

# DonorsChoose

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

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

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

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

## About the DonorsChoose Data Set

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

Feature		Desc
<code>project_id</code>		A unique identifier for the proposed project. <b>Example:</b> p0
<code>project_title</code>	• •	Title of the project. <b>Example:</b> Art Will Make You H First Grad
<code>project_grade_category</code>	• • • •	Grade level of students for which the project is targeted. One of the following enumerated values: Grades P Grade Grade Grades
<code>project_subject_categories</code>	• • • • • • • •	One or more (comma-separated) subject categories for the project from the following enumerated list of values: Applied Learning Care & Health Health & Safety History & Culture Literacy & Language Math & Science Music & The Arts Special Education
	• •	<b>Example:</b> Music & The Arts Literacy & Language, Math & Science

Feature	Desc
<code>school_state</code>	State where school is located ( <a href="https://en.wikipedia.org/wiki/List_of_U.S._state_abbreviations#Postal_codes">Two-letter U.S. postal codes</a> ) ( <a href="https://en.wikipedia.org/wiki/List_of_U.S._state_abbreviations#Postal_codes">https://en.wikipedia.org/wiki/List of U.S. state abbreviations#Postal codes</a> ) <b>Example:</b> CA
<code>project_subject_subcategories</code>	One or more (comma-separated) subject subcategories for the project. <b>Example:</b> Lit, Literature & Writing, Social Sci
<code>project_resource_summary</code>	An explanation of the resources needed for the project. <b>Example:</b> My students need hands on literacy materials to meet sensory needs!<
<code>project_essay_1</code>	First application
<code>project_essay_2</code>	Second application
<code>project_essay_3</code>	Third application
<code>project_essay_4</code>	Fourth application
<code>project_submitted_datetime</code>	Datetime when project application was submitted. <b>Example:</b> 2016-01-12:43:5
<code>teacher_id</code>	A unique identifier for the teacher of the proposed project. <b>Example:</b> bdf8baa8fedef6bfeec7ae4ff1c
<code>teacher_prefix</code>	Teacher's title. One of the following enumerated values: • • • • • • Tea
<code>teacher_number_of_previously_posted_projects</code>	Number of project applications previously submitted by the same teacher. <b>Example:</b>

\* 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
<code>id</code>	A <code>project_id</code> value from the <code>train.csv</code> file. <b>Example:</b> p036502
<code>description</code>	Description of the resource. <b>Example:</b> Tenor Saxophone Reeds, Box of 25
<code>quantity</code>	Quantity of the resource required. <b>Example:</b> 3
<code>price</code>	Price of the resource required. <b>Example:</b> 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
<code>project_is_approved</code>	A binary flag indicating whether DonorsChoose approved the project. A value of <code>0</code> indicates the project was not approved, and a value of <code>1</code> 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]:

```
import sys
!{sys.executable} -m pip install gensim
```

Collecting gensim

Using cached [https://files.pythonhosted.org/packages/ef/65/c90886ac34d4b12d3ae0bcc7aece1af57e1f30e7138aabb3e3c027e705a/gensim-3.8.0-cp35-cp35m-manylinux1\\_x86\\_64.whl](https://files.pythonhosted.org/packages/ef/65/c90886ac34d4b12d3ae0bcc7aece1af57e1f30e7138aabb3e3c027e705a/gensim-3.8.0-cp35-cp35m-manylinux1_x86_64.whl) ([https://files.pythonhosted.org/packages/ef/65/c90886ac34d4b12d3ae0bcc7aece1af57e1f30e7138aabb3e3c027e705a/gensim-3.8.0-cp35-cp35m-manylinux1\\_x86\\_64.whl](https://files.pythonhosted.org/packages/ef/65/c90886ac34d4b12d3ae0bcc7aece1af57e1f30e7138aabb3e3c027e705a/gensim-3.8.0-cp35-cp35m-manylinux1_x86_64.whl))

Collecting scipy>=0.18.1 (from gensim)

Using cached [https://files.pythonhosted.org/packages/7a/0e/3781e028d62a842244582abd8f084e6314297026760587c85607f687bf3/scipy-1.3.1-cp35-cp35m-manylinux1\\_x86\\_64.whl](https://files.pythonhosted.org/packages/7a/0e/3781e028d62a842244582abd8f084e6314297026760587c85607f687bf3/scipy-1.3.1-cp35-cp35m-manylinux1_x86_64.whl) ([https://files.pythonhosted.org/packages/7a/0e/3781e028d62a842244582abd8f084e6314297026760587c85607f687bf3/scipy-1.3.1-cp35-cp35m-manylinux1\\_x86\\_64.whl](https://files.pythonhosted.org/packages/7a/0e/3781e028d62a842244582abd8f084e6314297026760587c85607f687bf3/scipy-1.3.1-cp35-cp35m-manylinux1_x86_64.whl))

Collecting six>=1.5.0 (from gensim)

Using cached <https://files.pythonhosted.org/packages/73/fb/00a976f728d0d1fecfe898238ce23f502a721c0ac0ecfedb80e0d88c64e9/six-1.12.0-py2.py3-none-any.whl> (<https://files.pythonhosted.org/packages/73/fb/00a976f728d0d1fecfe898238ce23f502a721c0ac0ecfedb80e0d88c64e9/six-1.12.0-py2.py3-none-any.whl>)

Collecting smart-open>=1.7.0 (from gensim)

Collecting numpy>=1.11.3 (from gensim)

Using cached [https://files.pythonhosted.org/packages/69/25/eef8d362bd216b11e7d005331a3cca3d19b0aa57569bde680070109b745c/numpy-1.17.0-cp35-cp35m-manylinux1\\_x86\\_64.whl](https://files.pythonhosted.org/packages/69/25/eef8d362bd216b11e7d005331a3cca3d19b0aa57569bde680070109b745c/numpy-1.17.0-cp35-cp35m-manylinux1_x86_64.whl) ([https://files.pythonhosted.org/packages/69/25/eef8d362bd216b11e7d005331a3cca3d19b0aa57569bde680070109b745c/numpy-1.17.0-cp35-cp35m-manylinux1\\_x86\\_64.whl](https://files.pythonhosted.org/packages/69/25/eef8d362bd216b11e7d005331a3cca3d19b0aa57569bde680070109b745c/numpy-1.17.0-cp35-cp35m-manylinux1_x86_64.whl))

Collecting boto>=2.32 (from smart-open>=1.7.0->gensim)

Using cached <https://files.pythonhosted.org/packages/23/10/c0b78c27298029e4454a472a1919bde20cb182dab1662cec7f2ca1dcc523/boto-2.49.0-py2.py3-none-any.whl> (<https://files.pythonhosted.org/packages/23/10/c0b78c27298029e4454a472a1919bde20cb182dab1662cec7f2ca1dcc523/boto-2.49.0-py2.py3-none-any.whl>)

Collecting boto3 (from smart-open>=1.7.0->gensim)

Using cached <https://files.pythonhosted.org/packages/ff/3e/2262936ad70fd6e7b8827d79d6508ce33e2ffb49bfca6fedc5fe4abd6f1c/boto3-1.9.215-py2.py3-none-any.whl> (<https://files.pythonhosted.org/packages/ff/3e/2262936ad70fd6e7b8827d79d6508ce33e2ffb49bfca6fedc5fe4abd6f1c/boto3-1.9.215-py2.py3-none-any.whl>)

Collecting requests (from smart-open>=1.7.0->gensim)

Using cached <https://files.pythonhosted.org/packages/51/bd/23c926cd341ea6b7dd0b2a00aba99ae0f828be89d72b2190f27c11d4b7fb/requests-2.22.0-py2.py3-none-any.whl> (<https://files.pythonhosted.org/packages/51/bd/23c926cd341ea6b7dd0b2a00aba99ae0f828be89d72b2190f27c11d4b7fb/requests-2.22.0-py2.py3-none-any.whl>)

Collecting botocore<1.13.0,>=1.12.215 (from boto3->smart-open>=1.7.0->gensim)

Using cached <https://files.pythonhosted.org/packages/a1/b0/7a8794d914b95ef3335a5a4ba20595b46081dbd1e29f13812eceacf091ca/botocore-1.12.215-py2.py3-none-any.whl> (<https://files.pythonhosted.org/packages/a1/b0/7a8794d914b95ef3335a5a4ba20595b46081dbd1e29f13812eceacf091ca/botocore-1.12.215-py2.py3-none-any.whl>)

Collecting s3transfer<0.3.0,>=0.2.0 (from boto3->smart-open>=1.7.0->gensim)

Using cached <https://files.pythonhosted.org/packages/16/8a/1fc3dba0c4923c2a76e1ff0d52b305c44606da63f718d14d3231e21c51b0/s3transfer-0.2.1-py2.py3-none-any.whl> (<https://files.pythonhosted.org/packages/16/8a/1fc3dba0c4923c2a76e1ff0d52b305c44606da63f718d14d3231e21c51b0/s3transfer-0.2.1-py2.py3-none-any.whl>)

Collecting jmespath<1.0.0,>=0.7.1 (from boto3->smart-open>=1.7.0->gensim)

Using cached <https://files.pythonhosted.org/packages/83/94/7179c3832a6d45b266ddb2aac329e101367fbdb11f425f13771d27f225bb/jmespath-0.9.4-py2.py3-none-any.whl> (<https://files.pythonhosted.org/packages/83/94/7179c3832a6d45b266ddb2aac329e101367fbdb11f425f13771d27f225bb/jmespath-0.9.4-py2.py3-none-any.whl>)

```

ac329e101367fbbdb11f425f13771d27f225bb/jmespath-0.9.4-py2.py3-none-any.whl)
Collecting chardet<3.1.0,>=3.0.2 (from requests->smart-open>=1.7.0->gensim)
  Using cached https://files.pythonhosted.org/packages/bc/a9/01ffebfb562e4274b6487b4bb1ddec7ca55ec7510b22e4c51f14098443b8/chardet-3.0.4-py2.py3-none-any.whl (https://files.pythonhosted.org/packages/bc/a9/01ffebfb562e4274b6487b4bb1ddec7ca55ec7510b22e4c51f14098443b8/chardet-3.0.4-py2.py3-none-any.whl)
Collecting urllib3!=1.25.0,!1.25.1,<1.26,>=1.21.1 (from requests->smart-open>=1.7.0->gensim)
  Using cached https://files.pythonhosted.org/packages/e6/60/247f23a7121ae632d62811ba7f273d0e58972d75e58a94d329d51550a47d/urllib3-1.25.3-py2.py3-none-any.whl (https://files.pythonhosted.org/packages/e6/60/247f23a7121ae632d62811ba7f273d0e58972d75e58a94d329d51550a47d/urllib3-1.25.3-py2.py3-none-any.whl)
Collecting idna<2.9,>=2.5 (from requests->smart-open>=1.7.0->gensim)
  Using cached https://files.pythonhosted.org/packages/14/2c/cd551d81dbe15200be1cf41cd03869a46fe7226e7450af7a6545bfc474c9/idna-2.8-py2.py3-none-any.whl (https://files.pythonhosted.org/packages/14/2c/cd551d81dbe15200be1cf41cd03869a46fe7226e7450af7a6545bfc474c9/idna-2.8-py2.py3-none-any.whl)
Collecting certifi>=2017.4.17 (from requests->smart-open>=1.7.0->gensim)
  Using cached https://files.pythonhosted.org/packages/69/1b/b853c7a9d4f6a6d00749e94eb6f3a041e342a885b87340b79c1ef73e3a78/certifi-2019.6.16-py2.py3-none-any.whl (https://files.pythonhosted.org/packages/69/1b/b853c7a9d4f6a6d00749e94eb6f3a041e342a885b87340b79c1ef73e3a78/certifi-2019.6.16-py2.py3-none-any.whl)
Collecting python-dateutil<3.0.0,>=2.1; python_version >= "2.7" (from boto3-re<1.13.0,>=1.12.215->boto3->smart-open>=1.7.0->gensim)
  Using cached https://files.pythonhosted.org/packages/41/17/c62facbfbfd163c7f57f3844689e3a78bae1f403648a6afb1d0866d87fbb/python_dateutil-2.8.0-py2.py3-none-any.whl (https://files.pythonhosted.org/packages/41/17/c62facbfbfd163c7f57f3844689e3a78bae1f403648a6afb1d0866d87fbb/python_dateutil-2.8.0-py2.py3-none-any.whl)
Collecting docutils<0.16,>=0.10 (from botocore<1.13.0,>=1.12.215->boto3->smart-open>=1.7.0->gensim)
  Using cached https://files.pythonhosted.org/packages/22/cd/a6aa959dca619918ccb55023b4cb151949c64d4d5d55b3f4ffd7eee0c6e8/docutils-0.15.2-py3-none-any.whl (https://files.pythonhosted.org/packages/22/cd/a6aa959dca619918ccb55023b4cb151949c64d4d5d55b3f4ffd7eee0c6e8/docutils-0.15.2-py3-none-any.whl)
Installing collected packages: numpy, scipy, six, boto, urllib3, jmespath, python-dateutil, docutils, botocore, s3transfer, boto3, chardet, idna, certifi, requests, smart-open, gensim
Successfully installed boto-2.49.0 boto3-1.9.215 botocore-1.12.215 certifi-2019.6.16 chardet-3.0.4 docutils-0.15.2 gensim-3.8.0 idna-2.8 jmespath-0.9.4 numpy-1.17.0 python-dateutil-2.8.0 requests-2.22.0 s3transfer-0.2.1 scipy-1.3.1 six-1.12.0 smart-open-1.8.4 urllib3-1.25.3

