```
DonorsChoose
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
import time
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 [2]:
project data = pd.read csv('train data.csv')
resource data = pd.read csv('resources.csv')
In [3]:
print(len(project_data))
109248
In [4]:
# how to replace elements in list python: https://stackoverflow.com/a/2582163/4084039
cols = ['Date' if x=='project_submitted_datetime' else x for x in list(project_data.columns)]
#sort dataframe based on time pandas python: https://stackoverflow.com/a/49702492/4084039
project data['Date'] = pd.to datetime(project data['project submitted datetime'])
```

project data.drop('project submitted datetime', axis=1, inplace=True)

project data.sort values(by=['Date'], inplace=True)

```
# how to reorder columns pandas python: https://stackoverflow.com/a/13148611/4084039
project_data = project_data[cols]

print(cols)
project_data.head(2)

['Unnamed: 0', 'id', 'teacher_id', 'teacher_prefix', 'school_state', 'Date',
'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']

Out[4]:
```

	Unnamed:	id	teacher_id	teacher_prefix	school_state	Date	project_grade_cate
55660	8393	p205479	2bf07ba08945e5d8b2a3f269b2b3cfe5	Mrs.	CA	2016- 04-27 00:27:36	Grades PreK-2
76127	37728	p043609	3f60494c61921b3b43ab61bdde2904df	Ms.	-	2016- 04-27 00:31:25	Grades 3-5
4)

```
In [5]:
```

```
# https://stackoverflow.com/questions/22407798/how-to-reset-a-dataframes-indexes-for-all-groups-in
-one-step
price_data = resource_data.groupby('id').agg({'price':'sum', 'quantity':'sum'}).reset_index()
# join two dataframes in python:
project_data = pd.merge(project_data, price_data, on='id', how='left')
```

In [6]:

```
#project_data = project_data.sample(frac=0.5)
```

Preprocessing data

1.2 preprocessing of project subject categories

In [7]:

```
catogories = list(project_data['project_subject_categories'].values)
# remove special characters from list of strings python:
https://stackoverflow.com/a/47301924/4084039
# https://www.geeksforgeeks.org/removing-stop-words-nltk-python/
# https://stackoverflow.com/questions/23669024/how-to-strip-a-specific-word-from-a-string
# https://stackoverflow.com/questions/8270092/remove-all-whitespace-in-a-string-in-python
cat list = []
for i in catogories:
   temp = ""
    # consider we have text like this "Math & Science, Warmth, Care & Hunger"
   for j in i.split(','): # it will split it in three parts ["Math & Science", "Warmth", "Care & E
unger"]
       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(' ','') # we are placeing all the ' '(space) with ''(empty) ex:"Math &
Science"=>"Math&Science"
       temp+=j.strip()+" " #" abc ".strip() will return "abc", remove the trailing spaces
       temp = temp.replace('&',' ') # we are replacing the & value into
   cat_list.append(temp.strip())
```

```
project data['clean categories'] = cat list
project data.drop(['project subject categories'], axis=1, inplace=True)
from collections import Counter
my counter = Counter()
for word in project data['clean categories'].values:
   my counter.update(word.split())
cat dict = dict(my_counter)
sorted cat dict = dict(sorted(cat dict.items(), key=lambda kv: kv[1]))
4
In [8]:
preprocessed_grade=project_data['project_grade_category']
In [9]:
new=[i.replace("-","_") for i in preprocessed_grade]
new=[i.replace(" "," ") for i in new]
In [10]:
project_data['preprocessed_grade']=new
In [11]:
print(project_data['preprocessed_grade'])
0
         Grades PreK 2
            Grades_3_5
1
         Grades PreK 2
2
3
         Grades_PreK_2
4
           Grades 3 5
5
           Grades 3 5
           Grades_3_5
6
7
            Grades 3 5
8
         Grades PreK 2
           Grades 3 5
9
10
        Grades PreK 2
11
          Grades 9 12
         Grades PreK 2
12
13
           Grades 3 5
           Grades_3_5
14
15
        Grades PreK 2
           Grades 3 5
16
17
           Grades_3_5
18
            Grades_3_5
          Grades_9_12
19
20
        Grades_PreK_2
          Grades 3 5
21
22
          Grades_9_12
23
          Grades_3_5
            Grades_3_5
24
        Grades_PreK 2
25
26
        Grades PreK 2
27
        Grades PreK 2
          Grades 3 5
2.8
29
        Grades PreK 2
             . . .
109218 Grades_PreK_2
109219 Grades PreK 2
109220
         Grades_3_5
109221 Grades_PreK_2
        Grades_PreK 2
109222
         Grades 9 12
109223
109224 Grades_PreK_2
109225
          Grades 3 5
109226 Grades_PreK_2
109227
           Grades 3 5
        Grades_PreK_2
109228
        Grades_PreK 2
109229
109230
            Grades 3 5
```

```
109231
           Grades 6 8
109232
           Grades 3 5
        Grades PreK 2
109233
109234
         Grades PreK 2
         Grades 3 5
109235
109236
          Grades 9 12
109237
       Grades PreK 2
109238
        Grades_PreK_2
109239
           Grades 6 8
109240
           Grades_9_12
           Grades 3 5
109241
109242
       Grades PreK 2
         Grades 9 12
109243
109244
        Grades PreK 2
109245
            Grades 3 5
           Grades_9 12
109246
109247 Grades PreK 2
Name: preprocessed grade, Length: 109248, dtype: object
In [12]:
print(project data['clean categories'].unique())
['Math_Science' 'SpecialNeeds' 'Literacy_Language' 'AppliedLearning'
 'Math Science History Civics' 'Literacy Language Math Science'
 'AppliedLearning Music Arts' 'Math Science AppliedLearning'
 'Math Science Literacy_Language' 'History_Civics Literacy_Language'
 'AppliedLearning Health Sports' 'Math Science Music Arts'
 'AppliedLearning Literacy_Language' 'Music_Arts' 'Health_Sports'
 'Literacy Language SpecialNeeds' 'Math Science SpecialNeeds'
 'AppliedLearning History Civics' 'AppliedLearning SpecialNeeds'
 'Health Sports Literacy_Language' 'Literacy_Language Music_Arts'
 'History Civics Math Science' 'SpecialNeeds Health Sports'
 'Literacy Language History Civics' 'Health Sports SpecialNeeds'
 'History Civics Music Arts' 'Math Science Health Sports'
 'Music Arts SpecialNeeds' 'SpecialNeeds Music Arts'
 'Health Sports History_Civics' 'History_Civics'
 'Health Sports AppliedLearning' 'History Civics SpecialNeeds'
 'AppliedLearning Math Science' 'Health Sports Music Arts'
 'Literacy_Language Health_Sports' 'Literacy_Language AppliedLearning'
 'Music Arts Health Sports' 'Music Arts AppliedLearning'
 'Music_Arts History_Civics' 'Health_Sports Math_Science'
 'History_Civics AppliedLearning' 'History_Civics Health_Sports'
 'Health Sports Warmth Care Hunger' 'History Civics Warmth Care Hunger'
 'Math_Science Warmth Care_Hunger' 'SpecialNeeds Warmth Care_Hunger'
 'Warmth Care_Hunger' 'Literacy_Language Warmth Care_Hunger'
 'Music Arts Warmth Care Hunger' 'AppliedLearning Warmth Care Hunger']
```

1.3 preprocessing of project_subject_subcategories

In [13]:

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

1.4 Preprocessing of project_grade_category

1.3 Text preprocessing

In [14]:

In [15]:

```
# 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 [16]:

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

```
've', 'y', 'ain', 'aren', "aren't", 'couldn', "couldn't", 'didn', "didn't", 'doesn', "doesn', "doesn',
                            "hadn't", 'hasn', "hasn't", 'haven', "haven't", 'isn', "isn't", 'ma', 'mightn',
"mightn't", 'mustn',\
                           "mustn't", 'needn', "needn't", 'shan', "shan't", 'shouldn', "shouldn't", 'wasn',
"wasn't", 'weren', "weren't", \
                           'won', "won't", 'wouldn', "wouldn't"]
4
In [17]:
# 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%|
[1:02:23<00:00, 29.18it/s]
In [18]:
project data['project resource summary']
preprocessed resource summary=[]
for sentence in tqdm(project_data['project_resource_summary'].values):
        sent = decontracted(sentence)
         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.lower() for e in sent.split() if e not in stopwords)
         preprocessed resource summary.append(sent.lower().strip())
100%|
                                                                                                                                                                           109248/109248
[20:29<00:00, 88.87it/s]
In [19]:
project data['clean resource summary'] = preprocessed resource summary
```

1.4 Preprocessing of `project_title`

In [20]:

```
# Combining all the above statemennts
from tqdm import tqdm
preprocessed titles = []
# tqdm is for printing the status bar
for sentence in tqdm(project data['project title'].values):
   sent = decontracted(sentence)
   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 titles.append(sent.lower().strip())
                                                                   109248/109248
100%|
[05:35<00:00, 325.88it/s]
```

```
In [21]:
```

```
#Adding processed columns at place of original columns
project_data['clean_essays'] = preprocessed_essays
project_data.drop(['project_essay_1'], axis=1, inplace=True)
project_data.drop(['project_essay_2'], axis=1, inplace=True)
project_data.drop(['project_essay_3'], axis=1, inplace=True)
project_data.drop(['project_essay_4'], axis=1, inplace=True)
```

In [22]:

```
project_data['clean_titles'] = preprocessed_titles
```

In [23]:

```
# we cannot remove rows where teacher prefix is not available therefore we are replacing 'nan' val
ue with
# 'null'(string)
#https://stackoverflow.com/questions/42224700/attributeerror-float-object-has-no-attribute-split
project_data['teacher_prefix'] = project_data['teacher_prefix'].fillna('null')
```

In [24]:

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

Out[24]:

0 8393 p205479 2bf07ba08945e5d8b2a3f269b2b3cfe5 Mrs. CA 2016- 04-27 00:27:36 Grades PreK-2 1 37728 p043609 3f60494c61921b3b43ab61bdde2904df Ms. UT 2016- 04-27 00:31:25 Grades 3-5 00:31:25		Unnamed:	id	teacher_id	teacher_prefix	school_state	Date	project_grade_category
1 37728 p043609 3f60494c61921b3b43ab61bdde2904df Ms. UT 04-27 Grades 3-5	0	8393	p205479	2bf07ba08945e5d8b2a3f269b2b3cfe5	Mrs.	CA	04-27	
	1	37728	p043609	3f60494c61921b3b43ab61bdde2904df	Ms.	UT	04-27	

In [25]:

```
filtered_negative = project_data.loc[project_data['project_is_approved'] == 0]
#print(filtered_negative.count())

filtered_positive = project_data.loc[project_data['project_is_approved'] == 1]
sample_positive = filtered_positive.take(np.random.permutation(len(filtered_positive))[:50000])
```

In [26]:

```
project_data = pd.concat([filtered_negative, sample_positive]).sort_index(kind='merge')
```

In [27]:

```
project_data.count()
```

Out[27]:

```
Unnamed: 0 66542 id 66542 teacher_id 66542 teacher prefix 66542
```

```
66542
school_state
Date
                                                  66542
project_grade_category
                                                  66542
project title
                                                  66542
project_resource_summary
                                                  66542
teacher_number_of_previously_posted_projects
                                                 66542
project is approved
                                                  66542
                                                  66542
price
quantity
                                                  66542
clean_categories
                                                  66542
preprocessed grade
                                                  66542
clean subcategories
                                                  66542
                                                  66542
                                                 66542
clean_resource_summary
                                                 66542
clean essays
                                                  66542
clean_titles
dtype: int64
```

So far we have preprocessed the data. Next is to split and vectorize data for BoW,TFIDF,Avg W2Vec and TFIDF weighted W2Vec

1. Splitting data

```
In [28]:
```

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

In [29]:

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

In [30]:

```
x = np.count_nonzero(y_test)
print(len(y_test) - x)
```

5459

In [31]:

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

print("="*100)

(29870, 19) (29870,)
(14713, 19) (14713,)
(21959, 19) (21959,)
```

2. Vectorizing data

BoW

2.1 Text data

```
In [32]:
```

```
trom sklearn.teature_extraction.text import CountVectorizer
vectorizer = CountVectorizer(min_df=10, max_features=5000)
vectorizer.fit(X train['clean essays'].values) # fit has to happen only on train data
# we use the fitted CountVectorizer to convert the text to vector
X train essay bow = vectorizer.transform(X train['clean essays'].values)
X cv essay bow = vectorizer.transform(X cv['clean essays'].values)
X_test_essay_bow = vectorizer.transform(X_test['clean_essays'].values)
print("After vectorizations")
print(X_train_essay_bow.shape, y_train.shape)
print(X_cv_essay_bow.shape, y_cv.shape)
print(X_test_essay_bow.shape, y_test.shape)
print("="*100)
After vectorizations
(29870, 5000) (29870,)
(14713, 5000) (14713,)
(21959, 5000) (21959,)
In [33]:
from sklearn.feature extraction.text import CountVectorizer
vectorizer = CountVectorizer (min df=10, max features=5000)
vectorizer.fit(X_train['clean_titles'].values) # fit has to happen only on train data
# we use the fitted CountVectorizer to convert the text to vector
X train titles bow = vectorizer.transform(X train['clean titles'].values)
X_cv_titles_bow = vectorizer.transform(X_cv['clean_titles'].values)
X test titles bow = vectorizer.transform(X test['clean titles'].values)
print("After vectorizations")
print(X train titles bow.shape, y train.shape)
print(X_cv_titles_bow.shape, y_cv.shape)
print(X test titles bow.shape, y test.shape)
print("="*100)
After vectorizations
(29870, 1505) (29870,)
(14713, 1505) (14713,)
(21959, 1505) (21959,)
4
In [34]:
from sklearn.feature_extraction.text import CountVectorizer
vectorizer = CountVectorizer(min df=10, max features=5000)
vectorizer.fit(X train['clean resource summary'].values) # fit has to happen only on train data
#Bow FeatureList=Bow FeatureList + (vectorizer.get feature names())
# we use the fitted CountVectorizer to convert the text to vector
X train summary bow = vectorizer.transform(X train['clean resource summary'].values)
X cv summary_bow = vectorizer.transform(X_cv['clean_resource_summary'].values)
X test summary bow = vectorizer.transform(X test['clean resource summary'].values)
print("After vectorizations")
print(X train summary bow.shape, y train.shape)
print(X_cv_summary_bow.shape, y_cv.shape)
print(X test summary bow.shape, y test.shape)
print("="*100)
After vectorizations
(29870, 2952) (29870,)
(14713, 2952) (14713,)
(21959, 2952) (21959,)
In [35]:
```

X train summary bow.shape

```
Out[35]:
(29870, 2952)
```

2.2 one hot encoding the catogorical features: clean_categories

```
In [36]:
```

```
vectorizer = CountVectorizer()
vectorizer.fit(X_train['clean_categories'].values) # fit has to happen only on train data
# we use the fitted CountVectorizer to convert the text to vector
X train clean cat ohe = vectorizer.transform(X train['clean categories'].values)
X cv clean cat ohe = vectorizer.transform(X cv['clean categories'].values)
X test clean cat ohe = vectorizer.transform(X test['clean categories'].values)
print("After vectorizations")
print(X train clean_cat_ohe.shape, y_train.shape)
print(X_cv_clean_cat_ohe.shape, y_cv.shape)
print(X_test_clean_cat_ohe.shape, y_test.shape)
print(vectorizer.get feature names())
print("="*100)
After vectorizations
(29870, 9) (29870,)
(14713, 9) (14713,)
(21959, 9) (21959,)
['appliedlearning', 'care_hunger', 'health_sports', 'history_civics', 'literacy_language',
'math science', 'music arts', 'specialneeds', 'warmth']
```

