DonorsChoose

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

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

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

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

About the DonorsChoose Data Set

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

Feature	Description		
project_id	A unique identifier for the proposed project. Example		
project_title	Title of the project. Examples: • Art Will Make You Happy! • First Grade Fun		
project_grade_category	Grade level of students for which the project is targets enumerated values: • Grades PreK-2 • Grades 3-5 • Grades 6-8 • Grades 9-12		
project_subject_categories	One or more (comma-separated) subject categories for following enumerated list of values: • Applied Learning • Care & Hunger • Health & Sports • History & Civics • Literacy & Language • Math & Science • Music & The Arts • Special Needs • Warmth Examples: • Music & The Arts • Literacy & Language, Math & Science		
school_state	State where school is located (<u>Two-letter U.S. postal</u> (<u>https://en.wikipedia.org/wiki/List_of_U.Sstate_abbrexample:</u> WY		
project_subject_subcategories	One or more (comma-separated) subject subcategori Examples: • Literacy • Literature & Writing, Social Sciences		
project_resource_summary	An explanation of the resources needed for the project • My students need hands on literacy mater sensory needs!		

Feature	Description		
project_essay_1	First application essay*		
project_essay_2	Second application essay [*]		
project_essay_3	Third application essay*		
project_essay_4	Fourth application essay*		
project_submitted_datetime	Datetime when project application was submitted. Ex 12:43:56.245		
teacher_id	A unique identifier for the teacher of the proposed pro bdf8baa8fedef6bfeec7ae4ff1c15c56		
teacher_prefix	Teacher's title. One of the following enumerated value • nan • Dr. • Mr. • Mrs. • Ms. • Teacher.		
teacher_number_of_previously_posted_projects	Number of project applications previously submitted Example: 2		

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

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

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

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

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

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

Notes on the Essay Data

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

- __project_essay_1:__ "Introduce us to your classroom"
- __project_essay_2:__ "Tell us more about your students"
- project essay 3: "Describe how your students will use the materials you're requesting"
- __project_essay_3:__ "Close by sharing why your project will make a difference"

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

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

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

In [1]:

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

In [2]:

```
df1 = pd.read_csv('train_data.csv',nrows=80000)
df2 =pd.read csv('resources.csv', nrows =80000)
```

In [3]:

```
print("Number of data points in train data", df1.shape)
print('-'*50)
print("The attributes of data :", df1.columns.values)
Number of data points in train data (80000, 17)
The attributes of data : ['Unnamed: 0' 'id' 'teacher_id' 'teacher_prefix'
'school_state'
 'project_submitted_datetime' 'project_grade_category'
 'project_subject_categories' 'project_subject_subcategories'
 'project_title' 'project_essay_1' 'project_essay_2' 'project_essay_3'
 'project_essay_4' 'project_resource_summary'
 'teacher_number_of_previously_posted_projects' 'project_is_approved']
In [4]:
print("Number of data points in train data", df2.shape)
print(df2.columns.values)
print(df2.head(2))
Number of data points in train data (80000, 4)
['id' 'description' 'quantity' 'price']
        id
                                                  description quantity \
0 p233245 LC652 - Lakeshore Double-Space Mobile Drying Rack
                                                                       1
                  Bouncy Bands for Desks (Blue support pipes)
1 p069063
                                                                      3
   price
0
 149.00
   14.95
```

```
In [5]:
```

```
df1.head()
```

Out[5]:

	Unnamed:	id	teacher_id	teacher_prefix	school_s
0	160221	p253737	c90749f5d961ff158d4b4d1e7dc665fc	Mrs.	IN
1	140945	p258326	897464ce9ddc600bced1151f324dd63a	Mr.	FL
2	21895	p182444	3465aaf82da834c0582ebd0ef8040ca0	Ms.	AZ
3	45	p246581	f3cb9bffbba169bef1a77b243e620b60	Mrs.	KY
4	172407	p104768	be1f7507a41f8479dc06f047086a39ec	Mrs.	TX

preprocessing of project subject categories

```
In [7]:
```

```
categories = list(df1['project subject categories'].values)
# remove special characters from list of strings python: https://stackoverflow.com/a/47
301924/4084039
# https://www.geeksforgeeks.org/removing-stop-words-nltk-python/
# https://stackoverflow.com/questions/23669024/how-to-strip-a-specific-word-from-a-stri
# https://stackoverflow.com/questions/8270092/remove-all-whitespace-in-a-string-in-pyth
cat list = []
for i in categories:
    temp = ""
    # consider we have text like this "Math & Science, Warmth, Care & Hunger"
    for j in i.split(','): # it will split it in three parts ["Math & Science", "Warmt
h", "Care & Hunger"]
        if 'The' in j.split(): # this will split each of the catogory based on space "M
ath & Science"=> "Math", "&", "Science"
            j=j.replace('The','') # if we have the words "The" we are going to replace
 it with ''(i.e removing 'The')
        j = j.replace(' ','') # we are placeing all the ' '(space) with ''(empty) ex:"M
ath & Science"=>"Math&Science"
        temp+=j.strip()+" " #" abc ".strip() will return "abc", remove the trailing spa
ces
        temp = temp.replace('&','_') # we are replacing the & value into
    cat list.append(temp.strip())
df1['clean_categories'] = cat_list
df1.drop(['project_subject_categories'], axis=1, inplace=True)
from collections import Counter
my_counter = Counter()
for word in df1['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]))
sorted cat dict
Out[7]:
{'Warmth': 1029,
 'Care Hunger': 1029,
 'History_Civics': 4319,
 'Music_Arts': 7540,
 'AppliedLearning': 8917,
 'SpecialNeeds': 9974,
 'Health_Sports': 10445,
 'Math Science': 30285,
 'Literacy Language': 38266}
```

preprocessing project subject subcategories

In [8]:

```
sub catogories = list(df1['project subject subcategories'].values)
# remove special characters from list of strings python: https://stackoverflow.com/a/47
301924/4084039
# https://www.geeksforgeeks.org/removing-stop-words-nltk-python/
# https://stackoverflow.com/questions/23669024/how-to-strip-a-specific-word-from-a-stri
# https://stackoverflow.com/questions/8270092/remove-all-whitespace-in-a-string-in-pyth
on
sub_cat_list = []
for i in sub catogories:
    temp = ""
    # consider we have text like this "Math & Science, Warmth, Care & Hunger"
    for j in i.split(','): # it will split it in three parts ["Math & Science", "Warmt
h", "Care & Hunger"]
        if 'The' in j.split(): # this will split each of the catogory based on space "M
ath & Science"=> "Math", "&", "Science"
            j=j.replace('The','') # if we have the words "The" we are going to replace
 it with ''(i.e removing 'The')
        j = j.replace(' ','') # we are placeing all the ' '(space) with ''(empty) ex:"M
ath & Science"=>"Math&Science"
        temp +=j.strip()+" "#" abc ".strip() will return "abc", remove the trailing spa
ces
        temp = temp.replace('&','_')
    sub_cat_list.append(temp.strip())
df1['clean subcategories'] = sub cat list
df1.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 df1['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]))
sorted sub cat dict
```

```
Out[8]:
{'Economics': 198,
 'CommunityService': 332,
 'FinancialLiteracy': 417,
 'ParentInvolvement': 496,
 'Civics_Government': 591,
 'Extracurricular': 607,
 'ForeignLanguages': 623,
 'NutritionEducation': 995,
 'Warmth': 1029,
 'Care_Hunger': 1029,
 'SocialSciences': 1381,
 'PerformingArts': 1439,
 'CharacterEducation': 1499,
 'TeamSports': 1596,
 'Other': 1759,
 'College_CareerPrep': 1865,
 'History_Geography': 2310,
 'Music': 2321,
 'Health LifeScience': 3059,
 'EarlyDevelopment': 3130,
 'ESL': 3169,
 'Gym_Fitness': 3337,
 'EnvironmentalScience': 4060,
 'VisualArts': 4575,
 'Health Wellness': 7509,
 'AppliedSciences': 7904,
 'SpecialNeeds': 9974,
 'Literature_Writing': 16239,
 'Mathematics': 20552,
 'Literacy': 24757}
In [9]:
df1['teacher_prefix']= df1['teacher_prefix'].fillna(df1['teacher_prefix'].mode().iloc[0
])
prefix = list(df1['teacher prefix'].values)
# remove special characters from list of strings python: https://stackoverflow.com/a/47
301924/4084039
# https://www.geeksforgeeks.org/removing-stop-words-nltk-python/
# https://stackoverflow.com/questions/23669024/how-to-strip-a-specific-word-from-a-stri
# https://stackoverflow.com/questions/8270092/remove-all-whitespace-in-a-string-in-pyth
on
prefix_list = []
for i in prefix:
    temp = ""
    if "." in i:
        i = i.replace('.','')
        temp +=i.strip()+" "#" abc ".strip() will return "abc", remove the trailing spa
ces
    prefix list.append(temp.strip())
df1['teachers prefix'] = prefix list
df1.drop(['teacher_prefix'],inplace = True,axis = 1)
```