```

In [2]:

```
%matplotlib inline
import warnings
warnings.filterwarnings("ignore")

import sqlite3
import pandas as pd
import numpy as np
import nltk
import string
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.feature_extraction.text import TfidfTransformer
from sklearn.feature_extraction.text import TfidfVectorizer

from sklearn.feature_extraction.text import CountVectorizer
from sklearn.metrics import confusion_matrix
from sklearn import metrics
from sklearn.metrics import roc_curve, auc
from nltk.stem.porter import PorterStemmer

import re
# Tutorial about Python regular expressions: https://pymotw.com/2/re/
import string
from nltk.corpus import stopwords
from nltk.stem import PorterStemmer
from nltk.stem.wordnet import WordNetLemmatizer

from gensim.models import Word2Vec
from gensim.models import KeyedVectors
import pickle

from tqdm import tqdm
import os

from plotly import plotly
import plotly.offline as offline
import plotly.graph_objs as go
offline.init_notebook_mode()
from collections import Counter
```

## 1.1 Reading Data

In [3]:

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

In [4]:

```
print("Number of data points in train data", project_data.shape)
print('-'*50)
print("The attributes of data :", project_data.columns.values)
```

Number of data points in train data (109248, 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 [5]:

```
print("Number of data points in train data", resource_data.shape)
print(resource_data.columns.values)
resource_data.head(2)
```

Number of data points in train data (1541272, 4)

['id' 'description' 'quantity' 'price']

Out[5]:

	id	description	quantity	price
0	p233245	LC652 - Lakeshore Double-Space Mobile Drying Rack	1	149.00
1	p069063	Bouncy Bands for Desks (Blue support pipes)	3	14.95

In [6]:

```
# 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]

project_data.head(2)
```

Out[6]:

	Unnamed: 0	id	teacher_id	teacher_prefix	school_state	
55660	8393	p205479	2bf07ba08945e5d8b2a3f269b2b3cfe5	Mrs.	CA	00:
76127	37728	p043609	3f60494c61921b3b43ab61bdde2904df	Ms.	UT	00:



In [7]:

```
project_grade_category = []

for i in range(len(project_data)):
    a = project_data["project_grade_category"][i].replace(" ", "_")
    project_grade_category.append(a)

project_data.drop(['project_grade_category'], axis=1, inplace=True)
project_data["project_grade_category"] = project_grade_category
project_data.head(5)
```

Out[7]:

	Unnamed: 0	id	teacher_id	teacher_prefix	school_state	
55660	8393	p205479	2bf07ba08945e5d8b2a3f269b2b3cfe5	Mrs.	CA	00:
76127	37728	p043609	3f60494c61921b3b43ab61bdde2904df	Ms.	UT	00:
51140	74477	p189804	4a97f3a390bfe21b99cf5e2b81981c73	Mrs.	CA	00:
473	100660	p234804	cbc0e38f522143b86d372f8b43d4cff3	Mrs.	GA	00:
41558	33679	p137682	06f6e62e17de34fcf81020c77549e1d5	Mrs.	WA	01:

## 1.2 preprocessing of project\_subject\_categories

In [8]:

```

categories = list(project_data['project_subject_categories'].values)
# remove special characters from list of strings python: https://stackoverflow.com/a/473019
# https://www.geeksforgeeks.org/removing-stop-words-nltk-python/
# https://stackoverflow.com/questions/23669024/how-to-strip-a-specific-word-from-a-string
# https://stackoverflow.com/questions/8270092/remove-all-whitespace-in-a-string-in-python
cat_list = []
for i in 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", "Warmth", "
        if 'The' in j.split(): # this will split each of the category based on space "Math
            j=j.replace('The','') # if we have the words "The" we are going to replace it w
        j = j.replace(' ', '') # we are replacing all the ' '(space) with ''(empty) ex:"Math
        temp+=j.strip()+" " #" abc ".strip() will return "abc", remove the trailing spaces
        temp = temp.replace('&','_') # we are replacing the & value into
    cat_list.append(temp.strip())

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

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

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

```

## 1.3 preprocessing of project\_subject\_subcategories

In [9]:

```

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

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

sub_cat_list = []
for i in sub_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", "Warmth", "
        if 'The' in j.split(): # this will split each of the category based on space "Math
            j=j.replace('The', '') # if we have the words "The" we are going to replace it w
        j = j.replace(' ', '') # we are placing all the ' '(space) with ''(empty) ex:"Math
        temp +=j.strip()+" #" "abc ".strip() will return "abc", remove the trailing spaces
        temp = temp.replace('&', '_')
    sub_cat_list.append(temp.strip())

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

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

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

```

## Clean Titles (Text preprocessing)

In [10]:

```

# https://gist.github.com/sebleier/554280
# we are removing the words from the stop words list: 'no', 'nor', 'not'
stopwords= ['i', 'me', 'my', 'myself', 'we', 'our', 'ours', 'ourselves', 'you', "you're", "
    "you'll", "you'd", 'your', 'yours', 'yourself', 'yourselves', 'he', 'him', 'his',
    'she', "she's", 'her', 'hers', 'herself', 'it', "it's", 'its', 'itself', 'they',
    'theirs', 'themselves', 'what', 'which', 'who', 'whom', 'this', 'that', "that'll",
    'am', 'is', 'are', 'was', 'were', 'be', 'been', 'being', 'have', 'has', 'had',
    'did', 'doing', 'a', 'an', 'the', 'and', 'but', 'if', 'or', 'because', 'as', 'u
    'at', 'by', 'for', 'with', 'about', 'against', 'between', 'into', 'through', 'c
    'above', 'below', 'to', 'from', 'up', 'down', 'in', 'out', 'on', 'off', 'over',
    'then', 'once', 'here', 'there', 'when', 'where', 'why', 'how', 'all', 'any', '
    'most', 'other', 'some', 'such', 'only', 'own', 'same', 'so', 'than', 'too', 'v
    's', 't', 'can', 'will', 'just', 'don', "don't", 'should', "should've", 'now',
    've', 'y', 'ain', 'aren', "aren't", 'couldn', "couldn't", 'didn', "didn't", 'dc
    "hadn't", 'hasn', "hasn't", 'haven', "haven't", 'isn', "isn't", 'ma', 'mightn',
    "mustn't", 'needn', "needn't", 'shan', "shan't", 'shouldn', "shouldn't", 'wasn'
    'won', "won't", 'wouldn', "wouldn't"]

```

In [11]:

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

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

    # general
    phrase = re.sub(r"n't", " not", phrase)
    phrase = re.sub(r"'re", " are", phrase)
    phrase = re.sub(r"'s", " is", phrase)
    phrase = re.sub(r"'d", " would", phrase)
    phrase = re.sub(r"'ll", " will", phrase)
    phrase = re.sub(r"'t", " not", phrase)
    phrase = re.sub(r"'ve", " have", phrase)
    phrase = re.sub(r"'m", " am", phrase)
    return phrase
```

In [12]:

```
clean_titles = []

for titles in tqdm(project_data["project_title"]):
    title = decontracted(titles)
    title = title.replace('\\r', ' ')
    title = title.replace('\\n', ' ')
    title = title.replace('\\t', ' ')
    title = re.sub('[^A-Za-z0-9]+', ' ', title)
    title = ' '.join(f for f in title.split() if f not in stopwords)
    clean_titles.append(title.lower().strip())
```

100%|██████████| 109248/109248 [00:03<00:00, 34499.61it/s]

In [13]:

```
project_data["clean_titles"] = clean_titles
```

In [14]:

```
project_data.drop(['project_title'], axis=1, inplace=True)
```

## Feature "Number of Words in Title"

In [15]:

```

title_word_count = []
for a in project_data["clean_titles"] :
    b = len(a.split())
    title_word_count.append(b)

project_data["title_word_count"] = title_word_count
project_data.head(5)

```

Out[15]:

	Unnamed: 0	id	teacher_id	teacher_prefix	school_state	
55660	8393	p205479	2bf07ba08945e5d8b2a3f269b2b3cfe5	Mrs.	CA	00:
76127	37728	p043609	3f60494c61921b3b43ab61bdde2904df	Ms.	UT	00:
51140	74477	p189804	4a97f3a390bfe21b99cf5e2b81981c73	Mrs.	CA	00:
473	100660	p234804	cbc0e38f522143b86d372f8b43d4cff3	Mrs.	GA	00:
41558	33679	p137682	06f6e62e17de34fcf81020c77549e1d5	Mrs.	WA	01:

## 1.3 Text preprocessing

In [16]:

```

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

```

In [17]:

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

Out[17]:

Unnamed: 0	id	teacher_id	teacher_prefix	school_state
55660	8393 p205479	2bf07ba08945e5d8b2a3f269b2b3cfe5	Mrs.	CA
76127	37728 p043609	3f60494c61921b3b43ab61bdde2904df	Ms.	UT

## Clean Essays (Text preprocessing)

In [18]:

```
clean_essay = []

for ess in tqdm(project_data["essay"]):
    ess = decontracted(ess)
    ess = ess.replace('\\r', ' ')
    ess = ess.replace('\\\"', ' ')
    ess = ess.replace('\\n', ' ')
    ess = re.sub('[^A-Za-z0-9]+', ' ', ess)
    ess = ' '.join(f for f in ess.split() if f not in stopwords)
    clean_essay.append(ess.lower().strip())
```

100%|██████████| 109248/109248 [01:13<00:00, 1487.44it/s]

In [19]:

```
project_data["clean_essays"] = clean_essay
```

In [20]:

```
project_data.drop(['essay'], axis=1, inplace=True)
```

## Number of Words in Essay

In [21]:

```

essay_word_count = []
for ess in project_data["clean_essays"] :
    c = len(ess.split())
    essay_word_count.append(c)

project_data["essay_word_count"] = essay_word_count

project_data.head(5)

```

Out[21]:

	Unnamed: 0	id	teacher_id	teacher_prefix	school_state	
55660	8393	p205479	2bf07ba08945e5d8b2a3f269b2b3cfe5	Mrs.	CA	00:
76127	37728	p043609	3f60494c61921b3b43ab61bdde2904df	Ms.	UT	00:
51140	74477	p189804	4a97f3a390bfe21b99cf5e2b81981c73	Mrs.	CA	00:
473	100660	p234804	cbc0e38f522143b86d372f8b43d4cff3	Mrs.	GA	00:
41558	33679	p137682	06f6e62e17de34fcf81020c77549e1d5	Mrs.	WA	01:

## Train test Split

In [22]:

```

# train test split

from sklearn.model_selection import train_test_split

X_train, X_test, y_train, y_test = train_test_split(project_data, project_data['project_is_
X_train, X_cv, y_train, y_cv = train_test_split(X_train, y_train, test_size=0.33, stratify=

```

In [23]:

```
# printing some random reviews
print(project_data['clean_essays'].values[0])
print("="*50)
print(project_data['clean_essays'].values[150])
print("="*50)
print(project_data['clean_essays'].values[1000])
print("="*50)
print(project_data['clean_essays'].values[20000])
print("="*50)
print(project_data['clean_essays'].values[99999])
print("="*50)
```

i fortunate enough use fairy tale stem kits classroom well stem journals students really enjoyed i would love implement lakeshore stem kits classroom next school year provide excellent engaging stem lessons my students come variety backgrounds including language socioeconomic status many not lot experience science engineering kits give materials provide exciting opportunities students each month i try several science stem steam projects i would use kits robot help guide science instruction engaging meaningful ways i adapt kits current language arts pacing guide already teach material kits like tall tales paul bunyan johnny appleseed the following units taught next school year i implement kits magnets motion sink vs float robots i often get units not know if i teaching right way using right materials the kits give additional ideas strategies lessons prepare students science it challenging develop high quality science activities these kits give materials i need provide students science activities go along curriculum classroom although i things like magnets classroom i not know use effectively the kits provide right amount materials show use appropriate way

=====

i teach high school english students learning behavioral disabilities my students vary ability level however ultimate goal increase students literacy levels this includes reading writing communication levels i teach really dynamic group students however students face lot challenges my students live poverty dangerous neighborhood despite challenges i students desire defeat challenges my students learning disabilities currently performing grade level my students visual learners benefit classroom fulfills preferred learning style the materials i requesting allow students prepared classroom necessary supplies too often i challenged students come school unprepared class due economic challenges i want students able focus learning not able get school supplies the supplies last year students able complete written assignments maintain classroom journal the chart paper used make learning visual class create posters aid students learning the students access classroom printer the toner used print student work completed classroom chromebooks i want try remove barriers students learning create opportunities learning one biggest barriers students not resources get pens paper folders my students able increase literacy skills project

=====

life moves pretty fast if not stop look around awhile could miss movie ferris bueller day off think back remember grandparents how amazing would able flip book see day lives my second graders voracious readers they love read fiction nonfiction books their favorite characters include pete cat fly guy piggie elephant mercy watson they also love read insects space plants my students hungry bookworms my students eager learn read world around my kids love school like little sponges absorbing everything around their parents work long hours usually not see children my students usually cared grandparents family friend most students not someone speaks english home thus difficult students acquire language now think forward would not mean lot kids nieces nephews grandchildren able see day life today 30 years memories precious us able share memories future generations rewarding experience as part social studies



curriculum students learning changes time students studying photos learn community changed time in particular look photos study land buildings clothing schools changed time as culminating activity students capture slice history preserve scrap booking key important events young lives documented date location names students using photos home school create second grade memories their scrap books preserve unique stories future generations enjoy your donation project provide second graders opportunity learn social studies fun creative manner through scrapbooks children share story others historical documents rest lives

=====

a person person no matter small dr seuss i teach smallest students biggest enthusiasm learning my students learn many different ways using senses multiple intelligences i use wide range techniques help students succeed students class come variety different backgrounds makes wonderful sharing experiences cultures including native americans our school caring community successful learners seen collaborative student project based learning classroom kindergarten class love work hands materials many different opportunities practice skill mastered having social skills work cooperatively friends crucial aspect kindergarten curriculum montana perfect place learn agriculture nutrition my students love role play pretend kitchen early childhood classroom i several kids ask can try cooking real food i take idea create common core cooking lessons learn important math writing concepts cooking delicious healthy food snack time my students grounded appreciation work went making food knowledge ingredients came well healthy bodies this project would expand learning nutrition agricultural cooking recipes us peel apples make homemade applesauce make bread mix healthy plants classroom garden spring we also create cook books printed shared families students gain math literature skills well life long enjoyment healthy cooking nannan

=====

my classroom consists twenty two amazing sixth graders different cultures backgrounds they social bunch enjoy working partners working groups they hard working eager head middle school next year my job get ready make transition make smooth possible in order students need come school every day feel safe ready learn because getting ready head middle school i give lots choice choice sit work order complete assignments choice projects etc part students feeling safe ability come welcoming encouraging environment my room colorful atmosphere casual i want take ownership classroom all share together because time limited i want ensure get time enjoy best abilities currently twenty two desks differing sizes yet desks similar ones students use middle school we also kidney table crates seating i allow students choose spots working independently groups more often not move desks onto crates believe not proven successful making stay desks it i looking toward flexible seating option classroom the students look forward work time move around room i would like get rid constricting desks move toward fun seating options i requesting various seating students options sit currently i stool papasan chair i inherited previous sixth grade teacher well five milk crate seats i made i would like give options reduce competition good seats i also requesting two rugs not seating options make classroom welcoming appealing in order students able write complete work without desks i requesting class set clipboards finally due curriculum requires groups work together i requesting tables fold not using leave room flexible seating options i know seating options much excited coming school thank support making classroom one students remember forever nannan

In [24]:

```
# 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 [25]:

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

a person person no matter small dr seuss i teach smallest students biggest enthusiasm learning my students learn many different ways using senses multiple intelligences i use wide range techniques help students succeed students class come variety different backgrounds makes wonderful sharing experiences cultures including native americans our school caring community successful learners seen collaborative student project based learning classroom kindergarten class love work hands materials many different opportunities practice skill mastered having social skills work cooperatively friends crucial aspect kindergarten curriculum montana perfect place learn agriculture nutrition my students love role play pretend kitchen early childhood classroom i several kids ask can try cooking real food i take idea create common core cooking lessons learn important math writing concepts cooking delicious healthy food snack time my students grounded appreciation work went making food knowledge ingredients came well healthy bodies this project would expand learning nutrition agricultural cooking recipes us peel apples make homemade applesauce make bread mix healthy plants classroom garden spring we also create cook books printed shared families students gain math literature skills well life long enjoyment healthy cooking nannan

=====

In [26]:

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

a person person no matter small dr seuss i teach smallest students biggest enthusiasm learning my students learn many different ways using senses multiple intelligences i use wide range techniques help students succeed students class come variety different backgrounds makes wonderful sharing experiences cultures including native americans our school caring community successful learners seen collaborative student project based learning classroom kindergarten class love work hands materials many different opportunities practice skill mastered having social skills work cooperatively friends crucial aspect kindergarten curriculum montana perfect place learn agriculture nutrition my students love role play pretend kitchen early childhood classroom i several kids ask can try cooking real food i take idea create common core cooking lessons learn important math writing concepts cooking delicious healthy food snack time my students grounded appreciation work went making food knowledge ingredients came well healthy bodies this project would expand learning nutrition agricultural cooking recipes us peel apples make homemade applesauce make bread mix healthy plants classroom garden spring we also create cook books printed shared families students gain math literature skills well life long enjoyment healthy cooking nannan

In [27]:

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

a person person no matter small dr seuss i teach smallest students biggest enthusiasm learning my students learn many different ways using senses multiple intelligences i use wide range techniques help students succeed students class come variety different backgrounds makes wonderful sharing experiences cultures including native americans our school caring community successful learners seen collaborative student project based learning classroom kindergarten class love work hands materials many different opportunities practice skill mastered having social skills work cooperatively friends crucial aspect kindergarten curriculum montana perfect place learn agriculture nutrition my students love role play pretend kitchen early childhood classroom i several kids ask can try cooking real food i take idea create common core cooking lessons learn important math writing concepts cooking delicious healthy food snack time my students grounded appreciation work went making food knowledge ingredients came well healthy bodies this project would expand learning nutrition agricultural cooking recipes us peel apples make homemade applesauce make bread mix healthy plants classroom garden spring we also create cook books printed shared families students gain math literature skills well life long enjoyment healthy cooking nannan

In [28]:

```
# https://gist.github.com/sebleier/554280
# we are removing the words from the stop words list: 'no', 'nor', 'not'
stopwords= ['i', 'me', 'my', 'myself', 'we', 'our', 'ours', 'ourselves', 'you', "you're", "
you'll", "you'd", 'your', 'yours', 'yourself', 'yourselves', 'he', 'him', 'his', 'she', "she's", 'her', 'hers', 'herself', 'it', "it's", 'its', 'itself', 'they', 'theirs', 'themselves', 'what', 'which', 'who', 'whom', 'this', 'that', "that'll", 'am', 'is', 'are', 'was', 'were', 'be', 'been', 'being', 'have', 'has', 'had', 'did', 'doing', 'a', 'an', 'the', 'and', 'but', 'if', 'or', 'because', 'as', 'until', 'at', 'by', 'for', 'with', 'about', 'against', 'between', 'into', 'through', 'down', 'above', 'below', 'to', 'from', 'up', 'down', 'in', 'out', 'on', 'off', 'over', 'then', 'once', 'here', 'there', 'when', 'where', 'why', 'how', 'all', 'any', 'each', 'most', 'other', 'some', 'such', 'only', 'own', 'same', 'so', 'than', 'too', 'very', 's', 't', 'can', 'will', 'just', 'don', "don't", 'should', "should've", 'now', 've', 'y', 'ain', 'aren', "aren't", 'couldn', "couldn't", 'didn', "didn't", 'doesn', "doesn't", 'hadn't', 'hasn', "hasn't", 'haven', "haven't", 'isn', "isn't", 'ma', 'mightn', "mightn't", 'mustn't', 'needn', "needn't", 'shan', "shan't", 'shouldn', "shouldn't", 'wasn', 'won', "won't", 'wouldn', "wouldn't"]
```

## Preprocessed Train data (Essay)

In [29]:

```
# Combining all the above students
from tqdm import tqdm
preprocessed_essays_train = []
# tqdm is for printing the status bar
for sentence in tqdm(X_train['clean_essays'].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 for e in sent.split() if e.lower() not in stopwords)
    preprocessed_essays_train.append(sent.lower().strip())
```

100%|██████████| 49041/49041 [00:23<00:00, 2055.75it/s]

In [30]:

```
# after preprocessing
preprocessed_essays_train[20000]
```

Out[30]:

'students range academic levels honors students 95th percentile struggling students 1st percentile teach 50 boys 25 girls conduct learning style inventory students beginning year find learn best identify individual learning style seat groups varied learning styles setting good examples others good role model student expectation classroom school focuses giving students leadership roles classroom within school leadership roles given qualified student not outstanding student everyone job completing work accurately behavior warrant position course let struggling student feel accomplished give job know part classroom community much kid makes often times incentive getting classroom job brings best performance everyone school wide literacy goal every child read one million words april 2017 kids easy goal reach simply matter get one month two kids may well told needed swim across atlantic inconceivable somewhere along way either never pushed read given exposure reading material perhaps never priority home whatever reason truly bane existence goal classroom teacher help child find inner reader least inner listener audio books many different reasons children struggle read hate read job classroom teacher help succeed find way good book order find way reach reaching students getting interested books look differently students natural born readers simply need material keep reading however no longer live world handing student book enough read children learn differently realize students need read books kids discuss book move chapters children need technical aspect remain engaged like audio books research know audio books offer much originally thought important something offer everyone nannan'

## Preprocessed Test data (Essay)

In [31]:

```
preprocessed_essays_test = []
# tqdm is for printing the status bar
for sentence in tqdm(X_test['clean_essays'].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 for e in sent.split() if e.lower() not in stopwords)
    preprocessed_essays_test.append(sent.lower().strip())
```

100%|██████████| 36052/36052 [00:17<00:00, 2057.81it/s]

In [32]:

```
preprocessed_essays_test[0]
```

Out[32]:

'special needs teacher work many students grades 3 5 included general education classroom students varying disabilities work students autism learning disabilities students love learn sometimes need different approach understand concepts retain important skills taught class often need hands materials make learning engaging meaningful see tangible representations learning student s work hard day day learn fundamentals reading need exposure different materials order cement understanding well fun way outlet learning using reading games reinforcing skills already learning class fun interactive way students relate not students learning playing also able practice social skills completing tasks involving turn taking problem solving times conflict resolution n annan'

## Preprocessed Cross Validation data (essay)

In [33]:

```
preprocessed_essays_cv = []
# tqdm is for printing the status bar
for sentence in tqdm(X_cv['clean_essays'].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 for e in sent.split() if e.lower() not in stopwords)
    preprocessed_essays_cv.append(sent.lower().strip())
```

100%|██████████| 24155/24155 [00:11<00:00, 2050.99it/s]

In [34]:

```
preprocessed_essays_cv[0]
```

Out[34]:

'quote try live change wish see world try relay message kiddos constantly practicing growth mindset struggles persevere students student heart every morning read school contract reminder follow code conduct work together heterogeneous partnerships groups day focus bunch time collaborative conversations academic area keeping task building language positive take active stance learning class finds calming art time hope try integrate curriculum paintbrushes water colors would also like class use art projects gifts take home learn basics shading water colors also growing artists items improve classroom give students may not feel successful academic area feel understand area geometry fractions painting build equity students representing math social studies science vehicle visual arts students able express art'

## 1.4 Preprocessing of `project\_title`

### Preprocessing of Project Title for Train data

In [35]:

```
# similarly you can preprocess the titles also
preprocessed_titles_train = []

for titles in tqdm(X_train["clean_titles"]):
    title = decontracted(titles)
    title = title.replace('\\r', ' ')
    title = title.replace('\\\"', ' ')
    title = title.replace('\\n', ' ')
    title = re.sub('[^A-Za-z0-9]+', ' ', title)
    title = ' '.join(f for f in title.split() if f not in stopwords)
    preprocessed_titles_train.append(title.lower().strip())
```