2.3 one hot encoding the catogorical features: clean subcategories

```
In [37]:
```

```
vectorizer = CountVectorizer()
vectorizer.fit(X train['clean subcategories'].values) # fit has to happen only on train data
# we use the fitted CountVectorizer to convert the text to vector
X train clean subcat ohe = vectorizer.transform(X train['clean subcategories'].values)
X_cv_clean_subcat_ohe = vectorizer.transform(X_cv['clean_subcategories'].values)
X test clean subcat ohe = vectorizer.transform(X test['clean subcategories'].values)
print("After vectorizations")
print(X train clean subcat ohe.shape, y train.shape)
print(X_cv_clean_subcat_ohe.shape, y_cv.shape)
print(X_test_clean_subcat_ohe.shape, y_test.shape)
print(vectorizer.get feature names())
print("="*100)
After vectorizations
(29870, 30) (29870,)
(14713, 30) (14713,)
(21959, 30) (21959,)
['appliedsciences', 'care hunger', 'charactereducation', 'civics government',
'college careerprep', 'communityservice', 'earlydevelopment', 'economics', 'environmentalscience',
'esl', 'extracurricular', 'financialliteracy', 'foreignlanguages', 'gym fitness',
'health_lifescience', 'health_wellness', 'history_geography', 'literacy', 'literature_writing', 'm
athematics', 'music', 'nutritioneducation', 'other', 'parentinvolvement', 'performingarts', 'socia
lsciences', 'specialneeds', 'teamsports', 'visualarts', 'warmth']
4
```

```
In [38]:
vectorizer = CountVectorizer()
vectorizer.fit(X train['teacher prefix'].values) # fit has to happen only on train data
# we use the fitted CountVectorizer to convert the text to vector
X train teacher ohe = vectorizer.transform(X train['teacher prefix'].values)
X cv teacher ohe = vectorizer.transform(X cv['teacher prefix'].values)
X_test_teacher_ohe = vectorizer.transform(X_test['teacher_prefix'].values)
print("After vectorizations")
print(X_train_teacher_ohe.shape, y_train.shape)
print(X_cv_teacher_ohe.shape, y_cv.shape)
print(X test teacher ohe.shape, y test.shape)
print(vectorizer.get feature names())
print("="*100)
After vectorizations
(29870, 6) (29870,)
(14713, 6) (14713,)
(21959, 6) (21959,)
['dr', 'mr', 'mrs', 'ms', 'null', 'teacher']
4
2.4 one hot encoding the catogorical features: school state
In [39]:
vectorizer = CountVectorizer()
vectorizer.fit(X train['school state'].values) # fit has to happen only on train data
# we use the fitted CountVectorizer to convert the text to vector
X train state ohe = vectorizer.transform(X train['school state'].values)
X cv state ohe = vectorizer.transform(X cv['school state'].values)
X test state ohe = vectorizer.transform(X test['school state'].values)
print("After vectorizations")
```

```
print(X train state ohe.shape, y train.shape)
print(X_cv_state_ohe.shape, y_cv.shape)
print(X test state ohe.shape, y test.shape)
print(vectorizer.get feature names())
print("="*100)
```

```
After vectorizations
(29870, 51) (29870,)
(14713, 51) (14713,)
(21959, 51) (21959,)
['ak', 'al', 'ar', 'az', 'ca', 'co', 'ct', 'dc', 'de', 'fl', 'ga', 'hi', 'ia', 'id', 'il', 'in', 'k
s', 'ky', 'la', 'ma', 'md', 'me', 'mi', 'mn', 'mo', 'ms', 'mt', 'nc', 'nd', 'ne', 'nh', 'nj', 'nm',
'nv', 'ny', 'oh', 'ok', 'or', 'pa', 'ri', 'sc', 'sd', 'tn', 'tx', 'ut', 'va', 'vt', 'wa', 'wi', 'ww
', 'wy']
```

2.4 one hot encoding the catogorical features: project_grade_category

4

Out[40]:

In [40]: X train.head(2)

	Unnamed:	id	teacher_id	teacher_prefix	school_state	Date	project_grade_cate
						2016-	

39765	Uଡ଼ିନିର୍ଲୋନିed: 0	p219272 id	b94a387c93dfcccec1a93c72931f2f3f teacher_id	Ms. teacher_prefix	FL school_state	08-31 Date 23:57:37	Grades PreK-2 project_grade_cate
68257	107753	p017066	b03ea1aee38240e5e38e9fe33876b7fa	Mrs.	ОН	2016- 11-17 15:27:34	Grades PreK-2

In [41]:

In [42]:

```
vectorizer = CountVectorizer(vocabulary=list(sorted_project_grade_category_dict.keys()), lowercase
=False, binary=True)
vectorizer.fit(X_train['preprocessed_grade'].values) # fit has to happen only on train data

# we use the fitted CountVectorizer to convert the text to vector
X_train_grade_ohe = vectorizer.transform(X_train['preprocessed_grade'].values)
X_cv_grade_ohe = vectorizer.transform(X_cv['preprocessed_grade'].values)
X_test_grade_ohe = vectorizer.transform(X_test['preprocessed_grade'].values)

print("After vectorizations")
print(X_train_grade_ohe.shape, y_train.shape)
print(X_cv_grade_ohe.shape, y_test.shape)
print(X_test_grade_ohe.shape, y_test.shape)
print(vectorizer.get_feature_names())
print("="*100)
```

```
After vectorizations
(29870, 4) (29870,)
(14713, 4) (14713,)
(21959, 4) (21959,)
['Grades_9_12', 'Grades_6_8', 'Grades_3_5', 'Grades_PreK_2']
```

4

2.5 Normalizing the numerical features: Price

In [43]:

```
X_train.head(2)
```

Out[43]:

	Unnamed:	id	teacher_id	teacher_prefix	school_state	Date	project_grade_cate
						2016	
39765	169616	p219272	b94a387c93dfcccec1a93c72931f2f3f	Ms.	FL	2016- 08-31	Grades PreK-2

68257 107753 p017066 b03ea1aee38240e5e38e9fe	e33876b7fa	Mrs.	ОН	2016- 11-17 15:27:34	Grades PreK-2

In [44]:

```
from sklearn.preprocessing import StandardScaler
standard vec = StandardScaler(with_mean = False)
# this will rise an error Expected 2D array, got 1D array instead:
# array=[105.22 215.96 96.01 ... 368.98 80.53 709.67].
# Reshape your data either using
# array.reshape(-1, 1) if your data has a single feature
# array.reshape(1, -1) if it contains a single sample.
standard vec.fit(X train['price'].values.reshape(-1,1))
X train price std = standard vec.transform(X train['price'].values.reshape(-1,1))
X cv price std = standard vec.transform(X cv['price'].values.reshape(-1,1))
X test price std = standard vec.transform(X test['price'].values.reshape(-1,1))
print("After vectorizations")
print(X_train_price_std.shape, y_train.shape)
print(X_cv_price_std.shape, y_cv.shape)
print(X test price std.shape, y test.shape)
print("="*100)
```

After vectorizations
(29870, 1) (29870,)
(14713, 1) (14713,)
(21959, 1) (21959,)

[4]

2.6 Vectorizing numerical features: teacher_number_of_previously _posted_projects"

In [45]:

```
\textbf{from sklearn.preprocessing import} \ \texttt{StandardScaler}
standard vec = StandardScaler(with mean = False)
# this will rise an error Expected 2D array, got 1D array instead:
# array=[105.22 215.96 96.01 ... 368.98 80.53 709.67].
# Reshape your data either using
# array.reshape(-1, 1) if your data has a single feature
# array.reshape(1, -1) if it contains a single sample.
standard vec.fit(X train['teacher number of previously posted projects'].values.reshape(-1,1))
X train projects std =
standard_vec.transform(X_train['teacher_number_of_previously_posted_projects'].values.reshape(-1,1
X cv projects std = standard vec.transform(X cv['teacher number of previously posted projects'].va
lues.reshape(-1,1))
X_test_projects_std = standard_vec.transform(X_test['teacher_number_of_previously posted projects'
].values.reshape(-1,1))
print("After vectorizations")
print(X train projects std.shape, y train.shape)
print(X_cv_projects_std.shape, y_cv.shape)
print(X test_projects_std.shape, y_test.shape)
print("="*100)
```

After vectorizations (29870, 1) (29870,) (14713, 1) (14713,)

```
(21959, 1) (21959,)
4
In [46]:
from sklearn.preprocessing import StandardScaler
standard vec = StandardScaler(with mean = False)
# this will rise an error Expected 2D array, got 1D array instead:
# array=[105.22 215.96 96.01 ... 368.98 80.53 709.67].
# Reshape your data either using
# array.reshape(-1, 1) if your data has a single feature
# array.reshape(1, -1) if it contains a single sample.
standard vec.fit(X train['quantity'].values.reshape(-1,1))
X train qty std = standard vec.transform(X train['quantity'].values.reshape(-1,1))
X cv qty std = standard vec.transform(X cv['quantity'].values.reshape(-1,1))
X test qty std = standard vec.transform(X test['quantity'].values.reshape(-1,1))
print("After vectorizations")
print(X_train_qty_std.shape, y_train.shape)
print(X_cv_qty_std.shape, y_cv.shape)
print(X test qty std.shape, y test.shape)
print("="*100)
After vectorizations
(29870, 1) (29870,)
(14713, 1) (14713,)
(21959, 1) (21959,)
                                                                                                  - P
In [47]:
#function to get heatmap confusion matrix for obtaining colur encoded matrix
#Reference link https://seaborn.pydata.org/generated/seaborn.heatmap.html
def get confusion matrix(clf, X te, y test):
    y pred = clf.predict(X te)
    df cm = pd.DataFrame(confusion_matrix(y_test, y_pred), range(2), range(2))
    df cm.columns = ['Predicted NO', 'Predicted YES']
    df cm = df cm.rename({0: 'Actual NO', 1: 'Actual YES'})
    sns.set(font_scale=1.4)#for label size
```

