1])
plt.ylim([0,1])
plt.legend(loc = "upper left")
plt.show()
```



Set3 (Avg W2V)

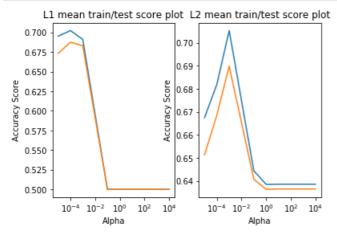
```
In [54]:
grid = GridSearchCV(sqd,par grid,scoring="roc auc",n jobs=-1,cv=10)
grid.fit(set3,y train)
print(grid.best estimator )
print(grid.best index )
print(grid.best_params_)
print(grid.best score )
SGDClassifier(alpha=0.001, average=False, class weight='balanced',
       early stopping=False, epsilon=0.1, eta0=0.0, fit intercept=True,
       11 ratio=0.15, learning rate='optimal', loss='hinge', max iter=None,
       n iter=None, n iter no change=5, n jobs=None, penalty='12',
       power_t=0.5, random_state=None, shuffle=True, tol=None,
       validation fraction=0.1, verbose=0, warm start=False)
5
{'alpha': 0.001, 'penalty': '12'}
0.6898722244176072
In [55]:
SGDClassifier(alpha=grid.best params ["alpha"], class weight="balanced", penalty=grid.best params ["
penalty"])
ccv = CalibratedClassifierCV(sgd,cv=10)
ccv.fit(set3,y train)
print(ccv.get params)
<bound method BaseEstimator.get params of</pre>
CalibratedClassifierCV(base estimator=SGDClassifier(alpha=0.001, average=False,
class weight='balanced',
       early stopping=False, epsilon=0.1, eta0=0.0, fit intercept=True,
       11_ratio=0.15, learning_rate='optimal', loss='hinge', max_iter=None,
       n iter=None, n iter no change=5, n jobs=None, penalty='12',
       power t=0.5, random state=None, shuffle=True, tol=None,
       validation fraction=0.1, verbose=0, warm start=False),
            cv=10, method='sigmoid')>
In [56]:
#converting results to dataframe
df = pd.DataFrame(data = grid.cv_results_)
# getting into list
11 train score = []
11_test_score = []
  train score = []
12_test_score = []
for i in range(len(df)):
  if df.iloc[i]["param penalty"] =="11":
    11 test score.append(df.iloc[i]["mean test score"])
    11_train_score.append(df.iloc[i]["mean_train_score"])
  if df.iloc[i]["param penalty"] =="12":
    12 test score.append(df.iloc[i]["mean test score"])
    12_train_score.append(df.iloc[i]["mean_train_score"])
print(11 train score)
print(l1 test score )
print(12 train score)
print(12 test score)
[0.6952675553506407,\ 0.7023841517798984,\ 0.6910450895707014,\ 0.5,\ 0.5,\ 0.5,\ 0.5,\ 0.5,\ 0.5]
[0.6731981961850698, 0.6875299992462821, 0.6830276960516257, 0.5, 0.5, 0.5, 0.5, 0.5, 0.5]
```

 $\begin{smallmatrix} 10.6674706271249987. & 0.6819791106428462. & 0.7053584602089862. & 0.6443475210746894. \end{smallmatrix}$

```
0.6384886081060215, 0.6385540012326143, 0.63855407921698, 0.6385534867118401, 0.638553579156604] [0.6513451514851318, 0.6685588686395242, 0.6898722244176072, 0.6406571319779433, 0.6363145752346241, 0.6364610563592599, 0.6364600639967104, 0.6364604524898966, 0.6364599565536718]
```

In [57]:

```
plt.figure()
plt.subplot(121)
plt.plot(alpha, 11 train score)
plt.plot(alpha, l1_test_score)
plt.xscale("log")
plt.xlabel("Alpha")
plt.ylabel("Accuracy Score")
plt.title("L1 mean train/test score plot")
plt.subplot(122)
plt.plot(alpha,12 train score)
plt.plot(alpha, 12_test_score)
plt.xscale("log")
plt.xlabel("Alpha")
plt.ylabel("Accuracy Score")
plt.title("L2 mean train/test score plot")
plt.show()
```



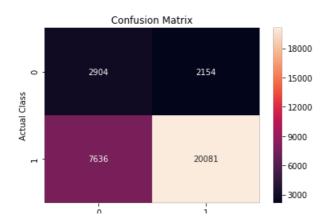
confusion matrix

In [58]:

```
y3_predict = grid.predict(set3_t)
cm3 = confusion_matrix(y_test,y3_predict)
# https://seaborn.pydata.org/generated/seaborn.heatmap.html
sns.heatmap(cm3, annot=True, fmt="d")
plt.ylabel("Actual Class")
plt.xlabel("Predicted Class")
plt.title("Confusion Matrix")
```

Out[58]:

Text(0.5, 1.0, 'Confusion Matrix')



Predicted Class

AUC CUrve

```
In [59]:
```

```
# probabilities calcultion
y3_predict_prob = ccv.predict_proba(set3_t)[:,1]
y3_predict_prob_train = ccv.predict_proba(set3)[:,1]

# took referance from https://stackoverflow.com/questions/25009284/how-to-plot-roc-curve-in-python
#fpr,tpr
fpr,tpr,thre = roc_curve(y_test,y3_predict_prob)

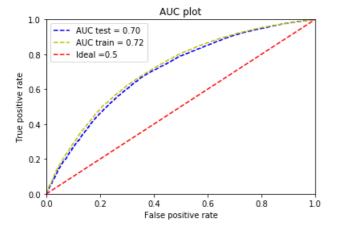
# am i doing it right here.....?
fpr_train,tpr_train,thre_train = roc_curve(y_train,y3_predict_prob_train)
```

In [60]:

```
# auc calculation for test data
roc_auc3 = metrics.auc(fpr,tpr)

# auc calculation for train data
roc_auc_train3 = metrics.auc(fpr_train,tpr_train)

# took referance from https://stackoverflow.com/questions/25009284/how-to-plot-roc-curve-in-python
plt.plot(fpr,tpr,"b--",label = 'AUC test = %0.2f'%roc_auc3)
plt.plot(fpr_train,tpr_train,"y--",label = 'AUC train = %0.2f'%roc_auc_train3)
plt.title("AUC plot")
plt.xlabel("False positive rate")
plt.ylabel("True positive rate")
plt.plot([0, 1], [0, 1],'r--',label = "Ideal =0.5")
plt.xlim([0,1])
plt.ylim([0,1])
plt.legend(loc = "upper left")
plt.show()
```



Set4 (TFIDF Avg_w2v)

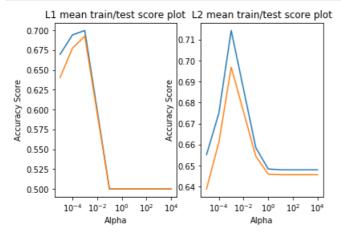
In [61]:

```
grid = GridSearchCV(sgd,par_grid,scoring="roc_auc",n_jobs=-1,cv=10)
grid.fit(set4,y_train)
print(grid.best_estimator_)
print(grid.best_index_)
print(grid.best_params_)
print(grid.best_score_)
```

```
SGDClassifier(alpha=0.001, average=False, class_weight='balanced', early_stopping=False, epsilon=0.1, eta0=0.0, fit_intercept=True, l1_ratio=0.15, learning_rate='optimal', loss='hinge', max_iter=None, n_iter=None, n_iter_no_change=5, n_jobs=None, penalty='l2', power t=0.5, random state=None, shuffle=True, tol=None,
```