100%|██████████| 49041/49041 [00:01<00:00, 35957.45it/s]

In [36]:

```
preprocessed_titles_train[0]
```

Out[36]:

'no place like chrome'

## Preprocessing of Project Title for Test data

In [37]:

```
preprocessed_titles_test = []

for titles in tqdm(X_test["clean_titles"]):
    title = decontracted(titles)
    title = title.replace('\\r', ' ')
    title = title.replace('\\\"', ' ')
    title = title.replace('\\n', ' ')
    title = re.sub('[^A-Za-z0-9]+', ' ', title)
    title = ' '.join(f for f in title.split() if f not in stopwords)
    preprocessed_titles_test.append(title.lower().strip())
```

100%|██████████| 36052/36052 [00:01<00:00, 35537.63it/s]

In [38]:

```
preprocessed_titles_test[0]
```

Out[38]:

'fun games reading'

## Preprocessing of Project Title for CV data

In [39]:

```
preprocessed_titles_cv = []

for titles in tqdm(X_cv["clean_titles"]):
    title = decontracted(titles)
    title = title.replace('\\r', ' ')
    title = title.replace('\\\"', ' ')
    title = title.replace('\\n', ' ')
    title = re.sub('[^A-Za-z0-9]+', ' ', title)
    title = ' '.join(f for f in title.split() if f not in stopwords)
    preprocessed_titles_cv.append(title.lower().strip())
```

100%|██████████| 24155/24155 [00:00<00:00, 35233.97it/s]

In [40]:

```
preprocessed_titles_cv[0]
```

Out[40]:

```
'visual art integration 04 28 16'
```

## 1.5 Preparing data for models

In [41]:

```
project_data.columns
```

Out[41]:

```
Index(['Unnamed: 0', 'id', 'teacher_id', 'teacher_prefix', 'school_state',
      'Date', 'project_essay_1', 'project_essay_2', 'project_essay_3',
      'project_essay_4', 'project_resource_summary',
      'teacher_number_of_previously_posted_projects', 'project_is_approved',
      'project_grade_category', 'clean_categories', 'clean_subcategories',
      'clean_titles', 'title_word_count', 'clean_essays', 'essay_word_count'],
      dtype='object')
```

we are going to consider

- school\_state : categorical data
- clean\_categories : categorical data
- clean\_subcategories : categorical data
- project\_grade\_category : categorical data
- teacher\_prefix : categorical data
- project\_title : text data
- text : text data
- project\_resource\_summary: text data (optional)
- quantity : numerical (optional)
- teacher\_number\_of\_previously\_posted\_projects : numerical
- price : numerical



## 1.5.1 Vectorizing Categorical data

- <https://www.appliedaicourse.com/course/applied-ai-course-online/lessons/handling-categorical-and-numerical-features/> (<https://www.appliedaicourse.com/course/applied-ai-course-online/lessons/handling-categorical-and-numerical-features/>)

In [42]:

```
# we use count vectorizer to convert the values into one
from sklearn.feature_extraction.text import CountVectorizer

vectorizer_proj = CountVectorizer(vocabulary=list(sorted_cat_dict.keys()), lowercase=False,
vectorizer_proj.fit(X_train['clean_categories'].values)

categories_one_hot_train = vectorizer_proj.transform(X_train['clean_categories'].values)
categories_one_hot_test = vectorizer_proj.transform(X_test['clean_categories'].values)
categories_one_hot_cv = vectorizer_proj.transform(X_cv['clean_categories'].values)

print(vectorizer_proj.get_feature_names())

print("Shape of matrix of Train data after one hot encoding ",categories_one_hot_train.shape)
print("Shape of matrix of Test data after one hot encoding ",categories_one_hot_test.shape)
print("Shape of matrix of CV data after one hot encoding ",categories_one_hot_cv.shape)

['Care_Hunger', 'Math_Science', 'History_Civics', 'Music_Arts', 'Warmth', 'SpecialNeeds', 'AppliedLearning', 'Literacy_Language', 'Health_Sports']
Shape of matrix of Train data after one hot encoding (49041, 9)
Shape of matrix of Test data after one hot encoding (36052, 9)
Shape of matrix of CV data after one hot encoding (24155, 9)
```

In [43]:

```
# we use count vectorizer to convert the values into one
vectorizer_sub_proj = CountVectorizer(vocabulary=list(sorted_sub_cat_dict.keys()), lowercase=True)
vectorizer_sub_proj.fit(X_train['clean_subcategories'].values)

sub_categories_one_hot_train = vectorizer_sub_proj.transform(X_train['clean_subcategories'].values)
sub_categories_one_hot_test = vectorizer_sub_proj.transform(X_test['clean_subcategories'].values)
sub_categories_one_hot_cv = vectorizer_sub_proj.transform(X_cv['clean_subcategories'].values)

print(vectorizer_sub_proj.get_feature_names())

print("Shape of matrix of Train data after one hot encoding ", sub_categories_one_hot_train.shape)
print("Shape of matrix of Test data after one hot encoding ", sub_categories_one_hot_test.shape)
print("Shape of matrix of Cross Validation data after one hot encoding ", sub_categories_one_hot_cv.shape)
```

```
['Mathematics', 'SocialSciences', 'Gym_Fitness', 'FinancialLiteracy', 'EnvironmentalScience', 'CommunityService', 'VisualArts', 'Warmth', 'History_Geography', 'NutritionEducation', 'Health_Wellness', 'Extracurricular', 'Literature_Writing', 'PerformingArts', 'ForeignLanguages', 'AppliedSciences', 'EarlyDevelopment', 'Health_LifeScience', 'College_CareerPrep', 'Music', 'Civics_Government', 'TeamSports', 'Economics', 'Literacy', 'Care_Hunger', 'ESL', 'Other', 'SpecialNeeds', 'CharacterEducation', 'ParentInvolvement']
Shape of matrix of Train data after one hot encoding (49041, 30)
Shape of matrix of Test data after one hot encoding (36052, 30)
Shape of matrix of Cross Validation data after one hot encoding (24155, 30)
```

In [44]:

```
# you can do the similar thing with state, teacher_prefix and project_grade_category also
my_counter = Counter()
for state in project_data['school_state'].values:
    my_counter.update(state.split())
```

In [45]:

```
school_state_cat_dict = dict(my_counter)
sorted_school_state_cat_dict = dict(sorted(school_state_cat_dict.items(), key=lambda kv: kv[1], reverse=True))
```

In [46]:

```
## we use count vectorizer to convert the values into one hot encoded features

vectorizer_states = CountVectorizer(vocabulary=list(sorted_school_state_cat_dict.keys()), 1
vectorizer_states.fit(X_train['school_state'].values)

school_state_categories_one_hot_train = vectorizer_states.transform(X_train['school_state'])
school_state_categories_one_hot_test = vectorizer_states.transform(X_test['school_state'].v
school_state_categories_one_hot_cv = vectorizer_states.transform(X_cv['school_state'].value

print(vectorizer_states.get_feature_names())

print("Shape of matrix of Train data after one hot encoding ",school_state_categories_one_h
print("Shape of matrix of Test data after one hot encoding ",school_state_categories_one_hc
print("Shape of matrix of Cross Validation data after one hot encoding ",school_state_categ
```

```
['ID', 'NY', 'WA', 'RI', 'NM', 'MI', 'HI', 'IA', 'AR', 'NH', 'WY', 'NE', 'N
C', 'KY', 'UT', 'OK', 'IN', 'CO', 'AZ', 'FL', 'WV', 'SD', 'GA', 'WI', 'CA',
'SC', 'KS', 'DC', 'MO', 'DE', 'VT', 'MS', 'AL', 'NV', 'ME', 'OR', 'TX', 'M
A', 'CT', 'NJ', 'MT', 'MD', 'MN', 'TN', 'OH', 'PA', 'IL', 'VA', 'ND', 'LA',
'AK']
```

```
Shape of matrix of Train data after one hot encoding (49041, 51)
```

```
Shape of matrix of Test data after one hot encoding (36052, 51)
```

```
Shape of matrix of Cross Validation data after one hot encoding (24155, 51)
```

In [47]:

```
my_counter = Counter()
for project_grade in project_data['project_grade_category'].values:
    my_counter.update(project_grade.split())
```

In [48]:

```
project_grade_cat_dict = dict(my_counter)
sorted_project_grade_cat_dict = dict(sorted(project_grade_cat_dict.items(), key=lambda kv:
```

In [49]:

```

## we use count vectorizer to convert the values into one hot encoded features

vectorizer_grade = CountVectorizer(vocabulary=list(sorted_project_grade_cat_dict.keys()), 1
vectorizer_grade.fit(X_train['project_grade_category'].values)

project_grade_categories_one_hot_train = vectorizer_grade.transform(X_train['project_grade_
project_grade_categories_one_hot_test = vectorizer_grade.transform(X_test['project_grade_ca
project_grade_categories_one_hot_cv = vectorizer_grade.transform(X_cv['project_grade_catego

print(vectorizer_grade.get_feature_names())

print("Shape of matrix of Train data after one hot encoding ",project_grade_categories_one_
print("Shape of matrix of Test data after one hot encoding ",project_grade_categories_one_h
print("Shape of matrix of Cross Validation data after one hot encoding ",project_grade_cate

['Grades_PreK-2', 'Grades_6-8', 'Grades_9-12', 'Grades_3-5']
Shape of matrix of Train data after one hot encoding (49041, 4)
Shape of matrix of Test data after one hot encoding (36052, 4)
Shape of matrix of Cross Validation data after one hot encoding (24155, 4)

```

In [50]:

```

my_counter = Counter()
for teacher_prefix in project_data['teacher_prefix'].values:
    teacher_prefix = str(teacher_prefix)
    my_counter.update(teacher_prefix.split())

```

In [51]:

```

teacher_prefix_cat_dict = dict(my_counter)
sorted_teacher_prefix_cat_dict = dict(sorted(teacher_prefix_cat_dict.items(), key=lambda kv

```

In [52]:

```

## we use count vectorizer to convert the values into one hot encoded features
## Unlike the previous Categories this category returns a
## ValueError: np.nan is an invalid document, expected byte or unicode string.
## The link below explains how to tackle such discrepancies.
## https://stackoverflow.com/questions/39303912/tfidfvectorizer-in-scikit-learn-valueerror-

vectorizer_teacher = CountVectorizer(vocabulary=list(sorted_teacher_prefix_cat_dict.keys()))
vectorizer_teacher.fit(X_train['teacher_prefix'].values.astype("U"))

teacher_prefix_categories_one_hot_train = vectorizer_teacher.transform(X_train['teacher_pre
teacher_prefix_categories_one_hot_test = vectorizer_teacher.transform(X_test['teacher_prefi
teacher_prefix_categories_one_hot_cv = vectorizer_teacher.transform(X_cv['teacher_prefix']).

print(vectorizer_teacher.get_feature_names())

print("Shape of matrix after one hot encoding ",teacher_prefix_categories_one_hot_train.sha
print("Shape of matrix after one hot encoding ",teacher_prefix_categories_one_hot_test.shap
print("Shape of matrix after one hot encoding ",teacher_prefix_categories_one_hot_cv.shape)

```

```

['nan', 'Dr.', 'Teacher', 'Ms.', 'Mrs.', 'Mr.']
Shape of matrix after one hot encoding (49041, 6)
Shape of matrix after one hot encoding (36052, 6)
Shape of matrix after one hot encoding (24155, 6)

```

## 1.5.2 Vectorizing Text data

### a) Bag of words Train Data (Essays)

In [53]:

```

# We are considering only the words which appeared in at least 10 documents(rows or project
vectorizer_bow_essay = CountVectorizer(min_df=10)

vectorizer_bow_essay.fit(preprocessed_essays_train)

text_bow_train = vectorizer_bow_essay.transform(preprocessed_essays_train)

print("Shape of matrix after one hot encoding ",text_bow_train.shape)

```

```

Shape of matrix after one hot encoding (49041, 12054)

```

### b) Bag of words Test Data (Essays)

In [54]:

```

text_bow_test = vectorizer_bow_essay.transform(preprocessed_essays_test)
print("Shape of matrix after one hot encoding ",text_bow_test.shape)

```

```

Shape of matrix after one hot encoding (36052, 12054)

```

## c) Bag of words CV Data (Essays)

In [55]:

```
text_bow_cv = vectorizer_bow_essay.transform(preprocessed_essays_cv)
print("Shape of matrix after one hot encoding ",text_bow_cv.shape)
```

Shape of matrix after one hot encoding (24155, 12054)