2.7 Concatinating all the features

sns.heatmap(df cm, annot=True, annot kws={"size": 16}, fmt='g')

```
In [48]:
```

```
# merge two sparse matrices: https://stackoverflow.com/a/19710648/4084039
from scipy.sparse import hstack
X tr =
hstack((X_train_essay_bow,X_train_titles_bow,X_train_summary_bow,X_train_clean_cat_ohe,X_train_clea
n_subcat_ohe, X_train_state_ohe, X_train_teacher_ohe, X_train_grade_ohe,
X_train_price_std, X_train_projects_std, X_train_qty_std)).tocsr()
X cr =
hstack((X_cv_essay_bow,X_cv_titles_bow,X_cv_summary_bow,X_cv_clean_cat_ohe,X_cv_clean_subcat_ohe,
X_cv_state_ohe, X_cv_teacher_ohe, X_cv_grade_ohe, X_cv_price_std,X_cv_projects_std,X_cv_qty_std)).t
ocsr()
X te =
hstack((X test essay bow, X test titles bow, X test summary bow, X test clean cat ohe, X test clean suk
cat ohe, X test state ohe, X test teacher ohe, X test grade ohe,
X_test_price_std, X_test_projects_std, X_test_qty_std)).tocsr()
print("Final Data matrix")
print(X_tr.shape, y_train.shape)
print(X cr.shape, y cv.shape)
print(X_te.shape, y_test.shape)
print("="*100)
```

Final Data matrix (29870, 9560) (29870,)

```
(14713, 9560) (14713,)
(21959, 9560) (21959,)
```

Applying Logistic Regression

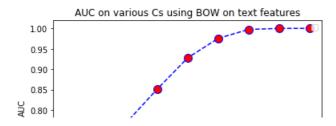
In [49]:

```
%%time
# https://scikit-learn.org/stable/modules/generated/sklearn.model selection.GridSearchCV.html
from sklearn.model_selection import GridSearchCV
from sklearn.linear_model import LogisticRegression
from sklearn.metrics import roc_auc_score
import math
def predAUC(X_tr,y_train):
    Cs = [10**i for i in range(-5,4)]
    parameters = { 'C':Cs}
    LoR = LogisticRegression(class_weight='balanced')
   clf = GridSearchCV(LoR, parameters, cv=3, scoring='roc auc')
    clf.fit(X tr, y train)
    auc_tr = clf.cv_results_['mean_train_score']
    auc tr std = clf.cv results ['std train score']
    auc_cv = clf.cv_results_['mean_test_score']
    auc_cv_std= clf.cv_results_['std_test_score']
   print(clf.best estimator )
    print(clf.score(X_te, y_test))
    return Cs, auc_tr, auc_cv
def plotPerformance(Cs, auc_tr, auc_cv, plt_title):
    #plt.plot(Cs, auc_tr, label='AUC Train')
    #plt.plot(Cs, auc_cv, label='AUC_Validation')
    #plt.scatter(Cs, auc_tr, label='Coordinates')
#plt.scatter(Cs, auc_cv, label='Coordinates')
    newmylist=[math.log10(i) for i in Cs]
   plt.plot(newmylist,list(auc tr),color='blue', linestyle='dashed',
marker='o',markerfacecolor='red', markersize=10)
   plt.plot(newmylist,list(auc_cv),color='orange', linestyle='dashed',
marker='o', markerfacecolor='red', markersize=10)
   plt.xlabel('Hyperparameter - C')
   plt.ylabel('AUC')
   plt.title("AUC on various Cs using %s on text features"%plt_title)
    plt.legend()
```

Wall time: 0 ns

```
In [50]:
```

No handles with labels found to put in legend.



```
0.75

0.70

0.65

0.60

-5 -4 -3 -2 -1 0 1 2 3

Hyperparameter - C
```

In [51]:

Logistic Regression with optimal alpha

```
In [52]:
```

```
def Get_LOR_OptimalC(optimal_c):
    LoR = LogisticRegression(C=optimal_c,class_weight = 'balanced')
    LoR.fit(X_tr, y_train)
    return LoR
```

In [53]:

```
def getROC_Data(data_pnts_test, y_test, data_pnts_tr, y_tr, LoR):
    predicted_y_test = LoR.predict_proba(data_pnts_test)[:, 1]
    predicted_y_tr = LoR.predict_proba(data_pnts_tr)[:, 1]

    fpr_test, tpr_test, thres_test = roc_curve(y_test, predicted_y_test)
    fpr_tr, tpr_tr, thres_tr = roc_curve(y_tr, predicted_y_tr)

    return [fpr_test, tpr_test, thres_test], [fpr_tr, tpr_tr, thres_tr]

def makeROC(test_data, train_data, plt_title):
    fpr_tr, tpr_tr, = train_data
    fpr_test, tpr_test, = test_data

plt.plot(fpr_tr, tpr_tr, label='AUC_Train')
    plt.plot(fpr_test, tpr_test, label='AUC_Trest')
    plt.title("ROC Curve using %s on text features"%plt_title)

plt.xlabel('FPR')
    plt.ylabel('TPR')
    plt.legend()
```

In [54]:

```
optimal_c=0.01
LOR=Get_LOR_OptimalC(optimal_c)
```

In [55]:

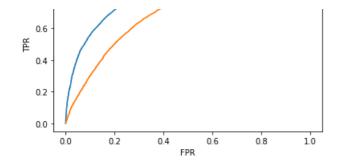
```
roc_data_test1, roc_data_train1 = getROC_Data(X_te, y_test, X_tr, y_train, LOR)
```

In [56]:

```
makeROC(roc_data_test1, roc_data_train1, plt_title1)
```

ROC Curve using BOW on text features

```
1.0 AUC_Train AUC_Test
```



In [57]:

```
get_confusion_matrix(LOR, X_tr, y_train)
```



In [58]:

```
get_confusion_matrix(LOR, X_te, y_test)
```



In [59]:

```
predbow = (LOR.predict_proba(X_te))
predbow_train=(LOR.predict_proba(X_tr))
```

In [61]:

```
print("AUC score for Logistic regression c_auc_score(y_test ,predbow[:,1]),3))
print("AUC score for Logistic regression print("AUC score for Logistic regression core(y_train ,predbow_train[:,1]),3))
Bag of Words for Training data ",round(metrics.roc_auc_s core(y_train ,predbow_train[:,1]),3))
```

AUC score for Logistic regression $\,$ with Bag of Words for Testing data is $\,$ 0.725 AUC score for Logistic regression $\,$ Bag of Words for Training data $\,$ 0.836 $\,$

2.4.2 Applying Logistic Regression on TFIDF, SET 2

```
In [62]:
%%time
from sklearn.feature extraction.text import TfidfVectorizer
from sklearn.feature selection import SelectKBest, chi2
vectorizer = TfidfVectorizer(min df=10)
vectorizer.fit(X train['clean essays'].values) # fit has to happen only on train data
# we use the fitted CountVectorizer to convert the text to vector
X train essay tfidf = vectorizer.transform(X train['clean essays'].values)
X cv essay tfidf = vectorizer.transform(X cv['clean essays'].values)
X_test_essay_tfidf = vectorizer.transform(X_test['clean_essays'].values)
Wall time: 19.4 s
In [63]:
from sklearn.feature_extraction.text import TfidfVectorizer
vectorizer = TfidfVectorizer(min df=5)
vectorizer.fit(X train['clean titles'].values) # fit has to happen only on train data
# we use the fitted CountVectorizer to convert the text to vector
X train titles tfidf = vectorizer.transform(X train['clean titles'].values)
X cv titles tfidf = vectorizer.transform(X cv['clean titles'].values)
X test titles tfidf = vectorizer.transform(X test['clean_titles'].values)
print("Train shape:",X train titles tfidf.shape)
print("CV shape:",X_cv_titles_tfidf.shape)
print("Test shape:", X_test_titles_tfidf.shape)
Train shape: (29870, 2443)
CV shape: (14713, 2443)
Test shape: (21959, 2443)
In [64]:
vectorizer = TfidfVectorizer(min df=5)
vectorizer.fit(X_train['project_resource_summary'].values) # fit has to happen only on train
# we use the fitted CountVectorizer to convert the text to vector
X train summary tfidf = vectorizer.transform(X train['project resource summary'].values)
X cv summary tfidf = vectorizer.transform(X cv['project resource summary'].values)
X test summary tfidf = vectorizer.transform(X test['project resource summary'].values)
print("After vectorizations")
print(X train summary_tfidf.shape, y_train.shape)
print(X_cv_summary_tfidf.shape, y_cv.shape)
print(X_test_summary_tfidf.shape, y_test.shape)
print("="*100)
After vectorizations
(29870, 4618) (29870,)
(14713, 4618) (14713,)
(21959, 4618) (21959,)
```