In [10]:

```
# count of all the words in corpus python: https://stackoverflow.com/a/22898595/4084039
my_counter = Counter()
for word in df1['teachers_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]))
sorted_prefix_dict

Out[10]:
{'Dr': 8, 'Mr': 7770, 'Ms': 28741, 'Mrs': 41776}
```

preprocessing project grade category

In [11]:

```
grade_categories = list(df1['project_grade_category'].values)
# remove special characters from list of strings python: https://stackoverflow.com/a/47
301924/4084039
# https://www.geeksforgeeks.org/removing-stop-words-nltk-python/
# https://stackoverflow.com/questions/23669024/how-to-strip-a-specific-word-from-a-stri
# https://stackoverflow.com/questions/8270092/remove-all-whitespace-in-a-string-in-pyth
on
grade_cat_list = []
for i in grade_categories:
    temp = ""
    if "-" in i:
        i = i.replace('-','_')
i = i.replace(' ','_')
        temp +=i.strip()+' '#" abc ".strip() will return "abc", remove the trailing spa
ces
    grade_cat_list.append(temp.strip())
df1['grade category'] = grade cat list
df1.drop(['project grade category'], axis=1, inplace=True)
```

In [12]:

```
# count of all the words in corpus python: https://stackoverflow.com/a/22898595/4084039
my_counter = Counter()
for word in df1['grade_category'].values:
    my_counter.update(word.split())

grade_dict = dict(my_counter)
sorted_grade_dict = dict(sorted(grade_dict.items(), key=lambda kv: kv[1]))

sorted_grade_dict

Out[12]:
```

```
{'Grades_9_12': 8021,
'Grades_6_8': 12383,
'Grades_3_5': 27244,
'Grades PreK 2': 32352}
```

train test splitting

```
In [13]:
```

```
y = df1['project_is_approved'].values
X = df1.drop(['Unnamed: 0','teacher_id','project_submitted_datetime'], axis=1)
X.head(1)
```

Out[13]:

	id	school_state	project_title	project_essay_1	project_essay_2	project_ess
0	p253737	IN	Educational Support for English Learners at Home	My students are English learners that are work	\"The limits of your language are the limits o	NaN
4						

In [14]:

```
from sklearn.model_selection import train_test_split

# split the data into test and train by maintaining same distribution of output varaible 
e 'y' [stratify=y]

X_train, X_test, y_train, y_test = train_test_split(X, y, stratify=y, test_size=0.2)

# split the train data into train and cross validation by maintaining same distribution 
of output varaible 'y_train' [stratify=y_train]

X_train, X_cv, y_train, y_cv = train_test_split(X_train, y_train, stratify=y_train, test_size=0.2)
```

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]:

```
sent = decontracted(df1['essay'].values[20000])
print(sent)
print("="*50)
```

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

In [17]:

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

My kindergarten students have varied disabilities ranging from speech and language delays, cognitive delays, gross/fine motor delays, to autism. The y are eager beavers and always strive to work their hardest working past t heir limitations. The materials we have are the ones I seek out for my students. I teach in a Title I school where most of the students receive f ree or reduced price lunch. Despite their disabilities and limitations, m y students love coming to school and come eager to learn and explore. Have you ever felt like you had ants in your pants and you needed to groove and move as you were in a meeting? This is how my kids feel all the time. The want to be able to move as they learn or so they say. Wobble chairs are the answer and I love then because they develop their core, which enhances gro ss motor and in Turn fine motor skills. They also want to learn through games, my kids do not want to sit and do worksheets. They want to learn to count by jumping and playing. Physical engagement is the key to our succes s. The number toss and color and shape mats can make that happen. My stude nts will forget they are doing work and just have the fun a 6 year old des erves.nannan

In [18]:

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

My kindergarten students have varied disabilities ranging from speech and language delays cognitive delays gross fine motor delays to autism They ar e eager beavers and always strive to work their hardest working past their limitations The materials we have are the ones I seek out for my students I teach in a Title I school where most of the students receive free or red uced price lunch Despite their disabilities and limitations my students lo ve coming to school and come eager to learn and explore Have you ever felt like you had ants in your pants and you needed to groove and move as you w ere in a meeting This is how my kids feel all the time The want to be able to move as they learn or so they say Wobble chairs are the answer and I lo ve then because they develop their core which enhances gross motor and in Turn fine motor skills They also want to learn through games my kids do no t want to sit and do worksheets They want to learn to count by jumping and playing Physical engagement is the key to our success The number toss and color and shape mats can make that happen My students will forget they are doing work and just have the fun a 6 year old deserves nannan

In [19]:

```
# https://gist.github.com/sebleier/554280
# we are removing the words from the stop words list: 'no', 'nor', 'not'
stopwords= ['i', 'me', 'my', 'myself', 'we', 'our', 'ours', 'ourselves', 'you', "you'r
e", "you've",\
            "you'll", "you'd", 'your', 'yours', 'yourself', 'yourselves', 'he', 'him',
'his', 'himself', \
            'she', "she's", 'her', 'hers', 'herself', 'it', "it's", 'its', 'itself', 't
hey', 'them', 'their',\
            'theirs', 'themselves', 'what', 'which', 'who', 'whom', 'this', 'that', "th
at'll", 'these', 'those', \
            'am', 'is', 'are', 'was', 'were', 'be', 'been', 'being', 'have', 'has', 'ha
d', 'having', 'do', 'does', \
             'did', 'doing', 'a', 'an', 'the', 'and', 'but', 'if', 'or', 'because', 'as'
, 'until', 'while', 'of', \
            'at', 'by', 'for', 'with', 'about', 'against', 'between', 'into', 'through'
 'during', 'before', 'after',\
            'above', 'below', 'to', 'from', 'up', 'down', 'in', 'out', 'on', 'off', 'ov
er', 'under', 'again', 'further',\
            'then', 'once', 'here', 'there', 'when', 'where', 'why', 'how', 'all', 'an
y', 'both', 'each', 'few', 'more', \
            'most', 'other', 'some', 'such', 'only', 'own', 'same', 'so', 'than', 'too'
, 'very', \
            's', 't', 'can', 'will', 'just', 'don', "don't", 'should', "should've", 'no
w', 'd', 'll', 'm', 'o', 're', \
            've', 'y', 'ain', 'aren', "aren't", 'couldn', "couldn't", 'didn', "didn't",
'doesn', "doesn't"
                  , 'hadn',\
            "hadn't", 'hasn', "hasn't", 'haven', "haven't", 'isn', "isn't", 'ma', 'migh
tn', "mightn't", 'mustn',\
            "mustn't", 'needn', "needn't", 'shan', "shan't", 'shouldn', "shouldn't", 'w
asn', "wasn't", 'weren', "weren't", \
            'won', "won't", 'wouldn', "wouldn't"]
```

In [20]:

```
# Combining all the above stundents
from tqdm import tqdm
preprocessed_train_essay = []
# 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('\\", ' ')
    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_train_essay.append(sent.lower().strip())
```

100%

| | 51200/51200 [00:55<00:00, 928.25it/s]

In [21]:

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

100%

| 12800/12800 [00:13<00:00, 939.27it/s]

In [22]:

```
# Combining all the above stundents
from tqdm import tqdm
preprocessed_test_essay = []
# 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('\\r', ' ')
    sent = sent.replace('\\n', ' ')
    sent = re.sub('[^A-Za-z0-9]+', ' ', sent)
    # https://gist.github.com/sebleier/554280
    sent = ' '.join(e for e in sent.split() if e.lower() not in stopwords)
    preprocessed_test_essay.append(sent.lower().strip())
```

100%

| 16000/16000 [00:16<00:00, 982.87it/s]

In [23]:

```
# Combining all the above stundents
from tqdm import tqdm
preprocessed_train_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('\\r', ' ')
    sent = sent.replace('\\r', ' ')
    sent = re.sub('[^A-Za-z0-9]+', ' ', sent)
    # https://gist.github.com/sebleier/554280
    sent = ' '.join(e for e in sent.split() if e.lower() not in stopwords)
    preprocessed_train_title.append(sent.lower().strip())
```

100%

| 51200/51200 [00:02<00:00, 20838.20it/s]

In [24]:

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

100%

| 12800/12800 [00:00<00:00, 20387.17it/s]

In [25]:

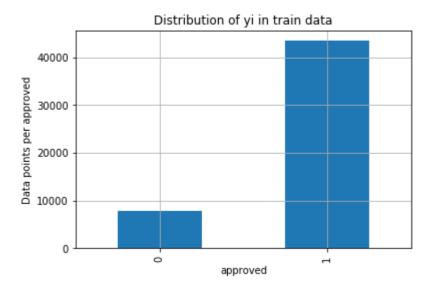
```
# Combining all the above stundents
from tqdm import tqdm
preprocessed_test_title = []
# tqdm is for printing the status bar
for sentance in tqdm(X_test['project_title'].values):
    sent = decontracted(sentance)
    sent = sent.replace('\\r', '')
    sent = sent.replace('\\r', '')
    sent = sent.replace('\\n',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

100%

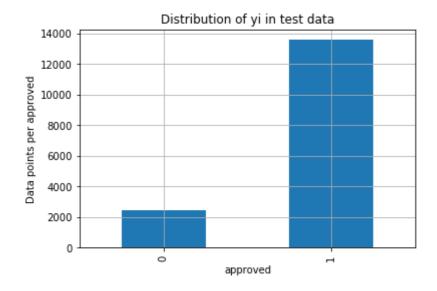
| 16000/16000 [00:00<00:00, 21510.82it/s]

In [26]:

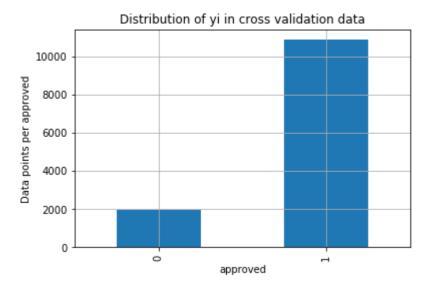
```
# it returns a dict, keys as aprvd labels and values as the number of data points in th
at aprvd
train aprvd distribution = X_train['project_is_approved'].value_counts().sort_index()
test aprvd distribution = X test['project is approved'].value counts().sort index()
cv_aprvd_distribution = X_cv['project_is_approved'].value_counts().sort_index()
my colors = 'rgbkymc'
train_aprvd_distribution.plot(kind='bar')
plt.xlabel('approved')
plt.ylabel('Data points per approved')
plt.title('Distribution of yi in train data')
plt.grid()
plt.show()
# ref: argsort https://docs.scipy.org/doc/numpy/reference/generated/numpy.argsort.html
# -(train aprvd distribution.values): the minus sign will give us in decreasing order
sorted_yi = np.argsort(-train_aprvd_distribution.values)
for i in sorted yi:
    print('Number of data points in approved', i+1, ':',train_aprvd_distribution.values
[i], '(', np.round((train_aprvd_distribution.values[i]/X_train.shape[0]*100), 3), '%)')
print('-'*80)
my_colors = 'rgbkymc'
test aprvd distribution.plot(kind='bar')
plt.xlabel('approved')
plt.ylabel('Data points per approved')
plt.title('Distribution of yi in test data')
plt.grid()
plt.show()
# ref: argsort https://docs.scipy.org/doc/numpy/reference/generated/numpy.argsort.html
# -(train_aprvd_distribution.values): the minus sign will give us in decreasing order
sorted yi = np.argsort(-test aprvd distribution.values)
for i in sorted yi:
    print('Number of data points in approved', i+1, ':',test_aprvd_distribution.values[
i], '(', np.round((test_aprvd_distribution.values[i]/X_test.shape[0]*100), 3), '%)')
print('-'*80)
my colors = 'rgbkymc'
cv aprvd distribution.plot(kind='bar')
plt.xlabel('approved')
plt.ylabel('Data points per approved')
plt.title('Distribution of yi in cross validation data')
plt.grid()
plt.show()
# ref: argsort https://docs.scipy.org/doc/numpy/reference/generated/numpy.argsort.html
# -(train aprvd distribution.values): the minus sign will give us in decreasing order
sorted_yi = np.argsort(-train_aprvd_distribution.values)
for i in sorted yi:
    print('Number of data points in approved', i+1, ':',cv aprvd distribution.values[i
], '(', np.round((cv aprvd distribution.values[i]/X cv.shape[0]*100), 3), '%)')
```



Number of data points in approved 2 : 43439 (84.842 %) Number of data points in approved 1 : 7761 (15.158 %)



Number of data points in approved 2 : 13575 (84.844 %)
Number of data points in approved 1 : 2425 (15.156 %)



Number of data points in approved 2 : 10860 (84.844 %) Number of data points in approved 1 : 1940 (15.156 %)

Response encoding

for school state

In [27]:

```
def get_project_fea_dict(alpha, feature, df):
    value_count = X_train[feature].value_counts()
    project_dict = dict()
    # denominator will contain the number of time that particular feature occured in wh
    for i, denominator in value_count.items():
        vec = []
        for k in range(0,2):
            aprvd_cnt = X_train.loc[(X_train['project_is_approved']==k) & (X_train[feat
ure]==i)]
            vec.append((aprvd cnt.shape[0] + alpha*10)/ (denominator + 90*alpha))
        project dict[i]=vec
    return project_dict
def get_project_feature(alpha, feature, df):
    project dict = get project fea dict(alpha, feature, df)
    value count = X train[feature].value counts()
    project fea = []
    # for every feature values in the given data frame we will check if it is there in
 the train data then we will add the feature to project_fea
    # if not we will add [1/2,1/2] to project_fea
    for index, row in df.iterrows():
        if row[feature] in dict(value count).keys():
            project_fea.append(project_dict[row[feature]])
        else:
            project_fea.append([1/2,1/2])
    return project_fea
```

In [28]:

```
# count of all the words in corpus python: https://stackoverflow.com/a/22898595/4084039
from collections import Counter
my_counter = Counter()
for word in df1['school_state'].values:
    my_counter.update(word.split())

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

Out[28]:

```
{'VT': 58,
 'WY': 79,
 'ND': 106,
 'MT': 168,
 'RI': 206,
 'SD': 221,
 'NE': 236,
 'NH': 237,
 'DE': 250,
 'AK': 256,
 'WV': 354,
 'HI': 369,
 'ME': 369,
 'DC': 382,
 'NM': 398,
 'KS': 460,
 'IA': 486,
 'ID': 501,
 'AR': 734,
 'CO': 858,
 'MN': 870,
 'OR': 904,
 'KY': 955,
 'MS': 955,
 'NV': 1016,
 'MD': 1087,
 'TN': 1202,
 'CT': 1235,
 'UT': 1270,
 'AL': 1273,
 'WI': 1331,
 'VA': 1513,
 'AZ': 1561,
 'NJ': 1625,
 'OK': 1710,
 'WA': 1715,
 'LA': 1764,
 'MA': 1765,
 'OH': 1819,
 'MO': 1896,
 'IN': 1897,
 'PA': 2237,
 'MI': 2341,
 'SC': 2881,
 'GA': 2908,
 'IL': 3178,
 'NC': 3737,
 'FL': 4568,
 'NY': 5391,
 'TX': 5406,
 'CA': 11262}
```

In [29]:

```
#response-coding of the feature
# alpha is used for laplace smoothing
alpha = 1
# train feature
train_state_responseCoding = np.array(get_project_feature(alpha, "school_state", X_trai
n))
# test feature
test_state_responseCoding = np.array(get_project_feature(alpha, "school_state", X_test
))
# cross validation feature
cv_state_responseCoding = np.array(get_project_feature(alpha, "school_state", X_cv))
```

In [30]:

```
print("Shape of X_train after response encodig ",train_state_responseCoding.shape, y_tr
ain.shape)
print("Shape of X_cv after response encodig ",cv_state_responseCoding.shape, y_cv.shape)
print("Shape of X_test after response encodig ",test_state_responseCoding.shape, y_test
.shape)
```

```
Shape of X_train after response encodig (51200, 2) (51200,)
Shape of X_cv after response encodig (12800, 2) (12800,)
Shape of X test after response encodig (16000, 2) (16000,)
```

for teacher_prefix

In [31]:

```
#response-coding of the feature
# alpha is used for laplace smoothing
alpha = 1
# train feature
train_teacher_responseCoding = np.array(get_project_feature(alpha, "teachers_prefix", X
_train))
# test feature
test_teacher_responseCoding = np.array(get_project_feature(alpha, "teachers_prefix", X_
test))
# cross validation feature
cv_teacher_responseCoding = np.array(get_project_feature(alpha, "teachers_prefix", X_cv
))
```

In [32]:

```
print("Shape of X_train after response encodig ",train_teacher_responseCoding.shape, y_
train.shape)
print("Shape of X_cv after response encodig ",cv_teacher_responseCoding.shape, y_cv.sha
pe)
print("Shape of X_test after response encodig ",test_teacher_responseCoding.shape, y_te
st.shape)
```

```
Shape of X_train after response encodig (51200, 2) (51200,) Shape of X_cv after response encodig (12800, 2) (12800,) Shape of X_test after response encodig (16000, 2) (16000,)
```

for project grade category

In [33]:

```
#response-coding of the feature
# alpha is used for laplace smoothing
alpha = 1
# train feature
train_grade_responseCoding = np.array(get_project_feature(alpha, "grade_category", X_tr
ain))
# test feature
test_grade_responseCoding = np.array(get_project_feature(alpha, "grade_category", X_tes
t))
# cross validation feature
cv_grade_responseCoding = np.array(get_project_feature(alpha, "grade_category", X_cv))
```

In [34]:

```
print("Shape of X_train after response encodig ",train_grade_responseCoding.shape, y_tr
ain.shape)
print("Shape of X_cv after response encodig ",cv_grade_responseCoding.shape, y_cv.shape)
print("Shape of X_test after response encodig ",test_grade_responseCoding.shape, y_test.shape)
```

```
Shape of X_train after response encodig (51200, 2) (51200,) Shape of X_cv after response encodig (12800, 2) (12800,) Shape of X test after response encodig (16000, 2) (16000,)
```

for project subject categories

In [35]:

```
#response-coding of the feature
# alpha is used for laplace smoothing
alpha = 1
# train feature
train_category_responseCoding = np.array(get_project_feature(alpha, "clean_categories",
X_train))
# test feature
test_category_responseCoding = np.array(get_project_feature(alpha, "clean_categories",
X_test))
# cross validation feature
cv_category_responseCoding = np.array(get_project_feature(alpha, "clean_categories", X_cv))
```