```
validation fraction=0.1, verbose=0, warm start=False)
{'alpha': 0.001, 'penalty': '12'}
0.6968222367442298
In [62]:
sad =
SGDClassifier(alpha=grid.best params ["alpha"], class weight="balanced", penalty=grid.best params ["
penalty"])
ccv = CalibratedClassifierCV(sgd,cv=10)
ccv.fit(set4,y_train)
print(ccv.get_params)
<bound method BaseEstimator.get params of</pre>
CalibratedClassifierCV(base estimator=SGDClassifier(alpha=0.001, average=False,
class weight='balanced',
      early stopping=False, epsilon=0.1, eta0=0.0, fit intercept=True,
      11_ratio=0.15, learning_rate='optimal', loss='hinge', max_iter=None,
      n iter=None, n iter no change=5, n jobs=None, penalty='12',
      power_t=0.5, random_state=None, shuffle=True, tol=None,
      validation fraction=0.1, verbose=0, warm start=False),
           cv=10, method='sigmoid')>
In [63]:
#converting results to dataframe
df = pd.DataFrame(data = grid.cv results )
# getting into list
11 train score = []
l1 test score = []
12 train score = []
12 test score = []
for i in range(len(df)):
  if df.iloc[i]["param_penalty"] =="l1":
    11 test score.append(df.iloc[i]["mean test score"])
    11 train score.append(df.iloc[i]["mean train score"])
  if df.iloc[i]["param penalty"] =="12":
    12 test score.append(df.iloc[i]["mean test score"])
    12 train score.append(df.iloc[i]["mean train score"])
print(11 train score)
print(l1_test_score )
print(l2 train score)
print(12 test score)
[0.6696517912765817, 0.6940314360150116, 0.6995576237021873, 0.5, 0.5, 0.5, 0.5, 0.5, 0.5]
[0.6405377724273922,\ 0.6772690762178323,\ 0.6924008180309225,\ 0.5,\ 0.5,\ 0.5,\ 0.5,\ 0.5,\ 0.5]
0.6484185323882781,\ 0.6479936252088834,\ 0.6479923810460161,\ 0.6479927170289383,\ 0.647992301850221]
0.6459914617241826, 0.6456755453684208, 0.6456751028230016, 0.6456759063258709,
0.6456746065732635]
In [64]:
plt.figure()
plt.subplot(121)
plt.plot(alpha, 11_train_score)
plt.plot(alpha, 11 test score)
plt.xscale("log")
plt.xlabel("Alpha")
plt.ylabel("Accuracy Score")
plt.title("L1 mean train/test score plot")
plt.subplot(122)
plt.plot(alpha, 12 train score)
plt.plot(alpha,12 test score)
plt.xscale("log")
plt.xlabel("Alpha")
```

```
pit.ylabel("Accuracy Score")
plt.title("L2 mean train/test score plot")
plt.show()
```



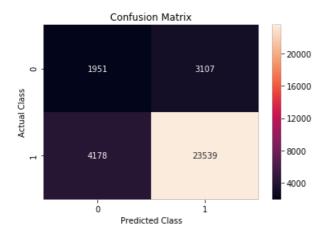
Confusion Matrix

In [65]:

```
y4_predict = grid.predict(set4_t)
cml = confusion_matrix(y_test,y4_predict)
# https://seaborn.pydata.org/generated/seaborn.heatmap.html
sns.heatmap(cml, annot=True, fmt="d")
plt.ylabel("Actual Class")
plt.xlabel("Predicted Class")
plt.title("Confusion Matrix")
```

Out[65]:

Text(0.5, 1.0, 'Confusion Matrix')



AUC Curve

In [66]:

```
# probabilities calcultion
y4_predict_prob = ccv.predict_proba(set4_t)[:,1]
y4_predict_prob_train = ccv.predict_proba(set4)[:,1]