Concatinating all features (TFIDF)

```
In [65]:
```

```
# merge two sparse matrices: https://stackoverflow.com/a/19710648/4084039
from scipy.sparse import hstack
X_tr =
hstack((X_train_essay_tfidf,X_train_titles_tfidf,X_train_summary_tfidf,X_train_clean_cat_ohe,X_train_clean_subcat_ohe, X_train_state_ohe, X_train_teacher_ohe, X_train_grade_ohe, X_train_price_std,X_train_projects_std,X_train_qty_std)).tocsr()
X_cr =
hstack((X_cv_essay_tfidf,X_cv_titles_tfidf,X_cv_summary_tfidf,X_cv_clean_cat_ohe,X_cv_clean_subcat_
```

```
ohe, X_cv_state_ohe, X_cv_teacher_ohe, X_cv_grade_ohe, X_cv_price_std, X_cv_projects_std, X_cv_qty_st
 d)).tocsr()
\textbf{X\_te} = \texttt{hstack((X\_test\_essay\_tfidf,X\_test\_titles\_tfidf,X\_test\_summary\_tfidf,X\_test\_clean\_cat\_ohe,X\_test\_titles\_tfidf,X\_test\_summary\_tfidf,X\_test\_clean\_cat\_ohe,X\_test\_titles\_tfidf,X\_test\_summary\_tfidf,X\_test\_titles\_tfidf,X\_test\_titles\_tfidf,X\_test\_titles\_tfidf,X\_test\_titles\_tfidf,X\_test\_titles\_tfidf,X\_test\_titles\_tfidf,X\_test\_titles\_tfidf,X\_test\_titles\_tfidf,X\_test\_titles\_tfidf,X\_test\_titles\_tfidf,X\_test\_titles\_tfidf,X\_test\_titles\_tfidf,X\_test\_titles\_tfidf,X\_test\_titles\_tfidf,X\_test\_titles\_tfidf,X\_test\_titles\_tfidf,X\_test\_titles\_tfidf,X\_test\_titles\_tfidf,X\_test\_titles\_tfidf,X\_test\_titles\_tfidf,X\_test\_titles\_tfidf,X\_test\_titles\_tfidf,X\_test\_titles\_tfidf,X\_test\_titles\_tfidf,X\_test\_titles\_tfidf,X\_test\_titles\_tfidf,X\_test\_titles\_tfidf,X\_test\_titles\_tfidf,X\_test\_titles\_tfidf,X\_test\_titles\_tfidf,X\_test\_titles\_tfidf,X\_test\_titles\_tfidf,X\_test\_titles\_tfidf,X\_test\_titles\_tfidf,X\_test\_titles\_tfidf,X\_test\_titles\_tfidf,X\_test\_titles\_tfidf,X\_test\_titles\_tfidf,X\_test\_titles\_tfidf,X\_test\_titles\_tfidf,X\_test\_titles\_tfidf,X\_test\_titles\_tfidf,X\_test\_titles\_tfidf,X\_test\_titles\_tfidf,X\_test\_titles\_tfidf,X\_test\_titles\_tfidf,X\_test\_titles\_tfidf,X\_test\_titles\_tfidf,X\_test\_titles\_tfidf,X\_test\_titles\_tfidf,X\_test\_titles\_tfidf,X\_test\_titles\_tfidf,X\_test\_titles\_tfidf,X\_test\_titles\_tfidf,X\_test\_titles\_tfidf,X\_test\_titles\_tfidf,X\_test\_titles\_tfidf,X\_test\_titles\_tfidf,X\_test\_titles\_tfidf,X\_test\_titles\_tfidf,X\_test\_titles\_tfidf,X\_test\_titles\_tfidf,X\_test\_titles\_tfidf,X\_test\_titles\_tfidf,X\_test\_titles\_tfidf,X\_test\_titles\_tfidf,X\_test\_titles\_tfidf,X\_test\_titles\_tfidf,X\_test\_titles\_tfidf,X\_test\_titles\_tfidf,X\_test\_titles\_tfidf,X\_test\_titles\_tfidf,X\_test\_titles\_tfidf,X\_test\_titles\_tfidf,X\_test\_titles\_tfidf,X\_test\_titles\_tfidf,X\_test\_titles\_tfidf,X\_test\_titles\_tfidf,X\_test\_titles\_tfidf,X\_test\_titles\_tfidf,X\_test\_titles\_tfidf,X\_test\_titles\_tfidf,X\_test\_titles\_tfidf,X\_test\_titles\_tfidf,X\_test\_titles\_tfidf,X\_test\_titles\_tfidf,X\_test\_titles\_tfidf,X\_test\_titles\_tfidf,X\_test\_titles\_tfidf,X\_test\_titles\_tfidf,X\_test\_titles\_tfidf,X\_test\_titles\_tfidf,
 test_clean_subcat_ohe, X_test_state_ohe, X_test_teacher_ohe, X_test_grade_ohe, X_test_price_std,X_t
est_projects_std, X_test_qty_std)).tocsr()
print("Final Data matrix")
print(X tr.shape, y train.shape)
print(X_cr.shape, y_cv.shape)
print(X_te.shape, y_test.shape)
 print("="*100)
 4
Final Data matrix
(29870, 16989) (29870,)
 (14713, 16989) (14713,)
 (21959, 16989) (21959,)
```

APPLYING Logistic regression ON TFIDF

In [66]:

```
Cs, auc_tr, auc_cv=predAUC(X_tr,y_train)
plt_title1 = 'TFIDF'
plotPerformance(Cs, auc_tr, auc_cv, plt_title1)
```

LogisticRegression(C=0.1, class weight='balanced', dual=False, fit_intercept=True, intercept_scaling=1, max_iter=100, multi class='ovr', n jobs=1, penalty='12', random state=None, solver='liblinear', tol=0.0001, verbose=0, warm_start=False) 0.723496078202801

No handles with labels found to put in legend.

AUC on various Cs using TFIDF on text features 1.0 0.9 8.0 AUC 0.7 -2 0 2 -4 Hyperparameter - C

In [67]:

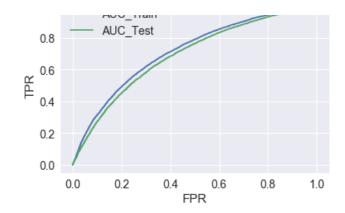
```
optimal_c=0.01
LOR=Get LOR OptimalC(optimal c)
```

In [68]:

```
roc_data_test1, roc_data_train1 = getROC_Data(X_te, y_test,X_tr, y_train, LOR)
```

In [69]:

```
makeROC(roc_data_test1, roc_data_train1, plt_title1)
```



In [70]:

```
{\tt get\_confusion\_matrix} \, ({\tt LOR}, {\tt X\_tr}, {\tt y\_train})
```



In [71]:

```
%%time
get_confusion_matrix(LOR, X_te, y_test)
```

Wall time: 140 ms



In [72]:

```
predbow = (LOR.predict_proba(X_te))
predbow_train=(LOR.predict_proba(X_tr))
```

In [73]:

```
print("AUC score for Logistic regression with TFIDF for Testing data is
",round(metrics.roc_auc_score(y_test ,predbow[:,1]),3))
print("AUC score for Logistic regression with TFIDF for Training data
",round(metrics.roc_auc_score(y_train ,predbow_train[:,1]),3))
```