In [36]:

```
print("Shape of X_train after response encodig ",train_category_responseCoding.shape, y
_train.shape)
print("Shape of X_cv after response encodig ",cv_category_responseCoding.shape, y_cv.sh
ape)
print("Shape of X_test after response encodig ",test_category_responseCoding.shape, y_t
est.shape)
```

```
Shape of X_train after response encodig (51200, 2) (51200,)
Shape of X_cv after response encodig (12800, 2) (12800,)
Shape of X_test after response encodig (16000, 2) (16000,)
```

for project categories

In [37]:

```
#response-coding of the feature
# alpha is used for laplace smoothing
alpha = 1
# train feature
train_subcategory_responseCoding = np.array(get_project_feature(alpha, "clean_subcatego
ries", X_train))
# test feature
test_subcategory_responseCoding = np.array(get_project_feature(alpha, "clean_subcategor
ies", X_test))
# cross validation feature
cv_subcategory_responseCoding = np.array(get_project_feature(alpha, "clean_subcategories", X_cv))
```

In [38]:

```
print("Shape of X_train after response encodig ",train_subcategory_responseCoding.shape
, y_train.shape)
print("Shape of X_cv after response encodig ",cv_subcategory_responseCoding.shape, y_cv
.shape)
print("Shape of X_test after response encodig ",test_subcategory_responseCoding.shape,
y_test.shape)
```

```
Shape of X_train after response encodig (51200, 2) (51200,) Shape of X_cv after response encodig (12800, 2) (12800,) Shape of X test after response encodig (16000, 2) (16000,)
```

bag of words on essay

In [39]:

```
vec1 = CountVectorizer(min_df=10)
# fit has to happen only on train data

# we use the fitted CountVectorizer to convert the text to vector
X_train_essay_bow = vec1.fit_transform(preprocessed_train_essay)
X_cv_essay_bow = vec1.transform(preprocessed_cv_essay)
X_test_essay_bow = vec1.transform(preprocessed_test_essay)

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)
After vectorizations
```

```
After vectorizations (51200, 12247) (51200,) (12800, 12247) (12800,) (16000, 12247) (16000,)
```

bag of words on title

In [40]:

```
from sklearn.feature_extraction.text import TfidfVectorizer
vec2 = CountVectorizer(min_df=10)

# fit has to happen only on train data
# we use the fitted CountVectorizer to convert the text to vector
X_train_title_bow = vec2.fit_transform(preprocessed_train_title)
X_cv_title_bow = vec2.transform(preprocessed_cv_title)
X_test_title_bow = vec2.transform(preprocessed_test_title)

print("After vectorizations")
print(X_train_title_bow.shape, y_train.shape)
print(X_cv_title_bow.shape, y_cv.shape)
print(X_test_title_bow.shape, y_test.shape)
```

```
After vectorizations (51200, 2053) (51200,) (12800, 2053) (12800,) (16000, 2053) (16000,)
```

tfidf on essay

In [41]:

```
from sklearn.feature_extraction.text import TfidfVectorizer
vec3 = TfidfVectorizer(min_df=10, max_features =5000)

# fit has to happen only on train data
X_train_ess_tfidf = vec3.fit_transform(preprocessed_train_essay)

# we use the fitted Tfidf Vectorizer to convert the text to vector
X_cv_ess_tfidf = vec3.transform(preprocessed_cv_essay)
X_test_ess_tfidf = vec3.transform(preprocessed_test_essay)

print("After vectorizations")
print(X_train_ess_tfidf.shape, y_train.shape)
print(X_cv_ess_tfidf.shape, y_cv.shape)
print(X_test_ess_tfidf.shape, y_test.shape)
```

```
After vectorizations (51200, 5000) (51200,) (12800, 5000) (12800,) (16000, 5000) (16000,)
```

tfidf on title

In [42]:

```
vec4 = TfidfVectorizer(min_df=10)
X_train_title_tfidf = vec4.fit_transform(preprocessed_train_title)

# we use the fitted Tfidf Vectorizer to convert the text to vector
X_cv_title_tfidf = vec4.transform(preprocessed_cv_title)
X_test_title_tfidf = vec4.transform(preprocessed_test_title)

print("After vectorizations")
print(X_train_title_tfidf.shape, y_train.shape)
print(X_cv_title_tfidf.shape, y_cv.shape)
print(X_test_title_tfidf.shape, y_test.shape)
```

```
After vectorizations (51200, 2053) (51200,) (12800, 2053) (12800,) (16000, 2053) (16000,)
```

Numerical features

In [43]:

```
# https://stackoverflow.com/questions/22407798/how-to-reset-a-dataframes-indexes-for-al
l-groups-in-one-step
price_data = df2.groupby('id').agg({'price':'sum', 'quantity':'sum'}).reset_index()
price_data.head(2)
```

Out[43]:

	id	price	quantity
0	p000027	782.13	15
1	p000031	357.98	2

In [44]:

```
# join two dataframes in python:
df1 = pd.merge(df1, price_data, on='id', how='left')
```

In [45]:

```
# https://stackoverflow.com/questions/32617811/imputation-of-missing-values-for-categor
ies-in-pandas
#replacing nan with most frequently occurring element
df1['price'] = df1['price'].fillna(df1['price'].mode().iloc[0])
```

In [46]:

```
X_train =pd.merge(X_train,price_data, how ='left', on = 'id')
X_cv =pd.merge(X_cv,price_data, how ='left', on ='id')
X_test = pd.merge(X_test,price_data,how ='left',on='id')
```

In [47]:

```
# https://stackoverflow.com/questions/32617811/imputation-of-missing-values-for-categor
ies-in-pandas
#replacing nan with most frequently occuring element
X_train['price'] = X_train['price'].fillna(X_train['price'].mode().iloc[0])
X_cv['price'] = X_cv['price'].fillna(X_train['price'].mode().iloc[0])
X_test['price'] = X_test['price'].fillna(X_test['price'].mode().iloc[0])
print(X_train['price'].isnull().sum())
print(X_cv['price'].isnull().sum())
print(X_test['price'].isnull().sum())
```

0

0

0

```
In [48]:
```

```
# price
# check this one: https://www.youtube.com/watch?v=0HOqOcln3Z4&t=530s
# standardization sklearn: https://scikit-learn.org/stable/modules/generated/sklearn.pr
eprocessing.StandardScaler.html
from sklearn.preprocessing import StandardScaler
# price_standardized = standardScalar.fit(df1['price'].values)
# this will rise the error
# ValueError: Expected 2D array, got 1D array instead: array=[725.05 213.03 329.
        287.73
                 5.5 ].
# Reshape your data either using array.reshape(-1, 1)
price_scalar = StandardScaler()
X_train_price_stndrd = price_scalar.fit_transform(X_train['price'].values.reshape(-1,1)
))
X_cv_price_stndrd = price_scalar.transform(X_cv['price'].values.reshape(-1,1))
X_test_price_stndrd = price_scalar.transform(X_test['price'].values.reshape(-1,1))
In [49]:
```

```
X_train_price_stndrd
Out[49]:
```

In [50]:

```
X cv price stndrd
```

```
Out[50]:
```

```
In [51]:
```

```
X test price stndrd
Out[51]:
array([[-1.69162597],
       [-1.69162597],
       [-1.69162597],
       [-1.69162597],
       [-1.69162597],
       [-1.69162597]])
In [52]:
df1['quantity'] = df1['quantity'].fillna(df1['quantity'].mode().iloc[0])
In [53]:
# https://stackoverflow.com/questions/32617811/imputation-of-missing-values-for-categor
ies-in-pandas
#replacing nan with most frequently occuring element
X_train['quantity'] = X_train['quantity'].fillna(X_train['quantity'].mode().iloc[0])
X cv['quantity'] = X cv['quantity'].fillna(X cv['quantity'].mode().iloc[0])
X_test['quantity'] = X_test['quantity'].fillna(X_test['quantity'].mode().iloc[0])
In [54]:
# quantity
quantity_scalar = StandardScaler()
X_train_quantity_stndrd = quantity_scalar.fit_transform(X_train['quantity'].values.resh
ape(-1,1))
X_cv_quantity_stndrd = quantity_scalar.transform(X_cv['quantity'].values.reshape(-1,1))
X test quantity stndrd = quantity scalar.transform(X test['quantity'].values.reshape(-1
,1))
In [55]:
X train quantity stndrd
Out[55]:
array([[-0.11135748],
       [-0.11135748],
       [-0.11135748],
       [-0.11135748],
       [-0.11135748],
       [-0.11135748]])
```

```
In [56]:
```

```
X_cv_quantity_stndrd
Out[56]:
array([[-0.11135748],
       [-0.11135748],
       [ 4.10490497],
       [-0.11135748],
       [-0.11135748],
       [-0.11135748]])
In [57]:
X test_quantity_stndrd
Out[57]:
array([[-0.11135748],
       [-0.11135748],
       [-0.11135748],
       . . . ,
       [-0.11135748],
       [-0.11135748],
       [-0.11135748]])
In [58]:
# previous_year_projects
# finding the mean and standard deviation of this data
price_scalar.fit(X_train['teacher_number_of_previously_posted_projects'].values.reshape
(-1,1))
X_train_prev_proj_stndrd =price_scalar.transform(X_train['teacher_number_of_previously_
posted_projects'].values.reshape(-1,1))
X test prev proj stndrd =price scalar.transform(X test['teacher number of previously po
sted projects'].values.reshape(-1, 1))
X cv prev proj stndrd = price scalar.transform(X cv['teacher number of previously poste
d_projects'].values.reshape(-1, 1))
In [59]:
X_train_prev_proj_stndrd
Out[59]:
array([[ 6.53997261],
       [-0.36636304],
       [0.24513543],
       . . . ,
       [-0.36636304]
       [-0.00665805],
       [-0.36636304]])
```

```
In [60]:
```

```
X_cv_prev_proj_stndrd
Out[60]:
array([[ 0.96454539],
       [-0.36636304],
       [-0.33039254],
       [ 6.6119136 ],
       [ 0.13722394],
       [-0.25845154]])
In [61]:
X_test_prev_proj_stndrd
Out[61]:
array([[ 1.10842739],
       [-0.40233353],
       [ 0.49692892],
       . . . ,
       [-0.40233353],
       [ 1.79186685],
       [-0.40233353]])
```