## d) Bag of words train Data (Titles)

In [56]:

```
# you can vectorize the title also
# before you vectorize the title make sure you preprocess it
vectorizer_bow_title = CountVectorizer(min_df=10)

vectorizer_bow_title.fit(preprocessed_titles_train)

title_bow_train = vectorizer_bow_title.transform(preprocessed_titles_train)

print("Shape of matrix after one hot encoding ",title_bow_train.shape)
```

Shape of matrix after one hot encoding (49041, 1968)

## e) Bag of words Test Data (Titles)

In [57]:

```
title_bow_test = vectorizer_bow_title.transform(preprocessed_titles_test)
print("Shape of matrix after one hot encoding ",title_bow_test.shape)
```

Shape of matrix after one hot encoding (36052, 1968)

## f) Bag of words Data (Titles)

In [58]:

```
title_bow_cv = vectorizer_bow_title.transform(preprocessed_titles_cv)
print("Shape of matrix after one hot encoding ",title_bow_cv.shape)
```

Shape of matrix after one hot encoding (24155, 1968)

### 1.5.2.2 TFIDF vectorizer

## a) TFIDF vectorizer Train Data (Essays)

In [59]:

```
from sklearn.feature_extraction.text import TfidfVectorizer

vectorizer_tfidf_essay = TfidfVectorizer(min_df=10)
vectorizer_tfidf_essay.fit(preprocessed_essays_train)

text_tfidf_train = vectorizer_tfidf_essay.transform(preprocessed_essays_train)

print("Shape of matrix after one hot encoding ",text_tfidf_train.shape)
```

Shape of matrix after one hot encoding (49041, 12054)

## b) TFIDF vectorizer Test Data (Essays)

In [60]:

```
text_tfidf_test = vectorizer_tfidf_essay.transform(preprocessed_essays_test)
print("Shape of matrix after one hot encoding ",text_tfidf_test.shape)
```

Shape of matrix after one hot encoding (36052, 12054)

## c) TFIDF vectorizer CV Data (Essays)

In [61]:

```
text_tfidf_cv = vectorizer_tfidf_essay.transform(preprocessed_essays_cv)
print("Shape of matrix after one hot encoding ",text_tfidf_cv.shape)
```

Shape of matrix after one hot encoding (24155, 12054)

## c) TFIDF vectorizer Train Data (Titles)

In [62]:

```
vectorizer_tfidf_titles = TfidfVectorizer(ngram_range=(2,2), min_df=10)

vectorizer_tfidf_titles.fit(preprocessed_titles_train)
title_tfidf_train = vectorizer_tfidf_titles.transform(preprocessed_titles_train)

print("Shape of matrix after one hot encoding ",title_tfidf_train.shape)
```

Shape of matrix after one hot encoding (49041, 1241)

## d) TFIDF vectorizer Test Data (Titles)

In [63]:

```
title_tfidf_test = vectorizer_tfidf_titles.transform(preprocessed_titles_test)
print("Shape of matrix after one hot encoding ",title_tfidf_test.shape)
```

Shape of matrix after one hot encoding (36052, 1241)

## e) TFIDF vectorizer CV Data (Titles)

In [64]:

```
title_tfidf_cv = vectorizer_tfidf_titles.transform(preprocessed_titles_cv)
print("Shape of matrix after one hot encoding ",title_tfidf_cv.shape)
```

Shape of matrix after one hot encoding (24155, 1241)

## C) Using Pretrained Models : AVG W2V



In [65]:

```
# Reading glove vectors in python: https://stackoverflow.com/a/38230349/4084039
def loadGloveModel(gloveFile):
    print ("Loading Glove Model")
    f = open(gloveFile,'r', encoding="utf8")
    model = {}
    for line in tqdm(f):
        splitLine = line.split()
        word = splitLine[0]
        embedding = np.array([float(val) for val in splitLine[1:]])
        model[word] = embedding
    print ("Done.",len(model)," words loaded!")
    return model

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

words = []
for i in preprocessed_essays_train :
    words.extend(i.split(' '))

print("all the words in the corpus", len(words))
words = set(words)
print("the unique words in the corpus", len(words))

inter_words = set(model.keys()).intersection(words)
print("The number of words that are present in both glove vectors and our corpus", \
      len(inter_words), "(" ,np.round(len(inter_words)/len(words)*100,3), "%)")

words_corpus = {}
words_glove = set(model.keys())
for i in words:
    if i in words_glove:
        words_corpus[i] = model[i]
print("word 2 vec length", len(words_corpus))

# stronging variables into pickle files python: http://www.jessicayung.com/how-to-use-pickl
```

Loading Glove Model

```
-----
FileNotFoundError                                Traceback (most recent call last)
<ipython-input-65-40504d584192> in <module>
     12         return model
     13
--> 14 model = loadGloveModel('glove.42B.300d.txt')
     15
     16

<ipython-input-65-40504d584192> in loadGloveModel(gloveFile)
      2 def loadGloveModel(gloveFile):
      3     print ("Loading Glove Model")
--> 4     f = open(gloveFile,'r', encoding="utf8")
      5     model = {}
```

```
6     for line in tqdm(f):
```

```
FileNotFoundError: [Errno 2] No such file or directory: 'glove.42B.300d.txt'
```

In [66]:

```
# stronging variables into pickle files python: http://www.jessicayung.com/how-to-use-pickl
# make sure you have the glove_vectors file
with open('glove_vectors', 'rb') as f:
    model = pickle.load(f)
    glove_words = set(model.keys())
```

## Train Essay

In [67]:

```
# average Word2Vec
# compute average word2vec for each review.

avg_w2v_vectors_train = [];

for sentence in tqdm(X_train["clean_essays"]): # for each review/sentence
    vector = np.zeros(300) # as word vectors are of zero length
    cnt_words = 0; # num of words with a valid vector in the sentence/review
    for word in sentence.split(): # for each word in a review/sentence
        if word in glove_words:
            vector += model[word]
            cnt_words += 1
    if cnt_words != 0:
        vector /= cnt_words
    avg_w2v_vectors_train.append(vector)

print(len(avg_w2v_vectors_train))
print(len(avg_w2v_vectors_train[0]))
```

```
100%|██████████| 49041/49041 [00:16<00:00, 2900.81it/s]
```

```
49041
```

```
300
```

## Test Essay

In [68]:

```

# average Word2Vec
# compute average word2vec for each review.

avg_w2v_vectors_test = [];

for sentence in tqdm(X_test["clean_essays"]): # for each review/sentence
    vector = np.zeros(300) # as word vectors are of zero length
    cnt_words = 0; # num of words with a valid vector in the sentence/review
    for word in sentence.split(): # for each word in a review/sentence
        if word in glove_words:
            vector += model[word]
            cnt_words += 1
    if cnt_words != 0:
        vector /= cnt_words
    avg_w2v_vectors_test.append(vector)

print(len(avg_w2v_vectors_test))
print(len(avg_w2v_vectors_test[0]))

```

100%|██████████| 36052/36052 [00:12<00:00, 2920.22it/s]

36052

300

## Cross validation Essay

In [69]:

```

avg_w2v_vectors_cv = [];

for sentence in tqdm(X_cv["clean_essays"]): # for each review/sentence
    vector = np.zeros(300) # as word vectors are of zero length
    cnt_words = 0; # num of words with a valid vector in the sentence/review
    for word in sentence.split(): # for each word in a review/sentence
        if word in glove_words:
            vector += model[word]
            cnt_words += 1
    if cnt_words != 0:
        vector /= cnt_words
    avg_w2v_vectors_cv.append(vector)

print(len(avg_w2v_vectors_cv))
print(len(avg_w2v_vectors_cv[0]))

```

100%|██████████| 24155/24155 [00:08<00:00, 2891.66it/s]

24155

300

## train Titles

In [70]:

```

avg_w2v_vectors_titles_train = []; # the avg-w2v for each sentence/review is stored in this
for sentence in tqdm(X_train["clean_titles"]): # for each title
    vector = np.zeros(300) # as word vectors are of zero length
    cnt_words = 0; # num of words with a valid vector in the sentence/review
    for word in sentence.split(): # for each word in a review/sentence
        if word in glove_words:
            vector += model[word]
            cnt_words += 1
    if cnt_words != 0:
        vector /= cnt_words
    avg_w2v_vectors_titles_train.append(vector)

print(len(avg_w2v_vectors_titles_train))
print(len(avg_w2v_vectors_titles_train[0]))

```

100%|██████████| 49041/49041 [00:00<00:00, 58925.56it/s]

49041

300

## Test Titles

In [71]:

```

avg_w2v_vectors_titles_test = []; # the avg-w2v for each sentence/review is stored in this
for sentence in tqdm(X_test["clean_titles"]): # for each title
    vector = np.zeros(300) # as word vectors are of zero length
    cnt_words = 0; # num of words with a valid vector in the sentence/review
    for word in sentence.split(): # for each word in a review/sentence
        if word in glove_words:
            vector += model[word]
            cnt_words += 1
    if cnt_words != 0:
        vector /= cnt_words
    avg_w2v_vectors_titles_test.append(vector)

print(len(avg_w2v_vectors_titles_test))
print(len(avg_w2v_vectors_titles_test[0]))

```

100%|██████████| 36052/36052 [00:00<00:00, 57293.30it/s]

36052

300

## CV Titles

In [72]:

```

avg_w2v_vectors_titles_cv = []; # the avg-w2v for each sentence/review is stored in this list
for sentence in tqdm(X_cv["clean_titles"]): # for each title
    vector = np.zeros(300) # as word vectors are of zero length
    cnt_words = 0; # num of words with a valid vector in the sentence/review
    for word in sentence.split(): # for each word in a review/sentence
        if word in glove_words:
            vector += model[word]
            cnt_words += 1
    if cnt_words != 0:
        vector /= cnt_words
    avg_w2v_vectors_titles_cv.append(vector)

print(len(avg_w2v_vectors_titles_cv))
print(len(avg_w2v_vectors_titles_cv[0]))

```

100%|██████████| 24155/24155 [00:00<00:00, 57484.91it/s]

24155

300

## D) Using Pretrained Models: TFIDF weighted W2V

### Train - Essays

In [73]:

```

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

```

In [74]:

```

# average Word2Vec
# compute average word2vec for each review.
tfidf_w2v_vectors_train = []; # the avg-w2v for each sentence/review is stored in this list
for sentence in tqdm(X_train["clean_essays"]): # for each review/sentence
    vector = np.zeros(300) # as word vectors are of zero length
    tf_idf_weight = 0; # num of words with a valid vector in the sentence/review
    for word in sentence.split(): # for each word in a review/sentence
        if (word in glove_words) and (word in tfidf_words):
            vec = model[word] # getting the vector for each word
            # here we are multiplying idf value(dictionary[word]) and the tf value((sentence.count(word)/len(sentence.split()))) # getting
            tf_idf = dictionary[word]*(sentence.count(word)/len(sentence.split())) # getting
            vector += (vec * tf_idf) # calculating tfidf weighted w2v
            tf_idf_weight += tf_idf
    if tf_idf_weight != 0:
        vector /= tf_idf_weight
    tfidf_w2v_vectors_train.append(vector)

print(len(tfidf_w2v_vectors_train))
print(len(tfidf_w2v_vectors_train[0]))

```

100%|██████████| 49041/49041 [01:50&lt;00:00, 445.75it/s]

49041

300

## Test essays

In [75]:

```

# compute average word2vec for each review.

tfidf_w2v_vectors_test = []; # the avg-w2v for each sentence/review is stored in this list
for sentence in tqdm(X_test["clean_essays"]): # for each review/sentence
    vector = np.zeros(300) # as word vectors are of zero length
    tf_idf_weight = 0; # num of words with a valid vector in the sentence/review
    for word in sentence.split(): # for each word in a review/sentence
        if (word in glove_words) and (word in tfidf_words):
            vec = model[word] # getting the vector for each word
            # here we are multiplying idf value(dictionary[word]) and the tf value((sentence.count(word)/len(sentence.split()))) # getting
            tf_idf = dictionary[word]*(sentence.count(word)/len(sentence.split())) # getting
            vector += (vec * tf_idf) # calculating tfidf weighted w2v
            tf_idf_weight += tf_idf
    if tf_idf_weight != 0:
        vector /= tf_idf_weight
    tfidf_w2v_vectors_test.append(vector)

print(len(tfidf_w2v_vectors_test))
print(len(tfidf_w2v_vectors_test[0]))

```

100%|██████████| 36052/36052 [01:19&lt;00:00, 455.48it/s]