AUC score for Logistic regression, with TFIDF for Testing data is 0.69

AUC score for Logistic regression with TFIDF for Training data 0.714

Avg W2V

```
In [74]:
```

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

```
In [75]:
```

```
# average Word2Vec
# compute average word2vec for each review.
train w2v vectors essays = []; # the avg-w2v for each essay is stored in this list
for sentence in tqdm(X_train['clean_essays'].values): # for each essay in training data
   vector = np.zeros(50) # as word vectors are of zero length
    cnt words =0; # num of words with a valid vector in the essay
    for word in sentence.split(): # for each word in a essay
       if word in glove words:
           vector += model[word][:50]
           cnt_words += 1
    if cnt words != 0:
       vector /= cnt words
    train_w2v_vectors_essays.append(vector)
print("train vector")
print(len(train w2v vectors essays))
print(len(train_w2v_vectors_essays[0]))
print('='*50)
# average Word2Vec
# compute average word2vec for each review.
test w2v vectors essays = []; # the avg-w2v for each essay is stored in this list
for sentence in tqdm(X test['clean essays'].values): # for each essay in training data
    vector = np.zeros(50) # as word vectors are of zero length
   cnt words =0; # num of words with a valid vector in the essay
    for word in sentence.split(): # for each word in a essay
       if word in glove words:
           vector += model[word][:50]
           cnt words += 1
    if cnt words != 0:
       vector /= cnt words
    test w2v vectors essays.append(vector)
print("Test vec")
print(len(test w2v vectors essays))
print(len(test w2v vectors essays[0]))
print('='*50)
# average Word2Vec
# compute average word2vec for each review.
cv_w2v_vectors_essays = []; # the avg-w2v for each essay is stored in this list
for sentence in tqdm(X cv['clean essays'].values): # for each essay in training data
   vector = np.zeros(50) # as word vectors are of zero length
    cnt_words =0; # num of words with a valid vector in the essay
    for word in sentence.split(): # for each word in a essay
        if word in glove words:
           vector += model[word][:50]
           cnt words += 1
    if cnt words != 0:
       vector /= cnt words
    cv w2v vectors essays.append(vector)
print("CV vec")
print(len(cv_w2v_vectors_essays))
print(len(cv_w2v_vectors_essays[0]))
print('='*50)
100%|
                                                                         29870/29870
[00:14<00:00, 2090.73it/s]
```

```
29870
```

```
| 21959/21959
[00:11<00:00, 1915.11it/s]
Test vec
21959
50
_____
```

```
14713/14713
[00:07<00:00, 1888.60it/s]
CV vec
14713
50
```

In [76]:

```
# Changing list to numpy arrays
train w2v vectors essays = np.array(train w2v vectors essays)
test_w2v_vectors_essays = np.array(test_w2v_vectors_essays)
cv w2v vectors_essays = np.array(cv_w2v_vectors_essays)
```

```
In [77]:
# average Word2Vec
# compute average word2vec for each review.
train w2v vectors titles = []; # the avg-w2v for each essay is stored in this list
for sentence in tqdm(X train['clean titles'].values): # for each essay in training data
   vector = np.zeros(50) # as word vectors are of zero length
   cnt words =0; # num of words with a valid vector in the essay
    for word in sentence.split(): # for each word in a essay
        if word in glove words:
           vector += model[word][:50]
           cnt words += 1
    if cnt words != 0:
       vector /= cnt_words
    train w2v vectors titles.append(vector)
print("train vector")
print(len(train w2v vectors titles))
print(len(train w2v vectors titles[0]))
print('='*50)
# average Word2Vec
# compute average word2vec for each review.
test w2v vectors titles = []; # the avq-w2v for each essay is stored in this list
for sentence in tqdm(X test['clean titles'].values): # for each essay in training data
   vector = np.zeros(50) # as word vectors are of zero length
    cnt words =0; # num of words with a valid vector in the essay
    for word in sentence.split(): # for each word in a essay
        if word in glove words:
           vector += model[word][:50]
           cnt words += 1
    if cnt words != 0:
        vector /= cnt words
    test w2v vectors titles.append(vector)
print("Test vec")
print(len(test_w2v_vectors_titles))
print(len(test w2v vectors titles[0]))
print('='*50)
# average Word2Vec
# compute average word2vec for each review.
cv_w2v_vectors_titles = []; # the avg-w2v for each essay is stored in this list
for sentence in tqdm(X cv['clean titles'].values): # for each essay in training data
   vector = np.zeros(50) # as word vectors are of zero length
    cnt words =0; # num of words with a valid vector in the essay
    for word in sentence.split(): # for each word in a essay
```

```
if word in glove words:
           vector += model[word][:50]
           cnt words += 1
    if cnt words != 0:
       vector /= cnt words
    cv w2v vectors titles.append(vector)
print("CV vec")
print(len(cv w2v vectors titles))
print(len(cv_w2v_vectors_titles[0]))
print('='*50)
100%|
                                                                       29870/29870
[00:00<00:00, 41829.58it/s]
train vector
29870
50
                                                                    21959/21959
100%|
[00:00<00:00, 34189.12it/s]
Test vec
21959
______
100%|
                                                                            | 14713/14713
[00:00<00:00, 28589.76it/s]
CV vec
14713
50
In [78]:
# Changing list to numpy arrays
train w2v vectors titles = np.array(train w2v vectors titles)
test_w2v_vectors_titles = np.array(test_w2v_vectors_titles)
cv w2v vectors titles = np.array(cv w2v vectors titles)
In [79]:
# average Word2Vec
# compute average word2vec for each review.
train_w2v_vectors_summary = []; # the avg-w2v for each essay is stored in this list
for sentence in tqdm(X train['project resource summary'].values): # for each essay in training
data
   vector = np.zeros(50) # as word vectors are of zero length
   cnt words =0; # num of words with a valid vector in the essay
    for word in sentence.split(): # for each word in a essay
       if word in glove words:
           vector += model[word][:50]
           cnt words += 1
    if cnt words != 0:
       vector /= cnt_words
    \verb|train_w2v_vectors_summary.append(vector)|
print("train vector")
print(len(train w2v vectors summary))
print(len(train w2v vectors summary[0]))
print('='*50)
# average Word2Vec
# compute average word2vec for each review.
test w2v vectors summary = []; # the avg-w2v for each essay is stored in this list
for sentence in tqdm(X_test['project_resource summary'].values): # for each essay in training data
   vector = np.zeros(50) # as word vectors are of zero length
    cnt words =0; # num of words with a valid vector in the essay
    for word in sentence.split(): # for each word in a essay
       if word in alove words.
```

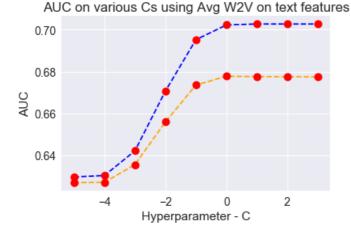
```
** WOIG *** GIOVE WOIGS.
                       vector += model[word][:50]
                      cnt words += 1
        if cnt words != 0:
             vector /= cnt_words
        test w2v vectors summary.append(vector)
print("Test vec")
print(len(test w2v vectors summary))
print(len(test_w2v_vectors_summary[0]))
print('='*50)
 # average Word2Vec
 # compute average word2vec for each review.
cv w2v vectors summary = []; # the avg-w2v for each essay is stored in this list
for sentence in tqdm(X_cv['project_resource_summary'].values): # for each essay in training data
       vector = np.zeros(50) # as word vectors are of zero length
        cnt words =0; # num of words with a valid vector in the essay
        for word in sentence.split(): # for each word in a essay
               if word in glove words:
                       vector += model[word][:50]
                       cnt_words += 1
        if cnt words != 0:
               vector /= cnt words
        cv w2v vectors summary.append(vector)
print("CV vec")
print(len(cv w2v vectors summary))
print(len(cv w2v vectors summary[0]))
print('='*50)
100%|
                                                                                                                                                       | 29870/29870
[00:02<00:00, 13227.38it/s]
train vector
29870
50
______
                                                                                                                                                     | 21959/21959
[00:01<00:00, 11110.13it/s]
Test vec
21959
______
100%|
                                                                                                                                             14713/14713
[00:01<00:00, 10963.98it/s]
CV vec
14713
50
______
In [80]:
# Changing list to numpy arrays
train w2v vectors summary = np.array(train w2v vectors summary)
test_w2v_vectors_summary = np.array(test_w2v_vectors_summary)
cv_w2v_vectors_summary = np.array(cv_w2v_vectors_summary)
In [81]:
 # merge two sparse matrices: https://stackoverflow.com/a/19710648/4084039
from scipy.sparse import hstack
X tr =
\verb|hstack|((train_w2v_vectors_essays, train_w2v_vectors_titles, train_w2v_vectors_summary, X_train_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_clean_cl
at ohe, X train clean subcat ohe, X train state ohe, X train teacher ohe, X train grade ohe,
X_train_price_std, X_train_projects_std, X_train_qty_std)).tocsr()
X cr =
hstack((cv_w2v_vectors_essays,cv_w2v_vectors_titles,cv_w2v_vectors_summary,X_cv_clean_cat_ohe,X_cv
```

```
clean subcat one, X cv state one, X cv teacher one, X cv grade one, X cv price std, X cv projects s
td, X_cv_qty_std)).tocsr()
X te =
hstack((test w2v vectors essays,test w2v vectors titles,test w2v vectors summary,X test clean cat c
he, X_test_clean_subcat_ohe, X_test_state_ohe, X_test_teacher_ohe, X_test_grade_ohe,
X test price std, X test projects std, X test qty std)).tocsr()
print("Final Data matrix")
print(X_tr.shape, y_train.shape)
print(X_cr.shape, y_cv.shape)
print(X_te.shape, y_test.shape)
print("="*100)
4
Final Data matrix
(29870, 253) (29870,)
(14713, 253) (14713,)
(21959, 253) (21959,)
                                                                                                  - 100 €
In [821:
Cs, auc_tr, auc_cv=predAUC(X_tr,y_train)
plt title1 = 'Avg W2V'
```

```
plotPerformance(Cs, auc tr, auc cv, plt title1)
```