Concatinating all

In [62]:

In [63]:

```
(12000) 14515) (12000)
```

In [64]:

```
from scipy.sparse import hstack
# with the same hstack function we are concatinating a sparse matrix and a dense matirx
X_test_bow = hstack((X_test_title_bow, X_test_essay_bow, test_category_responseCoding, tes
t subcategory responseCoding,
                      test_grade_responseCoding,test_teacher_responseCoding,test_state_
responseCoding,
                      X_test_price_stndrd, X_test_quantity_stndrd, X_test_prev_proj_stn
drd))
print(X_test_bow.shape, y_test.shape)
(16000, 14313) (16000,)
In [65]:
X_train_bow = X_train_bow.tocsr()
X_{cv_bow} = X_{cv_bow.tocsr()}
X_test_bow = X_test_bow.tocsr()
In [66]:
from scipy.sparse import hstack
# with the same hstack function we are concatinating a sparse matrix and a dense matirx
X_train_tfidf = hstack((X_train_title_tfidf,X_train_ess_tfidf,train_category_responseCo
ding,train_subcategory_responseCoding,
                         train_grade_responseCoding,train_teacher_responseCoding,train_
state_responseCoding,
                         X_train_price_stndrd, X_train_quantity_stndrd, X_train_prev_pro
j stndrd))
print(X_train_tfidf.shape, y_train.shape)
(51200, 7066) (51200,)
In [67]:
from scipy.sparse import hstack
# with the same hstack function we are concatinating a sparse matrix and a dense matirx
X_cv_tfidf = hstack((X_cv_title_tfidf,X_cv_ess_tfidf,cv_category_responseCoding,cv_subc
ategory_responseCoding,
                         cv grade responseCoding,cv teacher responseCoding,cv state res
ponseCoding,
                         X cv price stndrd, X cv quantity stndrd, X cv prev proj stndrd
))
print(X_cv_tfidf.shape, y_cv.shape)
```

(12800, 7066) (12800,)

In [68]:

Pretrained avg_w2v model

In [70]:

```
# Reading glove vectors in python: https://stackoverflow.com/a/38230349/4084039

def loadGloveModel(gloveFile):
    print ("Loading Glove Model")

    f = open(gloveFile,'r', encoding = 'utf8')

    model = {}

    for line in tqdm(f):
        splitLine = line.split()
        word = splitLine[0]
        embedding = np.array([float(val) for val in splitLine[1:]])
        model[word] = embedding

    print ("Done.",len(model)," words loaded!")

    return model
```

```
In [71]:
```

```
model = loadGloveModel('glove.42B.300d.txt')

0it [00:00, ?it/s]

Loading Glove Model

1917494it [09:38, 3311.79it/s]

Done. 1917494 words loaded!
```

```
In [72]:
```

```
glove_words = set(model.keys())
```

In [73]:

```
# compute average word2vec for each review.
def func(wordlist):
    train_avg_w2v_vectors = []; # the avg-w2v for each sentence/review is stored in thi
s list
    for sentence in tqdm(wordlist): # for each review/sentence
        vector = np.zeros(300) # as word vectors are of zero length
                                                                        # we are taking
 the 300 dimensions very large
        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
        train_avg_w2v_vectors.append(vector)
    print(len(train_avg_w2v_vectors))
    print(len(train_avg_w2v_vectors[0]))
    return train_avg_w2v_vectors
```

In [74]:

```
| 12800/12800 [00:06<00:00, 1987.16it/s]
| 1%|| | 198/16000 [00:00<00:08, 1910.98it/s]
```

12800 300

100%| 16000/16000 [00:08<00:00, 1933.17it/s]

16000 300

In [75]:

```
X_train_avg_w2v_ess=np.array(X_train_avg_w2v_ess)
X_cv_avg_w2v_ess=np.array(X_cv_avg_w2v_ess)
X_test_avg_w2v_ess =np.array(X_test_avg_w2v_ess)
```

In [76]:

```
# FOR TITLES
X_train_avg_w2v_title=func(preprocessed_train_title)
X_cv_avg_w2v_title=func(preprocessed_cv_title)
X_test_avg_w2v_title=func(preprocessed_test_title)
```

```
100% | 51200/51200 [00:01<00:00, 28325.70it/s]
16% | 2087/12800 [00:00<00:00, 20071.75it/s]
51200
300

100% | 12800/12800 [00:00<00:00, 32007.62it/s]
15% | 12800/12800 [00:00<00:00, 22486.37it/s]

12800
300

100% | 12800 [00:00<00:00, 22486.37it/s]
```

100%| 16000/16000 [00:00<00:00, 30059.13it/s]

16000 300

In [77]:

```
X_train_avg_w2v_title=np.asarray(X_train_avg_w2v_title)
X_cv_avg_w2v_title=np.asarray(X_cv_avg_w2v_title)
X_test_avg_w2v_title =np.asarray(X_test_avg_w2v_title)
```

In [78]:

In [79]:

In [80]:

In [81]:

```
print(X_train_avg_w2v.shape, y_train.shape)
print(X_cv_avg_w2v.shape, y_cv.shape)
print(X_test_avg_w2v.shape, y_test.shape)

(51200, 613) (51200,)
(12800, 613) (12800,)
(16000, 613) (16000,)

In [82]:

X_train_avg_w2v = X_train_avg_w2v[0:20000]
X_cv_avg_w2v = X_cv_avg_w2v[0:20000]
X_test_avg_w2v = X_test_avg_w2v[0:20000]
```

In [83]:

```
# S = ["abc def pqr", "def def def abc", "pqr pqr def"]
tfidf_model = TfidfVectorizer()
tfidf_model.fit(preprocessed_train_essay)
# we are converting a dictionary with word as a key, and the idf as a value
dictionary = dict(zip(tfidf_model.get_feature_names(), list(tfidf_model.idf_)))
tfidf_words = set(tfidf_model.get_feature_names())
```

In [84]:

```
def func(wordlist):
    # average Word2Vec
# compute average word2vec for each review.
    tfidf_w2v_ess = []; # the avg-w2v for each sentence/review is stored in this list
    for sentence in tqdm(wordlist): # 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 tfidf value for each word
                vector += (vec * tf idf) # calculating tfidf weighted w2v
                tf idf weight += tf idf
        if tf idf weight != 0:
            vector /= tf_idf_weight
        tfidf w2v ess.append(vector)
    print(len(tfidf_w2v_ess))
    print(len(tfidf w2v ess[0]))
    return tfidf_w2v_ess
```

In [85]:

```
#For essays
X_train_tfidf_w2v_ess =func(preprocessed_train_essay)
X_test_tfidf_w2v_ess =func(preprocessed_test_essay)
X_cv_tfidf_w2v_ess =func(preprocessed_cv_essay)
100%|
     | 51200/51200 [03:03<00:00, 279.36it/s]
  3%|
| 437/16000 [00:01<00:52, 296.02it/s]
51200
300
100%
     | 16000/16000 [00:56<00:00, 281.94it/s]
| 475/12800 [00:01<00:46, 262.72it/s]
16000
300
100%
       12800/12800 [00:44<00:00, 285.35it/s]
12800
300
In [86]:
X_train_tfidf_w2v_ess =np.array(X_train_tfidf_w2v_ess)
X_test_tfidf_w2v_ess = np.array(X_test_tfidf_w2v_ess)
X_cv_tfidf_w2v_ess = np.array(X_cv_tfidf_w2v_ess)
In [87]:
```

```
# S = ["abc def pqr", "def def def abc", "pqr pqr def"]
tfidf_model = TfidfVectorizer()
tfidf_model.fit(preprocessed_train_title)
# we are converting a dictionary with word as a key, and the idf as a value
dictionary = dict(zip(tfidf_model.get_feature_names(), list(tfidf_model.idf_)))
tfidf_words = set(tfidf_model.get_feature_names())
```