36052

300

## CV essays

In [76]:

```
# compute average word2vec for each review.

tfidf_w2v_vectors_cv = []; # the avg-w2v for each sentence/review is stored in this list
for sentence in tqdm(X_cv["clean_essays"]): # for each review/sentence
    vector = np.zeros(300) # as word vectors are of zero length
    tf_idf_weight = 0; # num of words with a valid vector in the sentence/review
    for word in sentence.split(): # for each word in a review/sentence
        if (word in glove_words) and (word in tfidf_words):
            vec = model[word] # getting the vector for each word
            # here we are multiplying idf value(dictionary[word]) and the tf value((sentence.count(word)/len(sentence.split()))) # getting
            tf_idf = dictionary[word]*(sentence.count(word)/len(sentence.split())) # getting
            vector += (vec * tf_idf) # calculating tfidf weighted w2v
            tf_idf_weight += tf_idf
    if tf_idf_weight != 0:
        vector /= tf_idf_weight
    tfidf_w2v_vectors_cv.append(vector)

print(len(tfidf_w2v_vectors_cv))
print(len(tfidf_w2v_vectors_cv[0]))
```

100%|██████████| 24155/24155 [00:54<00:00, 447.06it/s]

24155

300

## Train Titles

In [77]:

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

In [78]:

```

# compute average word2vec for each review.

tfidf_w2v_vectors_titles_train = [];

for sentence in tqdm(X_train["clean_titles"]): # for each review/sentence
    vector = np.zeros(300) # as word vectors are of zero length
    tf_idf_weight = 0; # num of words with a valid vector in the sentence/review
    for word in sentence.split(): # for each word in a review/sentence
        if (word in glove_words) and (word in tfidf_words):
            vec = model[word] # getting the vector for each word
            # here we are multiplying idf value(dictionary[word]) and the tf value((sentence.count(word)/len(sentence.split()))) # getting
            tf_idf = dictionary[word]*(sentence.count(word)/len(sentence.split())) # getting
            vector += (vec * tf_idf) # calculating tfidf weighted w2v
            tf_idf_weight += tf_idf
    if tf_idf_weight != 0:
        vector /= tf_idf_weight
    tfidf_w2v_vectors_titles_train.append(vector)

print(len(tfidf_w2v_vectors_titles_train))
print(len(tfidf_w2v_vectors_titles_train[0]))

```

100%|██████████| 49041/49041 [00:01<00:00, 27796.09it/s]

49041

300

## Test Titles



In [79]:

```
# compute average word2vec for each review.

tfidf_w2v_vectors_titles_test = [];

for sentence in tqdm(X_test["clean_titles"]): # for each review/sentence
    vector = np.zeros(300) # as word vectors are of zero length
    tf_idf_weight = 0; # num of words with a valid vector in the sentence/review
    for word in sentence.split(): # for each word in a review/sentence
        if (word in glove_words) and (word in tfidf_words):
            vec = model[word] # getting the vector for each word
            # here we are multiplying idf value(dictionary[word]) and the tf value((sentence.count(word)/len(sentence.split()))) # getting
            tf_idf = dictionary[word]*(sentence.count(word)/len(sentence.split())) # getting
            vector += (vec * tf_idf) # calculating tfidf weighted w2v
            tf_idf_weight += tf_idf
    if tf_idf_weight != 0:
        vector /= tf_idf_weight
    tfidf_w2v_vectors_titles_test.append(vector)

print(len(tfidf_w2v_vectors_titles_test))
print(len(tfidf_w2v_vectors_titles_test[0]))
```

100%|██████████| 36052/36052 [00:01<00:00, 24941.13it/s]

36052

300

## CV Titles

In [80]:

```
# compute average word2vec for each review.

tfidf_w2v_vectors_titles_cv = [];

for sentence in tqdm(X_cv["clean_titles"]): # for each review/sentence
    vector = np.zeros(300) # as word vectors are of zero length
    tf_idf_weight = 0; # num of words with a valid vector in the sentence/review
    for word in sentence.split(): # for each word in a review/sentence
        if (word in glove_words) and (word in tfidf_words):
            vec = model[word] # getting the vector for each word
            # here we are multiplying idf value(dictionary[word]) and the tf value((sentence.count(word)/len(sentence.split()))) # getting
            tf_idf = dictionary[word]*(sentence.count(word)/len(sentence.split())) # getting
            vector += (vec * tf_idf) # calculating tfidf weighted w2v
            tf_idf_weight += tf_idf
    if tf_idf_weight != 0:
        vector /= tf_idf_weight
    tfidf_w2v_vectors_titles_cv.append(vector)

print(len(tfidf_w2v_vectors_titles_cv))
print(len(tfidf_w2v_vectors_titles_cv[0]))
```

100%|██████████| 24155/24155 [00:00<00:00, 26973.01it/s]

24155

300

## 1.5.3 Vectorizing Numerical features

### a) Price

In [81]:

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

In [82]:

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

In [83]:

```
# check this one: https://www.youtube.com/watch?v=0H0q0cLn3Z4&t=530s
# standardization sklearn: https://scikitlearn.org/stable/modules/generated/sklearn.preprocessing
from sklearn.preprocessing import StandardScaler
from sklearn.preprocessing import MinMaxScaler
from sklearn.preprocessing import Normalizer
from sklearn import preprocessing
price_scalar = MinMaxScaler()
price_scalar.fit(X_train['price'].values.reshape(-1,1)) # finding the mean and standarddevi
#print(f"Mean : {price_scalar.mean_[0]}, Standard deviation : {np.sqrt(price_scalar.var_[0])}")
# Now standardize the data with above maen and variance.
price_train = price_scalar.transform(X_train['price'].values.reshape(-1, 1))
price_train
# Now standardize the data with above maen and variance.
price_test = price_scalar.transform(X_test['price'].values.reshape(-1, 1))
price_test
# Now standardize the data with above maen and variance.
price_cv = price_scalar.transform(X_cv['price'].values.reshape(-1, 1))
price_cv
```

Out[83]:

```
array([[0.00684016],
       [0.03753334],
       [0.00442175],
       ...,
       [0.01682884],
       [0.0035406 ],
       [0.02043045]])
```

In [84]:

```
print("After vectorizations")
print(price_train.shape, y_train.shape)
print(price_cv.shape, y_cv.shape)
print(price_test.shape, y_test.shape)
```

```
After vectorizations
(49041, 1) (49041,)
(24155, 1) (24155,)
(36052, 1) (36052,)
```

## b) Quantity

In [85]:

```
price_scalar.fit(X_train['quantity'].values.reshape(-1,1)) # finding the mean and standard
#print(f"Mean : {price_scalar.mean_[0]}, Standard deviation : {np.sqrt(price_scalar.var_[0])}")
# Now standardize the data with above mean and variance.
quantity_train = price_scalar.transform(X_train['quantity'].values.reshape(-1, 1))
quantity_train
# Now standardize the data with above mean and variance.
quantity_cv = price_scalar.transform(X_cv['quantity'].values.reshape(-1, 1))
quantity_cv
# Now standardize the data with above mean and variance.
quantity_test = price_scalar.transform(X_test['quantity'].values.reshape(-1, 1))
quantity_test
```

Out[85]:

```
array([[0.00616333],
       [0.03389831],
       [0.         ],
       ...,
       [0.00924499],
       [0.01232666],
       [0.01078582]])
```

In [86]:

```
print("After vectorizations")
print(quantity_train.reshape, y_train.shape)
print(quantity_cv.shape, y_cv.shape)
print(quantity_test.shape, y_test.shape)
```

After vectorizations

```
<built-in method reshape of numpy.ndarray object at 0x7f5e14d770d0> (49041,)
(24155, 1) (24155,)
(36052, 1) (36052,)
```

## c) Number of Projects previously proposed by Teacher

In [87]:

```
price_scalar.fit(X_train['teacher_number_of_previously_posted_projects'].values.reshape(-1,
# print(f"Mean : {price_scalar.mean_[0]}, Standard deviation : {np.sqrt(price_scalar.var_[0])}
# Now standardize the data with above mean and variance.
prev_projects_train = price_scalar.transform(X_train['teacher_number_of_previously_posted_projects'])
prev_projects_train
# Now standardize the data with above mean and variance.
prev_projects_cv = price_scalar.transform(X_cv['teacher_number_of_previously_posted_projects'])
prev_projects_cv
# Now standardize the data with above mean and variance.
prev_projects_test = price_scalar.transform(X_test['teacher_number_of_previously_posted_projects'])
prev_projects_test
```

Out[87]:

```
array([[0.00221729],
       [0.         ],
       [0.11529933],
       ...,
       [0.09756098],
       [0.00665188],
       [0.00221729]])
```

In [88]:

```
print("After vectorizations")
print(prev_projects_train.shape, y_train.shape)
print(prev_projects_cv.shape, y_cv.shape)
print(prev_projects_test.shape, y_test.shape)
```

```
After vectorizations
(49041, 1) (49041,)
(24155, 1) (24155,)
(36052, 1) (36052,)
```

## Assignment 7: Decision Trees

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

- **Set 1:** categorical, numerical features + project\_title(BOW) + preprocessed\_eassay (BOW)
- **Set 2:** categorical, numerical features + project\_title(TFIDF)+ preprocessed\_eassay (TFIDF)
- **Set 3:** categorical, numerical features + project\_title(AVG W2V)+ preprocessed\_eassay (AVG W2V)
- **Set 4:** categorical, numerical features + project\_title(TFIDF W2V)+ preprocessed\_eassay (TFIDF W2V)

### 2. Hyper parameter tuning (best `depth` in range [4,6, 8, 9,10,12,14,17] , and the best `min\_samples\_split` in range [2,10,20,30,40,50])

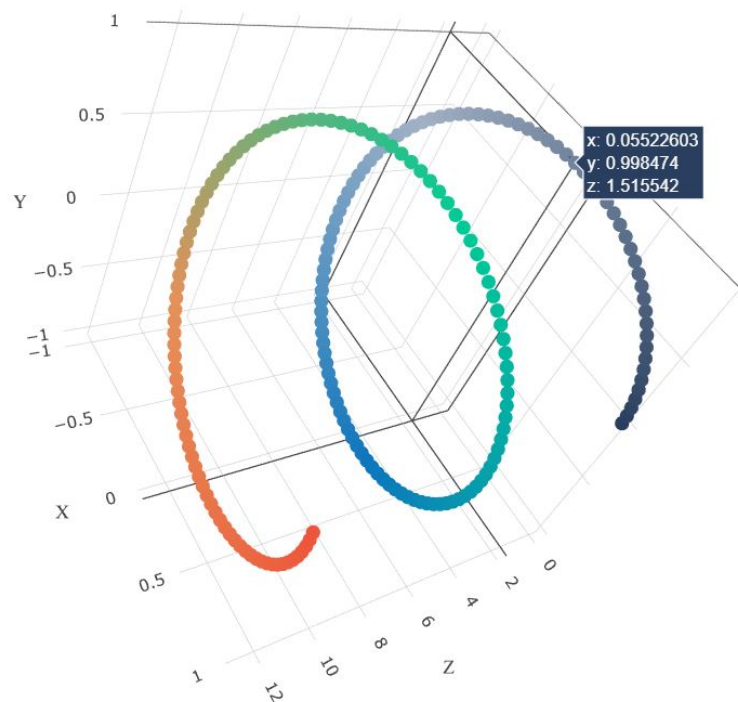
- Find the best hyper parameter which will give the maximum [AUC](https://www.appliedaicourse.com/course/applied-ai-course-online/lessons/receiver-operating-characteristic-curve-roc-curve-and-auc-1/) value
- Find the best hyper parameter using k-fold cross validation or simple cross validation data
-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

### 3. Graphviz

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

### 4. Representation of results

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

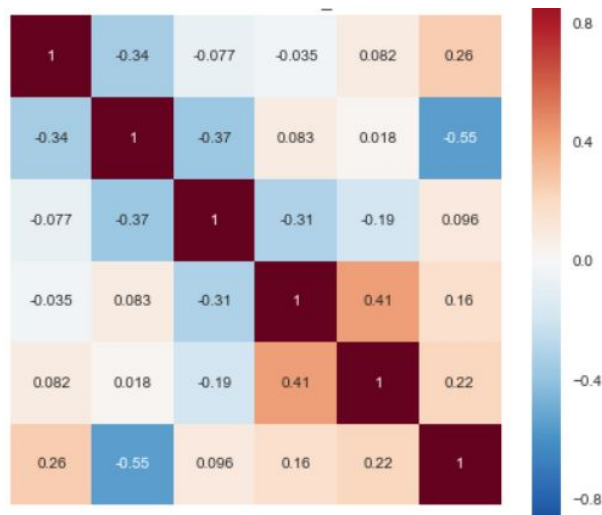


with X-axis as **min\_sample\_split**, Y-axis as **max\_depth**, and Z-axis as **AUC Score** , we have given the notebook which explains how to plot this 3d plot, you can find it in the same drive

[\*3d\\_scatter\\_plot.ipynb\*](#)

**or**

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



[seaborn heat maps \(https://seaborn.pydata.org/generated/seaborn.heatmap.html\)](https://seaborn.pydata.org/generated/seaborn.heatmap.html) with rows as **min\_sample\_split**, columns as **max\_depth**, and values inside the cell representing **AUC Score**

- You choose either of the plotting techniques out of 3d plot or heat map
- Once after you found the best hyper parameter, you need to train your model with it, and find the AUC on test data and plot the ROC curve on both train and test.