LogisticRegression(C=1, class weight='balanced', dual=False, fit_intercept=True, intercept_scaling=1, max_iter=100, multi_class='ovr', n_jobs=1, penalty='12', random_state=None, solver='liblinear', tol=0.0001, verbose=0, warm start=False) 0.6805203639250168

No handles with labels found to put in legend.



In [83]:

```
optimal c=0.1
LOR=Get LOR OptimalC(optimal c)
```

In [84]:

```
roc data test1, roc data train1 = getROC Data(X te, y test, X tr, y train, LOR)
```

In [85]:

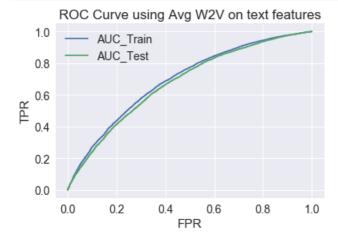
0.67773913

```
print(auc tr,auc cv)
[0.62968315 0.63037755 0.64213868 0.67067409 0.69516801 0.70220467
          0.70267509 \ 0.70266767 \ 0.70266603] \ [0.62705544 \ 0.62716149 \ 0.63557449 \ 0.65595911 \ 0.67352891 \ 0.67352891] \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \ 0.67352891 \
```

0.6775784 0.67747909 0.67746424]

In [86]:

makeROC(roc_data_test1, roc_data_train1, plt_title1)



In [87]:

get_confusion_matrix(LOR,X_tr,y_train)



In [88]:

%%time
get_confusion_matrix(LOR, X_te, y_test)

Wall time: 181 ms



In [89]:

predbow = (LOR.predict_proba(X_te))
predbow_train=(LOR.predict_proba(X_tr))

```
111 [JU].
```

```
print("AUC score for Logistic regression AVG W2V for Testing data is
",round(metrics.roc_auc_score(y_test ,predbow[:,1]),3))
print("AUC score for Logistic regression AVG W2V for Training data ",round(metrics.roc_auc_score(y_train ,predbow_train[:,1]),3))
```

AUC score for Logistic regression AVG W2V for Testing data is 0.678 AUC score for Logistic regression AVG W2V for Training data 0.694

TFIDF W2V

In [91]:

```
\# S = ["abc def pqr", "def def def abc", "pqr pqr def"]
tfidf model = TfidfVectorizer()
tfidf model.fit(X train['clean essays'].values)
# we are converting a dictionary with word as a key, and the idf as a value
dictionary = dict(zip(tfidf model.get feature names(), list(tfidf model.idf )))
tfidf words = set(tfidf model.get feature names())
# average Word2Vec
# compute average word2vec for each review.
train tfidf w2v essays = []; # the avg-w2v for each sentence/review is stored in this list
for sentence in tqdm(X_train['clean_essays'].values): # for each review/sentence
   vector = np.zeros(50) # 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][:50] # getting the vector for each word
            # here we are multiplying idf value(dictionary[word]) and the tf
value((sentence.count(word)/len(sentence.split())))
            tf idf = dictionary[word]*(sentence.count(word)/len(sentence.split())) # getting the tf
idf value for each word
            vector += (vec * tf idf) # calculating tfidf weighted w2v
            tf idf weight += tf idf
    if tf idf weight != 0:
        vector /= tf idf weight
    train tfidf w2v essays.append(vector)
print("Train matrix:")
print(len(train_tfidf_w2v_essays))
print(len(train tfidf w2v essays[0]))
print('='*50)
cv tfidf w2v essays = []; # the avg-w2v for each sentence/review is stored in this list
for sentence in tqdm(X cv['clean essays'].values): # for each review/sentence
    vector = np.zeros(50) # 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][:50] # getting the vector for each word
            # here we are multiplying idf value(dictionary[word]) and the tf
value((sentence.count(word)/len(sentence.split())))
            tf idf = dictionary[word]*(sentence.count(word)/len(sentence.split()))  # getting the tf
idf value for each word
           vector += (vec * tf idf) # calculating tfidf weighted w2v
            tf idf weight += tf idf
    if tf_idf_weight != 0:
        vector /= tf idf weight
    cv tfidf w2v essays.append(vector)
print("CV matrix:")
print(len(cv_tfidf_w2v_essays))
print(len(cv_tfidf_w2v_essays[0]))
print('='*50)
test tfidf w2v essays = []; # the avg-w2v for each sentence/review is stored in this list
for sentence in tqdm(X_test['clean_essays'].values): # for each review/sentence
    vector = np.zeros(50) # 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][:50] # 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 essays.append(vector)
print("Test matrix:")
print(len(test tfidf w2v essays))
print(len(test_tfidf_w2v_essays[0]))
print('='*50)
100%|
                                                                       | 29870/29870 [02:
22<00:00, 209.68it/s]
Train matrix:
29870
50
______
100%|
                                                                       | 14713/14713 [01:
08<00:00, 214.90it/s]
CV matrix:
14713
50
_____
100%|
                                                                             | 21959/21959 [01:
46<00:00, 206.74it/s]
Test matrix:
21959
50
In [92]:
# Changing list to numpy arrays
train_tfidf_w2v_essays = np.array(train_tfidf_w2v_essays)
test_tfidf_w2v_essays = np.array(test_tfidf_w2v_essays)
cv_tfidf_w2v_essays = np.array(cv_tfidf_w2v_essays)
In [93]:
tfidf model = TfidfVectorizer()
tfidf model.fit(X train['clean titles'].values)
# we are converting a dictionary with word as a key, and the idf as a value
dictionary = dict(zip(tfidf model.qet feature names(), list(tfidf model.idf )))
tfidf words = set(tfidf model.get feature names())
# average Word2Vec
# compute average word2vec for each review.
train tfidf w2v titles = []; # the avg-w2v for each sentence/review is stored in this list
for sentence in tqdm(X_train['clean_titles'].values): # for each review/sentence
    vector = np.zeros(50) # 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][:50] # 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 /- tr rar wergiit
    train tfidf_w2v_titles.append(vector)
print("Train matrix:")
print(len(train_tfidf_w2v_titles))
print(len(train_tfidf_w2v_titles[0]))
print('='*50)
cv tfidf w2v titles = []; # the avg-w2v for each sentence/review is stored in this list
for sentence in tqdm(X cv['clean titles'].values): # for each review/sentence
    vector = np.zeros(50) # 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][:50] # getting the vector for each word
            # here we are multiplying idf value(dictionary[word]) and the tf
value((sentence.count(word)/len(sentence.split())))
           tf idf = dictionary[word]*(sentence.count(word)/len(sentence.split())) # getting the tf
idf value for each word
           vector += (vec * tf idf) # calculating tfidf weighted w2v
            tf idf weight += tf idf
    if tf_idf_weight != 0:
       vector /= tf idf weight
    cv tfidf w2v titles.append(vector)
print("CV matrix:")
print(len(cv tfidf w2v titles))
print(len(cv_tfidf_w2v_titles[0]))
print('='*50)
test tfidf w2v titles = []; # the avg-w2v for each sentence/review is stored in this list
for sentence in tqdm(X_test['clean_titles'].values): # for each review/sentence
    vector = np.zeros(50) # 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][:50] # 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 titles.append(vector)
print("Test matrix:")
print(len(test_tfidf_w2v_titles))
print(len(test_tfidf_w2v_titles[0]))
print('='*50)
100%|
                                                                        29870/29870
[00:02<00:00, 12704.98it/s]
Train matrix:
29870
50
_____
100%|
                                                                           | 14713/14713
[00:01<00:00, 12599.60it/s]
CV matrix:
14713
100%|
                                                                      | 21959/21959
[00:01<00:00, 13612.35it/s]
Test matrix:
21959
```