In [88]:

```
#For essays
X_train_tfidf_w2v_title =func(preprocessed_train_title)
X_test_tfidf_w2v_title =func(preprocessed_test_title)
X_cv_tfidf_w2v_title =func(preprocessed_cv_title)
100%
    | 51200/51200 [00:03<00:00, 13463.33it/s]
 17%
2732/16000 [00:00<00:01, 13050.64it/s]
51200
```

300

100% 16000/16000 [00:01<00:00, 13471.43it/s] 21%|

2634/12800 [00:00<00:00, 12737.50it/s]

16000 300

100% | 12800/12800 [00:01<00:00, 12503.10it/s]

12800 300

In [89]:

```
X train tfidf w2v title =np.array(X train tfidf w2v title)
X_test_tfidf_w2v_title =np.array(X_test_tfidf_w2v_title)
X_cv_tfidf_w2v_title =np.array(X_cv_tfidf_w2v_title)
```

In [90]:

```
X_train_tfidf_w2v = np.hstack((train_category_responseCoding,train_subcategory_response
Coding, train_grade_responseCoding
                          ,train_teacher_responseCoding,train_state_responseCoding,
                            X_train_tfidf_w2v_title,X_train_tfidf_w2v_ess,X_train_price
_stndrd, X_train_quantity_stndrd, X_train_prev_proj_stndrd))
```

In [91]:

```
X_cv_tfidf_w2v = np.hstack((cv_category_responseCoding,cv_subcategory_responseCoding,cv
grade responseCoding
                          ,cv_teacher_responseCoding,cv_state_responseCoding,
                            X_cv_tfidf_w2v_title,X_cv_tfidf_w2v_ess,X_cv_price_stndrd,
X_cv_quantity_stndrd, X_cv_prev_proj_stndrd))
```

In [92]:

```
X_test_tfidf_w2v = np.hstack((test_category_responseCoding,test_subcategory_responseCod
ing,test_grade_responseCoding
                          ,test_teacher_responseCoding,test_state_responseCoding,
                            X_test_tfidf_w2v_title,X_test_tfidf_w2v_ess,X_test_price_st
ndrd, X_test_quantity_stndrd, X_test_prev_proj_stndrd))
```

```
In [93]:
```

```
print(X_train_tfidf_w2v.shape, y_train.shape)
print(X_cv_tfidf_w2v.shape, y_cv.shape)
print(X_test_tfidf_w2v.shape, y_test.shape)

(51200, 613) (51200,)
(12800, 613) (12800,)
(16000, 613) (16000,)

In [94]:

X_train_tfidf_w2v = X_train_tfidf_w2v[0:20000]
X_cv_tfidf_w2v = X_cv_tfidf_w2v[0:20000]
X_test_tfidf_w2v = X_test_tfidf_w2v[0:20000]
```

Random Forest Classifier on bag of words

In [116]:

```
from sklearn.metrics import f1 score
from sklearn.ensemble import RandomForestClassifier
from sklearn.metrics import f1 score
from sklearn.model_selection import RandomizedSearchCV
from scipy.stats import randint as sp randint
from xgboost.sklearn import XGBClassifier
from scipy.stats import uniform
param_dist = {"n_estimators":[10, 50, 100, 150, 200, 300, 500, 1000],
              "max_depth":[2, 3, 4, 5, 6, 7, 8, 9, 10],
              "min_samples_split": sp_randint(110,190),
              "min samples leaf": sp randint(20,75)}
clf = RandomForestClassifier(random_state=42, class_weight = 'balanced')
rf = RandomizedSearchCV(clf, param distributions=param dist, return train score = True,
                                   n_iter=5,cv=10,scoring='f1',random_state=42)
rf.fit(X_train_bow,y_train)
train_auc= rf.cv_results_['mean_train_score']
train auc std= rf.cv results ['std train score']
cv auc = rf.cv results ['mean test score']
cv_auc_std= rf.cv_results_['std_test_score']
```

```
In [117]:
```

```
train_auc
Out[117]:
array([0.75948755, 0.76957798, 0.76308804, 0.7551134 , 0.77102644])
```

```
In [118]:
train auc std
Out[118]:
array([0.00270147, 0.001997 , 0.00331074, 0.00155028, 0.00351632])
In [119]:
cv_auc
Out[119]:
array([0.75095856, 0.75822382, 0.75256765, 0.7472037, 0.76074461])
In [120]:
cv auc std
Out[120]:
array([0.00759748, 0.00825959, 0.00610561, 0.00704392, 0.00689054])
In [121]:
rf.best_estimator_
Out[121]:
RandomForestClassifier(bootstrap=True, ccp_alpha=0.0, class_weight='balanc
ed',
                       criterion='gini', max_depth=7, max_features='auto',
                       max_leaf_nodes=None, max_samples=None,
                       min_impurity_decrease=0.0, min_impurity_split=None,
                       min_samples_leaf=21, min_samples_split=173,
                       min_weight_fraction_leaf=0.0, n_estimators=150,
                       n_jobs=None, oob_score=False, random_state=42, verb
ose=0,
                       warm_start=False)
```

In [122]:

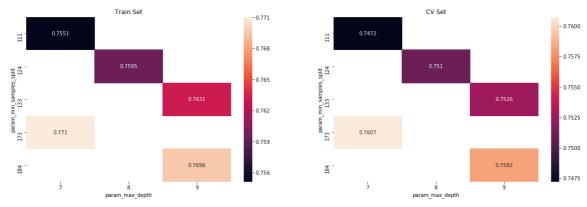
```
max_scores = pd.DataFrame(rf.cv_results_).groupby(['param_min_samples_split', 'param_ma
x_depth']).max().unstack()[['mean_test_score', 'mean_train_score']]

fig, ax = plt.subplots(1,2, figsize=(20,6))

sns.heatmap(max_scores.mean_train_score, annot = True, fmt='.4g', ax=ax[0])
sns.heatmap(max_scores.mean_test_score, annot = True, fmt='.4g', ax=ax[1])

ax[0].set_title('Train Set')
ax[1].set_title('CV Set')

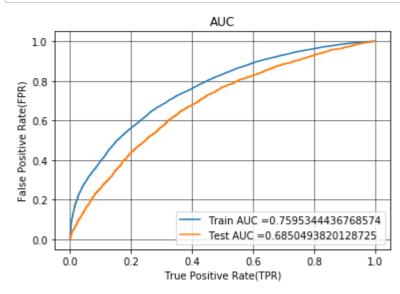
plt.show()
```



Hyperparameter tuning

In [124]:

```
# https://scikit-learn.org/stable/modules/generated/sklearn.metrics.roc curve.html#skle
arn.metrics.roc_curve
from sklearn.metrics import roc_curve, auc
model = RandomForestClassifier(max_depth = 7, n_estimators = 150, class_weight= 'balanc
ed')
model.fit(X_train_bow,y_train)
y_train_pred = model.predict_proba(X_train_bow)[:,1]
y test pred = model.predict proba(X test bow)[:,1]
train fpr, train_tpr, tr_thresholds = roc_curve(y_train, y_train_pred)
test_fpr, test_tpr, te_thresholds = roc_curve(y_test, y_test_pred)
plt.plot(train_fpr, train_tpr, label="Train AUC ="+str(auc(train_fpr, train_tpr)))
plt.plot(test_fpr, test_tpr, label="Test AUC ="+str(auc(test_fpr, test tpr)))
plt.legend()
plt.xlabel("True Positive Rate(TPR)")
plt.ylabel("False Positive Rate(FPR)")
plt.title("AUC")
plt.grid(color='black', linestyle='-', linewidth=0.5)
plt.show()
```



best threshold

In [125]:

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

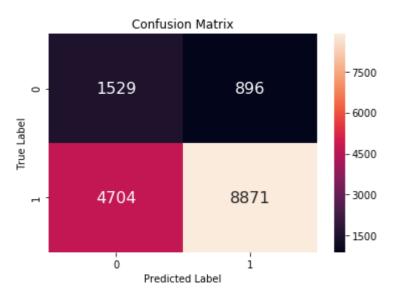
confusion matrix

In [127]:

```
#stackoverflow.com/questions/54018742/valueerror-classification-metrics-cant-handle-a-m
ix-of-unknown-and-binary-targ
matrix = confusion_matrix(y_test, predict_with_best_t(y_test_pred, best_t))
sns.heatmap(matrix, annot=True, annot_kws={'size':16}, fmt='g')
plt.ylabel('True Label')
plt.xlabel('Predicted Label')
plt.title('Confusion Matrix')
```

Out[127]:

Text(0.5, 1, 'Confusion Matrix')

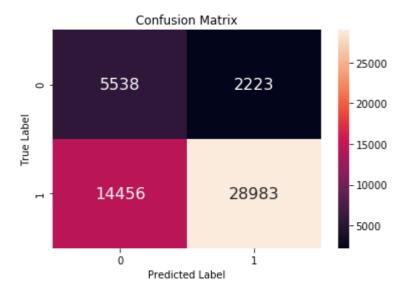


In [128]:

```
#stackoverflow.com/questions/54018742/valueerror-classification-metrics-cant-handle-a-m
ix-of-unknown-and-binary-targ
matrix = confusion_matrix(y_train, predict_with_best_t(y_train_pred, best_t))
sns.heatmap(matrix, annot=True, annot_kws={'size':16}, fmt='g')
plt.ylabel('True Label')
plt.xlabel('Predicted Label')
plt.title('Confusion Matrix')
```