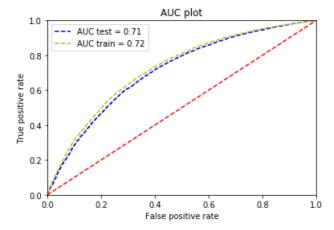
# took referance from https://stackoverflow.com/questions/25009284/how-to-plot-roc-curve-in-python
#fpr,tpr
fpr,tpr,thre = roc_curve(y_test,y4_predict_prob)

# am i doing it right here.....?
fpr_train,tpr_train,thre_train = roc_curve(y_train,y4_predict_prob_train)
```

```
# auc calculation for test data
roc_auc4 = metrics.auc(fpr,tpr)

# auc calculation for train data
roc_auc_train4 = metrics.auc(fpr_train,tpr_train)

# took referance from https://stackoverflow.com/questions/25009284/how-to-plot-roc-curve-in-python
plt.plot(fpr,tpr,"b--",label = 'AUC test = %0.2f'%roc_auc4)
plt.plot(fpr_train,tpr_train,"y--",label = 'AUC train = %0.2f'%roc_auc_train4)
plt.title("AUC plot")
plt.xlabel("False positive rate")
plt.ylabel("True positive rate")
plt.plot([0, 1], [0, 1],'r--')
plt.xlim([0,1])
plt.ylim([0,1])
plt.legend(loc = "upper left")
plt.show()
```



Task-2

Please refer attachments

DonorsChoose_7 SVM(Set5)

Observation

- 1. Best model found is Avg-W2v with Test AUC of 0.705 and Train AUC of 0.716
- 2. Maximum Train AUC found is 0.795 in case of BOW
- 3. 40k Training data point considered in Task2 due to Memory Error

Summary

```
In [77]:
```

```
from prettytable import PrettyTable
summary = PrettyTable()
```

```
In [78]:
```

```
summary.field_names = ["Set", "Vectorizer", "Model", "Hyperparameter", "Test", "Train"]
```

In [79]:

```
summary.add_row(["set1","BOW","SVM(Hinge Loss)","'alpha': 0.001, 'penalty': '12'","%0.3f"%roc_auc1
,"%0.3f"%roc_auc_train1])
summary.add_row(["set2","TFIDF","SVM(Hinge Loss)","'alpha': 0.001, 'penalty':
'12'","%0.3f"%roc_auc2,"%0.3f"%roc_auc_train2])
summary.add_row(["set1","Avg-W2v","SVM(Hinge Loss)","'alpha': 0.001, 'penalty': '12'","%0.3f"%roc_auc3,"%0.3f"%roc_auc_train3])
summary.add_row(["set2","TFIDF W2V","SVM(Hinge Loss)","'alpha': 0.001, 'penalty': '12'","%0.3f"%roc_auc3,"%0.3f"%roc_auc_train3])
```

```
c auc4,"%0.3f"%roc auc train4])
summary.add row(["set1", "TFIDF (Truncated SVD)", "SVM(Hinge Loss)", "'alpha': 0.0001, 'penalty': '12
"","0.64","0.63"])
In [80]:
print(summary)
| Set | Vectorizer | Model | Hyperparameter
                                                                 | Test | Train
+----+
             BOW
                         | SVM(Hinge Loss) | 'alpha': 0.001, 'penalty': '12' | 0.686 | 0.795
| set1 |
             TFIDF
                         | SVM(Hinge Loss) | 'alpha': 0.001, 'penalty': '12' | 0.686 | 0.731
| set2 |
                         | SVM(Hinge Loss) | 'alpha': 0.001, 'penalty': '12' | 0.705 | 0.716
| set1 |
            Avg-W2v
| set2 |
           TFIDF W2V
                         | SVM(Hinge Loss) | 'alpha': 0.001, 'penalty': '12' | 0.709 | 0.724
\mid set1 \mid TFIDF (Truncated SVD) \mid SVM(Hinge Loss) \mid 'alpha': 0.0001, 'penalty': '12' \mid 0.64 \mid
In [ ]:
```