50

In [94]:

```
# Changing list to numpy arrays
train_tfidf_w2v_titles = np.array(train_tfidf_w2v_titles)
test_tfidf_w2v_titles = np.array(test_tfidf_w2v_titles)
cv_tfidf_w2v_titles = np.array(cv_tfidf_w2v_titles)
```

In [95]:

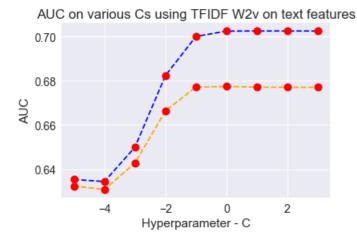
```
tfidf model = TfidfVectorizer()
tfidf model.fit(X_train['project_resource_summary'].values)
\# we are converting a dictionary with word as a key, and the idf as a value
dictionary = dict(zip(tfidf model.get feature names(), list(tfidf model.idf )))
tfidf words = set(tfidf model.get feature names())
# average Word2Vec
# compute average word2vec for each review.
train tfidf w2v summary = []; # the avg-w2v for each sentence/review is stored in this list
for sentence in tqdm (X train['project resource summary'].values): # for each review/sentence
    vector = np.zeros(50) # 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][:50] # getting the vector for each word
            # here we are multiplying idf value(dictionary[word]) and the tf
value((sentence.count(word)/len(sentence.split())))
            tf idf = dictionary[word]*(sentence.count(word)/len(sentence.split())) # getting the tf
idf value for each word
            vector += (vec * tf idf) # calculating tfidf weighted w2v
            tf_idf_weight += tf_idf
    if tf_idf_weight != 0:
        vector /= tf idf weight
    train_tfidf_w2v_summary.append(vector)
print("Train matrix:")
print(len(train_tfidf_w2v_summary))
print(len(train tfidf w2v summary[0]))
print('='*50)
cv tfidf w2v summary = []; # the avg-w2v for each sentence/review is stored in this list
for sentence in tqdm(X_cv['project_resource_summary'].values): # for each review/sentence
    vector = np.zeros(50) # 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][:50] # getting the vector for each word
            \# here we are multiplying idf value(dictionary[word]) and the tf
value((sentence.count(word)/len(sentence.split())))
            tf_idf = dictionary[word]*(sentence.count(word)/len(sentence.split())) # getting the tf
idf value for each word
            vector += (vec * tf idf) # calculating tfidf weighted w2v
            tf idf weight += tf idf
    if tf idf weight != 0:
        vector /= tf idf weight
    cv tfidf w2v summary.append(vector)
print("CV matrix:")
print(len(cv_tfidf_w2v_summary))
print(len(cv tfidf w2v summary[0]))
print('='*50)
test tfidf w2v summary = []; # the avg-w2v for each sentence/review is stored in this list
for sentence in tqdm(X_test['project_resource_summary'].values): # for each review/sentence
    vector = np.zeros(50) # 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][:50] # 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 summary.append(vector)
print("Test matrix:")
print(len(test_tfidf_w2v_summary))
print(len(test_tfidf_w2v_summary[0]))
print('='*50)
100%|
                                                                         29870/29870
[00:07<00:00, 3834.02it/s]
Train matrix:
29870
50
_____
                                                                              | 14713/14713
100%1
[00:04<00:00, 3375.79it/s]
CV matrix:
14713
50
100%|
                                                                             | 21959/21959
[00:06<00:00, 3164.53it/s]
Test matrix:
21959
50
In [96]:
# Changing list to numpy arrays
train_tfidf_w2v_summary = np.array(train_tfidf_w2v_summary)
test_tfidf_w2v_summary = np.array(test_tfidf_w2v_summary)
cv tfidf w2v summary = np.array(cv tfidf w2v summary)
In [97]:
# merge two sparse matrices: https://stackoverflow.com/a/19710648/4084039
from scipy.sparse import hstack
X tr =
\verb|hstack|((train_tfidf_w2v_essays, train_tfidf_w2v_titles, train_tfidf_w2v_summary, X_train_clean_cat_ohermore)| \\
,X_train_clean_subcat_ohe, X_train_state_ohe, X_train_teacher_ohe, X_train_grade_ohe,
X_train_price_std,X_train_projects_std,X_train_qty_std)).tocsr()
hstack((cv_tfidf_w2v_essays,cv_tfidf_w2v_titles,cv_tfidf_w2v_summary,X_cv_clean_cat_ohe,X_cv_clean_
subcat_ohe, X_cv_state_ohe, X_cv_teacher_ohe, X_cv_grade_ohe, X_cv_price_std, X_cv_projects_std, X_cv
_qty_std)).tocsr()
X_te =
hstack((test_tfidf_w2v_essays,test_tfidf_w2v_titles,test_tfidf_w2v_summary,X_test_clean_cat_ohe,X_
test clean subcat ohe, X test state ohe, X test teacher ohe, X test grade ohe, X test price std,X t
est_projects_std, X_test_qty_std)).tocsr()
print("Final Data matrix")
print(X_tr.shape, y_train.shape)
print(X cr.shape, y cv.shape)
print(X_te.shape, y_test.shape)
print("="*100)
4
Final Data matrix
(29870, 253) (29870,)
(14713, 253) (14713,)
(21959, 253) (21959,)
```

In [98]:

```
Cs, auc_tr, auc_cv=predAUC(X_tr,y_train)
plt_title1 = 'TFIDF W2v'
plotPerformance(Cs, auc_tr, auc_cv, plt_title1)
```

No handles with labels found to put in legend.



In [99]:

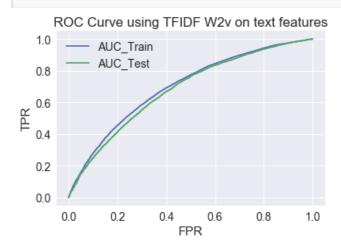
```
optimal_c=0.1
LOR=Get_LOR_OptimalC(optimal_c)
```

In [100]:

```
roc_data_test1, roc_data_train1 = getROC_Data(X_te, y_test, X_tr, y_train, LOR)
```

In [101]:

makeROC(roc_data_test1, roc_data_train1, plt_title1)



In [102]:

```
get confusion matrix(LOR,X tr,y train)
```



In [103]:

```
%%time
get_confusion_matrix(LOR, X_te, y_test)
```

Wall time: 166 ms



In [104]:

```
predbow = (LOR.predict_proba(X_te))
predbow_train=(LOR.predict_proba(X_tr))
```

In [105]:

```
print("AUC score for Logistic regression with TFIDF-W2V for Testing data is ",round(metrics.roc_a
uc_score(y_test ,predbow[:,1]),3))
print("AUC score for Logistic regression TFIDF-W2V for Training data
",round(metrics.roc_auc_score(y_train ,predbow_train[:,1]),3))
```

AUC score for Logistic regression with TFIDF-W2V for Testing data is 0.68 AUC score for Logistic regression TFIDF-W2V for Training data 0.697

CONCLUSION

In [107]:

```
#http://zetcode.com/python/prettytable/
from prettytable import PrettyTable

x = PrettyTable()
x.field_names = ["Vectorizer", "Hyperparameter", "Test AUC", "Train AUC"]
x.add_row(["Bag of Words", 0.01, 0.725,0.836])
x.add_row(["TFIDF",0.01,0.69, 0.714])
x.add_row(["AVG W2V",0.1, 0.678,0.694])
x.add_row(["TFIDF W2V", 0.1, 0.68,0.697])
print(x)
```

Bag of Words 0.01 0.725 0.836		Hyperparameter		
0.01	g of Words	0.01	0.725	0.836
TFIDF 0.01 0.69 0.714 AVG W2V 0.1 0.678 0.694				•
TFIDF W2V 0.1 0.68 0.697	- '			