Out[128]:

Text(0.5, 1, 'Confusion Matrix')



Random forest classifier on tfidf

In [129]:

In [130]:

```
rf.best_estimator_
```

Out[130]:

warm_start=False)

In [131]:

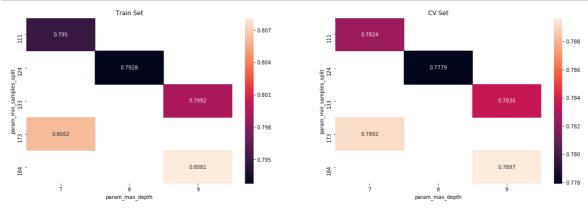
```
max_scores = pd.DataFrame(rf.cv_results_).groupby(['param_min_samples_split', 'param_ma
x_depth']).max().unstack()[['mean_test_score', 'mean_train_score']]

fig, ax = plt.subplots(1,2, figsize=(20,6))

sns.heatmap(max_scores.mean_train_score, annot = True, fmt='.4g', ax=ax[0])
sns.heatmap(max_scores.mean_test_score, annot = True, fmt='.4g', ax=ax[1])

ax[0].set_title('Train Set')
ax[1].set_title('CV Set')

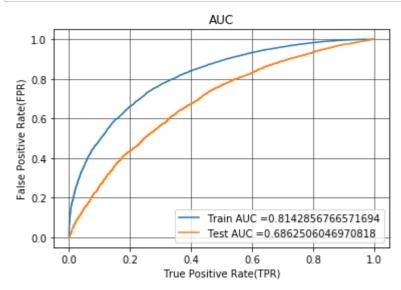
plt.show()
```



Hyperparameter tuning

In [132]:

```
# https://scikit-learn.org/stable/modules/generated/sklearn.metrics.roc curve.html#skle
arn.metrics.roc_curve
from sklearn.metrics import roc_curve, auc
model = RandomForestClassifier(max_depth = 9, n_estimators = 100, class_weight= 'balanc
ed')
model.fit(X_train_tfidf,y_train)
y_train_pred = model.predict_proba(X_train_tfidf)[:,1]
y test pred = model.predict proba(X test tfidf)[:,1]
train_fpr, train_tpr, tr_thresholds = roc_curve(y_train, y_train_pred)
test_fpr, test_tpr, te_thresholds = roc_curve(y_test, y_test_pred)
plt.plot(train_fpr, train_tpr, label="Train AUC ="+str(auc(train_fpr, train_tpr)))
plt.plot(test_fpr, test_tpr, label="Test AUC ="+str(auc(test_fpr, test tpr)))
plt.legend()
plt.xlabel("True Positive Rate(TPR)")
plt.ylabel("False Positive Rate(FPR)")
plt.title("AUC")
plt.grid(color='black', linestyle='-', linewidth=0.5)
plt.show()
```



best threshold

In [133]:

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

Confusion matrix

In [134]:

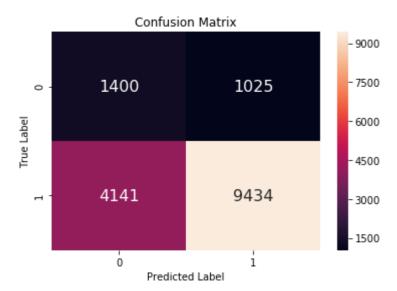
```
from sklearn.metrics import confusion_matrix
best_t = find_best_threshold(tr_thresholds, train_fpr, train_tpr)
print("Train confusion matrix")
print(confusion_matrix(y_train, predict_with_best_t(y_train_pred, best_t)))
print("Test confusion matrix")
print(confusion_matrix(y_test, predict_with_best_t(y_test_pred, best_t)))
print('='*100)
the maximum value of tpr*(1-fpr) 0.5440109365026429 for threshold 0.505
Train confusion matrix
[[ 5775 1986]
[11681 31758]]
Test confusion matrix
[[1400 1025]
 [4141 9434]]
______
```

In [135]:

```
#stackoverflow.com/questions/54018742/valueerror-classification-metrics-cant-handle-a-m
ix-of-unknown-and-binary-targ
matrix = confusion_matrix(y_test, predict_with_best_t(y_test_pred, best_t))
sns.heatmap(matrix, annot=True, annot_kws={'size':16}, fmt='g')
plt.ylabel('True Label')
plt.xlabel('Predicted Label')
plt.title('Confusion Matrix')
```

Out[135]:

Text(0.5, 1, 'Confusion Matrix')

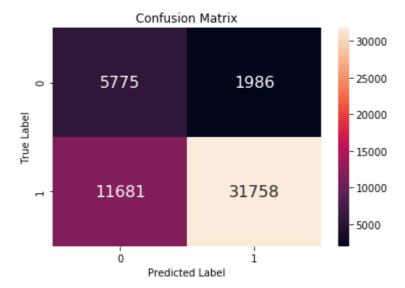


In [136]:

```
#stackoverflow.com/questions/54018742/valueerror-classification-metrics-cant-handle-a-m
ix-of-unknown-and-binary-targ
matrix = confusion_matrix(y_train, predict_with_best_t(y_train_pred, best_t))
sns.heatmap(matrix, annot=True, annot_kws={'size':16}, fmt='g')
plt.ylabel('True Label')
plt.xlabel('Predicted Label')
plt.title('Confusion Matrix')
```

Out[136]:

Text(0.5, 1, 'Confusion Matrix')



Random forest classifier on avg_w2v

In [138]:

```
y_train1 = y_train[0:20000]
y_cv1 = y_cv[0:20000]
y_test1 = y_test[0:20000]
```

In [139]:

In [140]:

```
rf.best_estimator_
```

Out[140]:

RandomForestClassifier(bootstrap=True, ccp_alpha=0.0, class_weight='balanc
ed',

criterion='gini', max_depth=9, max_features='auto',
max_leaf_nodes=None, max_samples=None,
min_impurity_decrease=0.0, min_impurity_split=None,
min_samples_leaf=72, min_samples_split=133,
min_weight_fraction_leaf=0.0, n_estimators=100,
n_jobs=None, oob_score=False, random_state=42, verb

ose=0,

warm_start=False)

In [141]:

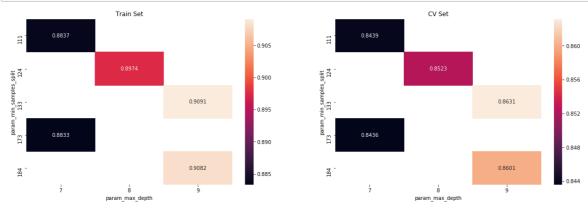
```
max_scores = pd.DataFrame(rf.cv_results_).groupby(['param_min_samples_split', 'param_ma
x_depth']).max().unstack()[['mean_test_score', 'mean_train_score']]

fig, ax = plt.subplots(1,2, figsize=(20,6))

sns.heatmap(max_scores.mean_train_score, annot = True, fmt='.4g', ax=ax[0])
sns.heatmap(max_scores.mean_test_score, annot = True, fmt='.4g', ax=ax[1])

ax[0].set_title('Train Set')
ax[1].set_title('CV Set')

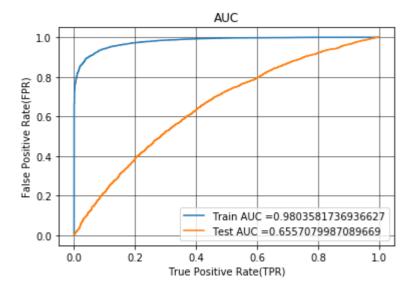
plt.show()
```



Hyperparameter tuning

In [142]:

```
# https://scikit-learn.org/stable/modules/generated/sklearn.metrics.roc curve.html#skle
arn.metrics.roc_curve
from sklearn.metrics import roc_curve, auc
model = RandomForestClassifier(max_depth =9 , n_estimators = 100, class_weight= 'balanc
ed')
model.fit(X_train_avg_w2v,y_train1)
y_train_pred = model.predict_proba(X_train_avg_w2v)[:,1]
y test pred = model.predict proba(X test avg w2v)[:,1]
train_fpr, train_tpr, tr_thresholds = roc_curve(y_train1, y_train_pred)
test_fpr, test_tpr, te_thresholds = roc_curve(y_test1, y_test_pred)
plt.plot(train_fpr, train_tpr, label="Train AUC ="+str(auc(train_fpr, train_tpr)))
plt.plot(test fpr, test tpr, label="Test AUC ="+str(auc(test fpr, test tpr)))
plt.legend()
plt.xlabel("True Positive Rate(TPR)")
plt.ylabel("False Positive Rate(FPR)")
plt.title("AUC")
plt.grid(color='black', linestyle='-', linewidth=0.5)
plt.show()
```



best threshold

In [143]:

confusion matrix

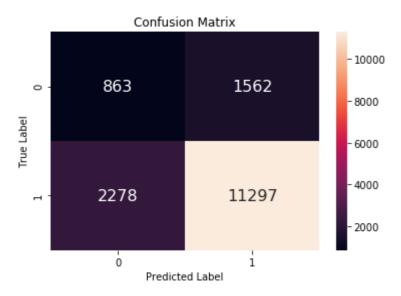
```
In [144]:
from sklearn.metrics import confusion matrix
best_t = find_best_threshold(tr_thresholds, train_fpr, train_tpr)
print("Train confusion matrix")
print(confusion_matrix(y_train1, predict_with_best_t(y_train_pred, best_t)))
print("Test confusion matrix")
print(confusion_matrix(y_test1, predict_with_best_t(y_test_pred, best_t)))
print('='*100)
the maximum value of tpr*(1-fpr) 0.8583300568527676 for threshold 0.544
Train confusion matrix
[[ 2855
        137]
[ 1709 15299]]
Test confusion matrix
[[ 863 1562]
 [ 2278 11297]]
```

In [145]:

```
#stackoverflow.com/questions/54018742/valueerror-classification-metrics-cant-handle-a-m
ix-of-unknown-and-binary-targ
matrix = confusion_matrix(y_test1, predict_with_best_t(y_test_pred, best_t))
sns.heatmap(matrix, annot=True, annot_kws={'size':16}, fmt='g')
plt.ylabel('True Label')
plt.xlabel('Predicted Label')
plt.title('Confusion Matrix')
```

Out[145]:

Text(0.5, 1, 'Confusion Matrix')

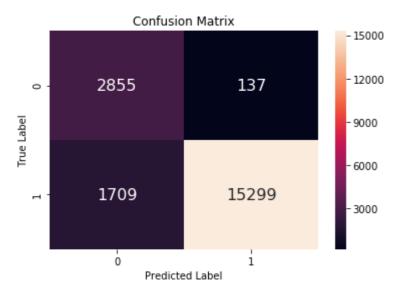


In [146]:

```
#stackoverflow.com/questions/54018742/valueerror-classification-metrics-cant-handle-a-m
ix-of-unknown-and-binary-targ
matrix = confusion_matrix(y_train1, predict_with_best_t(y_train_pred, best_t))
sns.heatmap(matrix, annot=True, annot_kws={'size':16}, fmt='g')
plt.ylabel('True Label')
plt.xlabel('Predicted Label')
plt.title('Confusion Matrix')
```

Out[146]:

Text(0.5, 1, 'Confusion Matrix')



Random forest classifier on tfidf_w2v

In [147]:

In [149]:

```
rf.best_estimator_
```

Out[149]:

RandomForestClassifier(bootstrap=True, ccp_alpha=0.0, class_weight='balanc
ed',

criterion='gini', max_depth=9, max_features='auto',
max_leaf_nodes=None, max_samples=None,
min_impurity_decrease=0.0, min_impurity_split=None,
min_samples_leaf=72, min_samples_split=133,
min_weight_fraction_leaf=0.0, n_estimators=100,
n_jobs=None, oob_score=False, random_state=42, verb

ose=0,

warm_start=False)

In [150]:

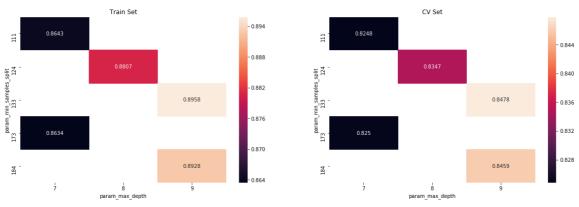
```
max_scores = pd.DataFrame(rf.cv_results_).groupby(['param_min_samples_split', 'param_ma
x_depth']).max().unstack()[['mean_test_score', 'mean_train_score']]

fig, ax = plt.subplots(1,2, figsize=(20,6))

sns.heatmap(max_scores.mean_train_score, annot = True, fmt='.4g', ax=ax[0])
sns.heatmap(max_scores.mean_test_score, annot = True, fmt='.4g', ax=ax[1])

ax[0].set_title('Train Set')
ax[1].set_title('CV Set')

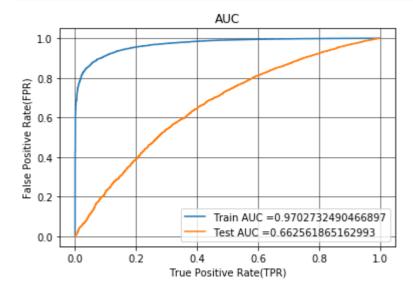
plt.show()
```



Hyperparameter tuning

In [151]:

```
# https://scikit-learn.org/stable/modules/generated/sklearn.metrics.roc curve.html#skle
arn.metrics.roc_curve
from sklearn.metrics import roc_curve, auc
model = RandomForestClassifier(max_depth = 9, n_estimators = 100, class_weight= 'balanc
ed')
model.fit(X_train_tfidf_w2v,y_train1)
y_train_pred = model.predict_proba(X_train_tfidf_w2v)[:,1]
y test pred = model.predict proba(X test tfidf w2v)[:,1]
train_fpr, train_tpr, tr_thresholds = roc_curve(y_train1, y_train_pred)
test_fpr, test_tpr, te_thresholds = roc_curve(y_test1, y_test_pred)
plt.plot(train_fpr, train_tpr, label="Train AUC ="+str(auc(train_fpr, train_tpr)))
plt.plot(test fpr, test tpr, label="Test AUC ="+str(auc(test fpr, test tpr)))
plt.legend()
plt.xlabel("True Positive Rate(TPR)")
plt.ylabel("False Positive Rate(FPR)")
plt.title("AUC")
plt.grid(color='black', linestyle='-', linewidth=0.5)
plt.show()
```



best threshold

In [152]:

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

confusion matrix

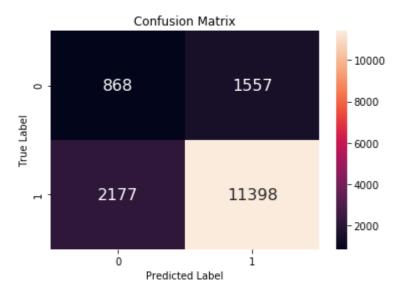
```
In [153]:
from sklearn.metrics import confusion_matrix
best_t = find_best_threshold(tr_thresholds, train_fpr, train_tpr)
print("Train confusion matrix")
print(confusion_matrix(y_train1, predict_with_best_t(y_train_pred, best_t)))
print("Test confusion matrix")
print(confusion_matrix(y_test1, predict_with_best_t(y_test_pred, best_t)))
print('='*100)
the maximum value of tpr*(1-fpr) 0.8296152549790976 for threshold 0.521
Train confusion matrix
[[ 2776
        216]
[ 1800 15208]]
Test confusion matrix
[[ 868 1557]
 [ 2177 11398]]
______
```

In [154]:

```
#stackoverflow.com/questions/54018742/valueerror-classification-metrics-cant-handle-a-m
ix-of-unknown-and-binary-targ
matrix = confusion_matrix(y_test1, predict_with_best_t(y_test_pred, best_t))
sns.heatmap(matrix, annot=True, annot_kws={'size':16}, fmt='g')
plt.ylabel('True Label')
plt.xlabel('Predicted Label')
plt.title('Confusion Matrix')
```

Out[154]:

Text(0.5, 1, 'Confusion Matrix')

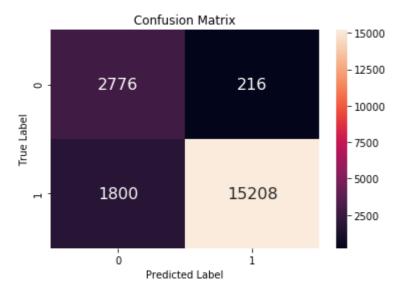


In [156]:

```
#stackoverflow.com/questions/54018742/valueerror-classification-metrics-cant-handle-a-m
ix-of-unknown-and-binary-targ
matrix = confusion_matrix(y_train1, predict_with_best_t(y_train_pred, best_t))
sns.heatmap(matrix, annot=True, annot_kws={'size':16}, fmt='g')
plt.ylabel('True Label')
plt.xlabel('Predicted Label')
plt.title('Confusion Matrix')
```

Out[156]:

Text(0.5, 1, 'Confusion Matrix')



Summary

In [158]:

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

ptable = PrettyTable()
ptable.title = 'Classification Report'

ptable.field_names = ["Vectorization", "Model", "max_depth", "n_estimator", "AUC"]

ptable.add_row(["BOW","RF",7,150,68.50])
ptable.add_row(["tf-idf", "RF",9,100,68.62])
ptable.add_row(["avg-w2v", "RF",9,100,65.57])
ptable.add_row(["tf-idf-w2v", "RF",9,100,66.25])

print(ptable)
```

4						_
	Vectorization	Model	max_depth	n_estimator	AUC	
	BOW tf-idf avg-w2v	RF RF RF	7 9 9	150 100 100	68.5 68.62 65.57	
	tf-idf-w2v	RF	9	100	66.25	

Initially after loading the dataset if any null values exists replace them with most occurring element then split the data into training, validation, testing data and preprocessed the data to avoid the leakage.

Applied bag of words, tfidf, avg_w2v, tfidf_w2v featurising on the data. After concatenating all the features applied random forest on each.

More accuracy is obtained for tfidf.