- Along with plotting ROC curve, you need to print the [confusion matrix \(https://www.appliedaicourse.com/course/applied-ai-course-online/lessons/confusion-matrix-tpr-fpr-fnr-tnr-1/\)](https://www.appliedaicourse.com/course/applied-ai-course-online/lessons/confusion-matrix-tpr-fpr-fnr-tnr-1/) with predicted and original labels of test data points



- Once after you plot the confusion matrix with the test data, get all the `false positive data points`
  - Plot the WordCloud [WordCloud \(https://www.geeksforgeeks.org/generating-word-cloud-python/\)](https://www.geeksforgeeks.org/generating-word-cloud-python/)
  - Plot the box plot with the `price` of these `false positive data points`
  - Plot the pdf with the `teacher\_number\_of\_previously\_posted\_projects` of these `false positive data points`

## 5. [Task-2]

- Select 5k best features from features of **Set 2** using `feature\_importances\_` (<https://scikit-learn.org/stable/modules/generated/sklearn.tree.DecisionTreeClassifier.html>), discard all the other remaining features and then apply any of the model of your choice i.e. (Decision tree, Logistic Regression, Linear SVM), you need to do hyperparameter tuning corresponding to the model you selected and procedure in step 2 and step 3

## 6. Conclusion

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



## Note: Data Leakage

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

## 2.1 Set 1: categorical, numerical features + project\_title(BOW) + preprocessed\_essay (BOW)

In [89]:

```
# merge two sparse matrices: https://stackoverflow.com/a/19710648/4084039
from scipy.sparse import hstack

X_tr = hstack((categories_one_hot_train, sub_categories_one_hot_train, school_state_categories_one_hot_train))
X_te = hstack((categories_one_hot_test, sub_categories_one_hot_test, school_state_categories_one_hot_test))
X_cr = hstack((categories_one_hot_cv, sub_categories_one_hot_cv, school_state_categories_one_hot_cv))
```

In [90]:

```
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

```
(49041, 14125) (49041,)
(24155, 14125) (24155,)
(36052, 14125) (36052,)
```

```
=====
=====
```

In [91]:

```
def batch_predict(clf, data):
    # roc_auc_score(y_true, y_score) the 2nd parameter should be probability estimates of the model
    # not the predicted outputs

    y_data_pred = []
    tr_loop = data.shape[0] - data.shape[0]%1000
    # consider you X_tr shape is 49041, then your cr_loop will be 49041 - 49041%1000 = 49000
    # in this for loop we will iterate until the last 1000 multiplier
    for i in range(0, tr_loop, 1000):
        y_data_pred.extend(clf.predict_proba(data[i:i+1000]))[:,1])
    # we will be predicting for the last data points
    y_data_pred.extend(clf.predict_proba(data[tr_loop:]))[:,1])

    return y_data_pred
```

### A) Gridsearch-cv

In [92]:

```
from sklearn.model_selection import GridSearchCV
from sklearn.tree import DecisionTreeClassifier
```



In [93]:

```
dt1 = DecisionTreeClassifier(class_weight = 'balanced')
parameters = {'max_depth': [1, 5, 10, 50, 100, 500, 1000], 'min_samples_split': [5, 10, 20,
clf1 = GridSearchCV(dt1, parameters, cv= 2, scoring='roc_auc',n_jobs=-1,return_train_score=
clf1.fit(X_tr, y_train)
```

Out[93]:

```
GridSearchCV(cv=2, error_score='raise-deprecating',
             estimator=DecisionTreeClassifier(class_weight='balanced',
                                              criterion='gini', max_depth=No
ne,
                                              max_features=None,
                                              max_leaf_nodes=None,
                                              min_impurity_decrease=0.0,
                                              min_impurity_split=None,
                                              min_samples_leaf=1,
                                              min_samples_split=2,
                                              min_weight_fraction_leaf=0.0,
                                              presort=False, random_state=No
ne,
                                              splitter='best'),
             iid='warn', n_jobs=-1,
             param_grid={'max_depth': [1, 5, 10, 50, 100, 500, 1000],
                         'min_samples_split': [5, 10, 20, 45, 75, 100, 135,
270,
                                              500]}},
             pre_dispatch='2*n_jobs', refit=True, return_train_score=True,
             scoring='roc_auc', verbose=0)
```

In [94]:

```
import seaborn as sns; sns.set()

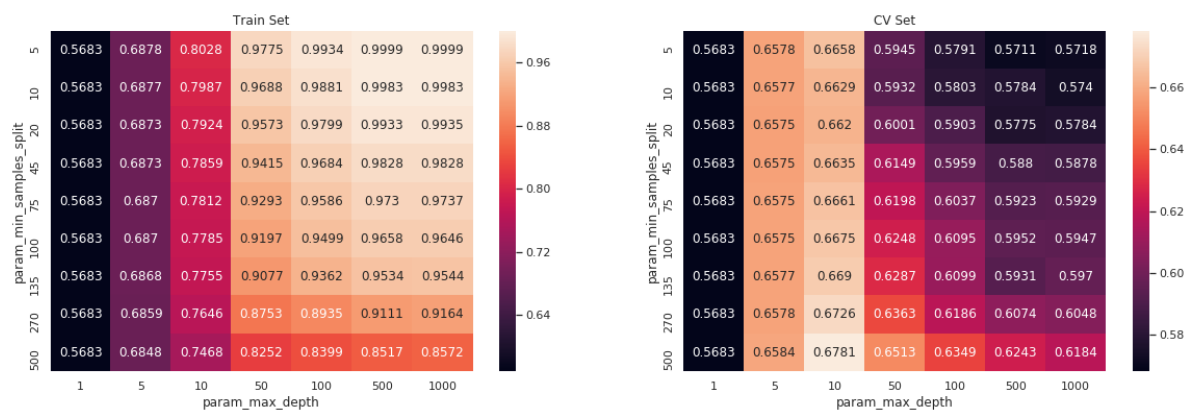
max_scores1 = pd.DataFrame(clf1.cv_results_).groupby(['param_min_samples_split', 'param_max_depth'])

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

sns.heatmap(max_scores1.mean_train_score, annot = True, fmt='.4g', ax=ax[0])
sns.heatmap(max_scores1.mean_test_score, annot = True, fmt='.4g', ax=ax[1])

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

plt.show()
```

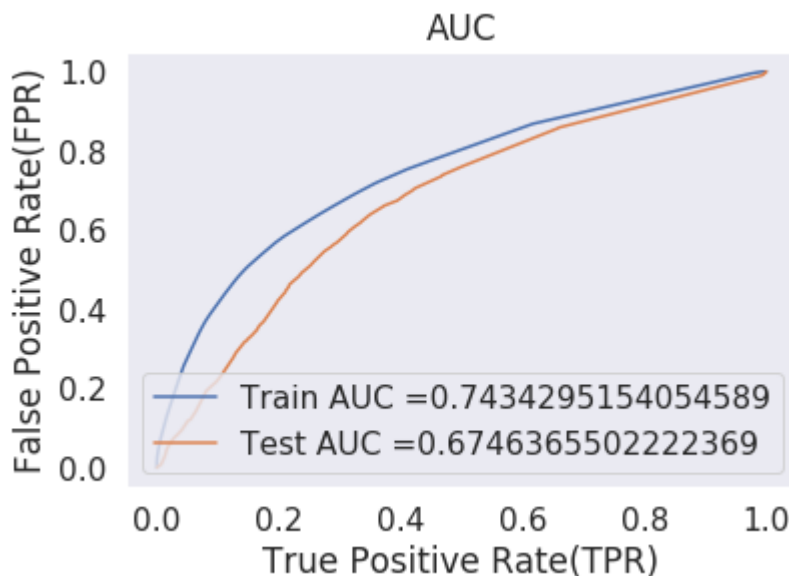


## B) Train model using the best hyper-parameter value

In [107]:

```
# https://scikitlearn.org/stable/modules/generated/sklearn.metrics.roc_curve.html#sklearn.m
from sklearn.metrics import roc_curve, auc
model = DecisionTreeClassifier(max_depth = 10, min_samples_split = 500, class_weight = 'bala
model.fit(X_tr, y_train)
clfV1=DecisionTreeClassifier (class_weight = 'balanced', max_depth=3, min_samples_split=500)
# for visulation
clfV1.fit(X_tr, y_train)
# roc_auc_score(y_true, y_score) the 2nd parameter should be probability estimates of the p
# not the predicted outputs
y_train_pred = batch_predict(model, X_tr)
y_test_pred = batch_predict(model, X_te)

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()
plt.show()
```



## D) Confusion Matrix

In [108]:

```
def predict(proba, threshold, fpr, tpr):

    t = threshold[np.argmax(fpr*(1-tpr))]

    # (tpr*(1-fpr)) will be maximum if your fpr is very low and tpr is very high

    print("the maximum value of tpr*(1-fpr)", max(tpr*(1-fpr)), "for threshold", np.round(t, 2))
    predictions = []
    global predictions1 # make it global
    for i in proba:
        if i >= t:
            predictions.append(1)
        else:
            predictions.append(0)
    predictions1 = predictions
    return predictions
```

## Train Data

In [109]:

```
from sklearn.metrics import confusion_matrix
print("Train confusion matrix")
print(confusion_matrix(y_train, predict(y_train_pred, tr_thresholds, train_fpr, train_fpr)))
```

Train confusion matrix

the maximum value of tpr\*(1-fpr) 0.2475040382279047 for threshold 0.36

```
[[ 4084  3342]
 [ 9334 32281]]
```

In [110]:

```
conf_matr_df_train_1 = pd.DataFrame(confusion_matrix(y_train, predict(y_train_pred, tr_thre
```

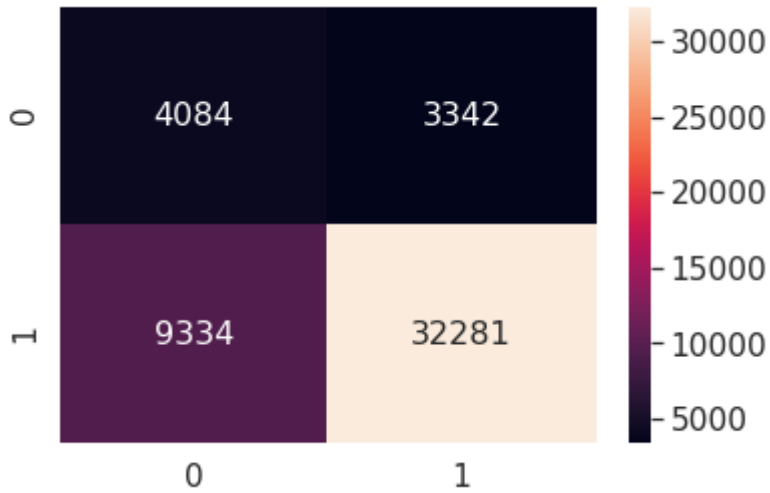
the maximum value of tpr\*(1-fpr) 0.2475040382279047 for threshold 0.36

In [111]:

```
sns.set(font_scale=1.4)#for label size
sns.heatmap(conf_matr_df_train_1, annot=True,annot_kws={"size": 16}, fmt='g')
```

Out[111]:

<matplotlib.axes.\_subplots.AxesSubplot at 0x7f5e140872b0>



## Test Data

In [112]:

```
print("Test confusion matrix")
print(confusion_matrix(y_test, predict(y_test_pred, tr_thresholds, test_fpr, test_fpr)))
```

Test confusion matrix

the maximum value of  $tpr \cdot (1 - fpr)$  0.24999294479210055 for threshold 0.36

```
[[ 2715  2744]
 [ 7295 23298]]
```

In [113]:

```
conf_matr_df_test_2 = pd.DataFrame(confusion_matrix(y_test, predict(y_test_pred, tr_threshc
```

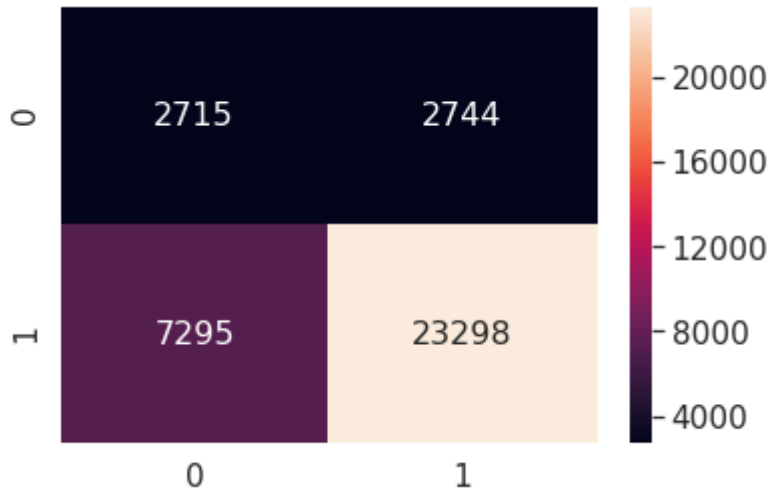
the maximum value of  $tpr \cdot (1 - fpr)$  0.24999294479210055 for threshold 0.36

In [114]:

```
sns.set(font_scale=1.4)#for label size
sns.heatmap(conf_matr_df_test_2, annot=True,annot_kws={"size": 16}, fmt='g')
```

Out[114]:

<matplotlib.axes.\_subplots.AxesSubplot at 0x7f5e13e20390>



## Feature aggregation

In [123]:

```
f1=vectorizer_proj.get_feature_names()
f2=vectorizer_sub_proj.get_feature_names()
f3= vectorizer_states.get_feature_names()
f4= vectorizer_grade.get_feature_names()
f5 = vectorizer_teacher.get_feature_names()
fbow= vectorizer_bow_essay.get_feature_names()
ftbow = vectorizer_bow_title.get_feature_names()
ftfidf = vectorizer_tfidf_essay.get_feature_names()
fttfidf = vectorizer_tfidf_titles.get_feature_names()

feature_agg_bow = f1 + f2 + f3 + f4 + f5 + fbow + ftbow
feature_agg_tfidf = f1 + f2 + f3 + f4 + f5 + ftfidf + fttfidf
# p is price, q is quantity, t is teacher previous year projects
feature_agg_bow.append('price')
feature_agg_tfidf.append('price')
feature_agg_bow.append('quantity')
feature_agg_tfidf.append('quantity')
feature_agg_bow.append('teacher_previous_projects')
feature_agg_tfidf.append('teacher_previous_projects')
```

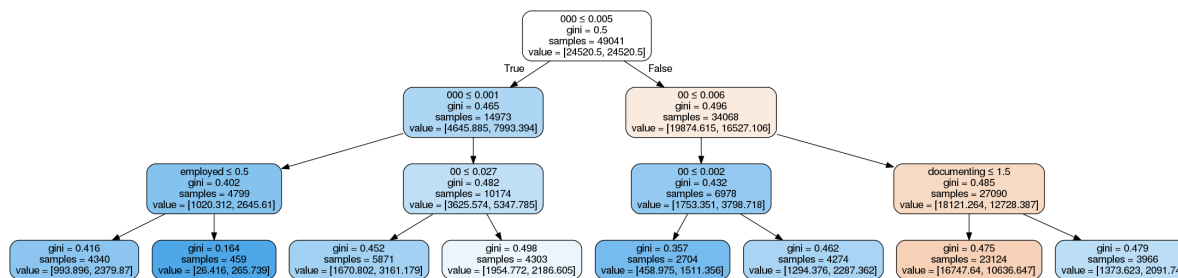
## Visualizing Decision Tree

In [124]:

```
import warnings
warnings.filterwarnings("ignore")
from sklearn.externals.six import StringIO
from IPython.display import Image
from sklearn.tree import export_graphviz
import pydotplus

dot_data = StringIO()
export_graphviz(clfV1, out_file=dot_data, filled=True, rounded=True, special_characters=True)
graph = pydotplus.graph_from_dot_data(dot_data.getvalue())
Image(graph.create_png())
```

Out[124]:



## False Positives

In [125]:

```
fpi = []

for i in range(len(y_test)) :
    if (y_test.values[i] == 0) & (predictions1[i] == 1) :
        fpi.append(i)
```

In [126]:

```
fp_essay1 = []
for i in fpi :
    fp_essay1.append(X_test['clean_essays'].values[i])
```

In [127]:

```
len(fp_essay1)
```

Out[127]:

2744

In [ ]:

## Word Cloud (False positives essay)

In [128]:

```
from wordcloud import WordCloud, STOPWORDS

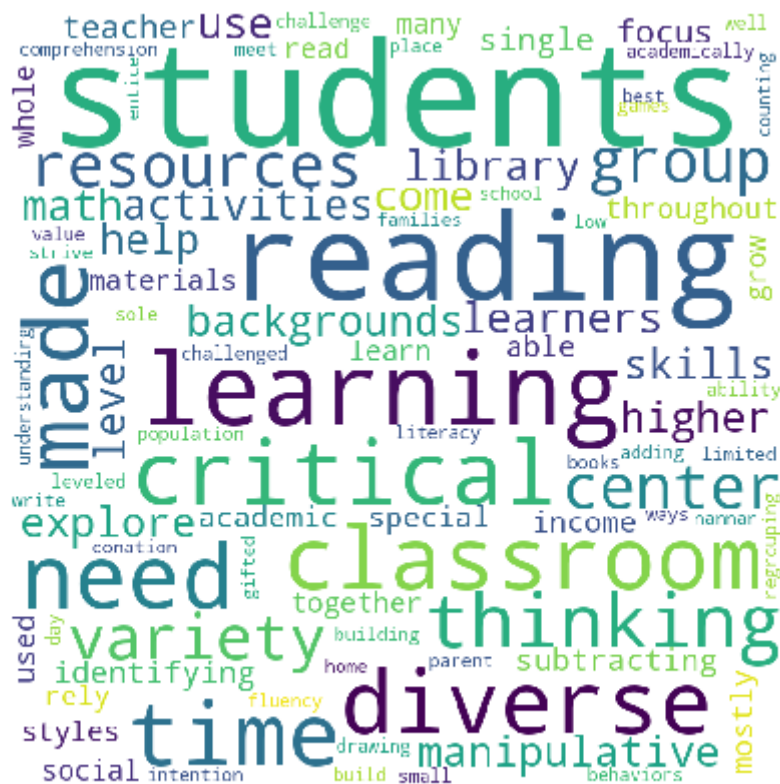
comment_words = ''
stopwords = set(STOPWORDS)

for val in fp_essay1 :
    val = str(val)
    tokens = val.split()
    for i in range(len(tokens)):
        tokens[i] = tokens[i].lower()
    for words in tokens :
        comment_words = comment_words + words + ' '

wordcloud = WordCloud(width = 800, height = 800, background_color = 'white', stopwords = stopwords, min_font_size = 10)

plt.figure(figsize = (6, 6), facecolor = None)
plt.imshow(wordcloud)
plt.axis("off")
plt.tight_layout(pad = 0)

plt.show()
```



## Building DataFrame of False Positives

In [129]:

```
cols = X_test.columns
X_test_falsePos1 = pd.DataFrame(columns=cols)
```



In [130]:

```
for i in fpi :
    X_test_falsePos1 = X_test_falsePos1.append(X_test.filter(items=[i], axis=0))
```

In [131]:

```
len(X_test_falsePos1)
```

Out[131]:

2744

In [132]:

```
X_test_falsePos1.head(2)
```

Out[132]:

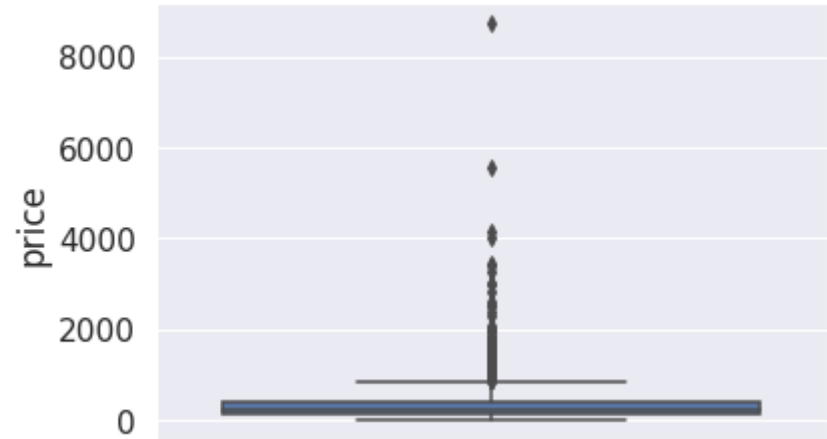
	Unnamed: 0	id	teacher_id	teacher_prefix	school_state	Dat
35	136188	p166914	dd32fbe60f0d5981c9f8822db9bfb80e	Ms.	NY	2016-10-01 13:53:20
65	77315	p007303	25cf951f507a0aaa6f251bbc02ab45a7	Mrs.	NC	2016-08-01 00:00:50

2 rows × 22 columns

## box plot for Price

In [133]:

```
graph = sns.boxplot(y='price', data=X_test_falsePos1)
```

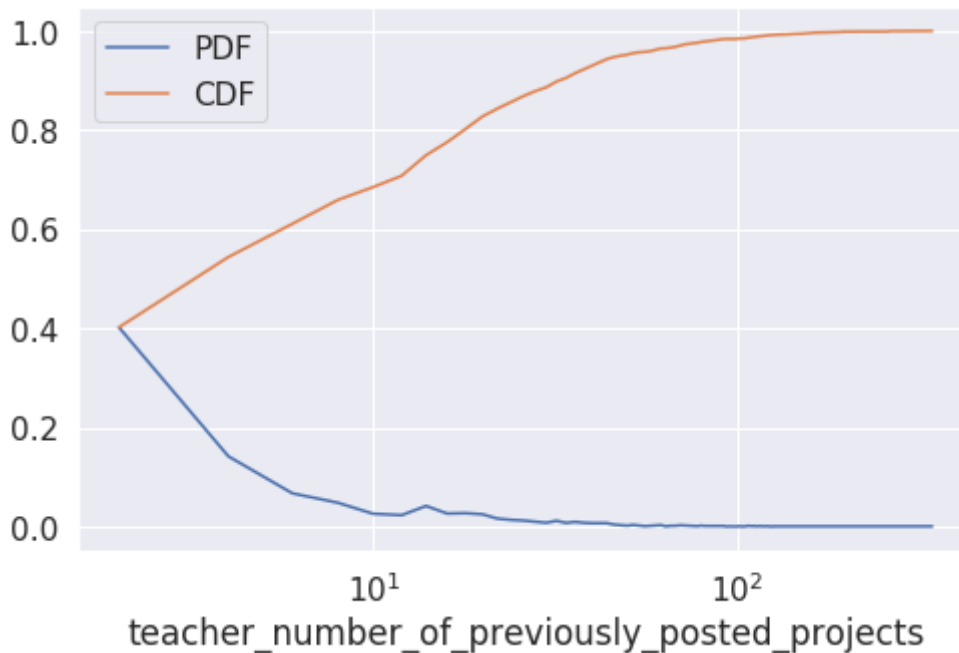


## PDF on teacher\_number\_of\_previously\_posted\_projects

In [134]:

```
plt.figure(figsize=(8,5))

counts, bin_edges = np.histogram(X_test_falsePos1['teacher_number_of_previously_posted_proj
pdf = counts/sum(counts)
cdf = np.cumsum(pdf)
pdfP, = plt.plot(bin_edges[1:], pdf)
cdfP, = plt.plot(bin_edges[1:], cdf)
plt.legend([pdfP, cdfP], ["PDF", "CDF"])
plt.xscale('log')
plt.xlabel('teacher_number_of_previously_posted_projects')
plt.show()
```



## Set 2 : categorical, numerical features + project\_title(TFIDF) + preprocessed\_essay (TFIDF)

In [135]:

```
# merge two sparse matrices: https://stackoverflow.com/a/19710648/4084039
from scipy.sparse import hstack

X_tr2 = hstack((categories_one_hot_train, sub_categories_one_hot_train, school_state_categor
X_te2 = hstack((categories_one_hot_test, sub_categories_one_hot_test, school_state_categorie
X_cr2 = hstack((categories_one_hot_cv, sub_categories_one_hot_cv, school_state_categories_or
```

In [136]:

```
print("Final Data matrix")
print(X_tr2.shape, y_train.shape)
print(X_cr2.shape, y_cv.shape)
print(X_te2.shape, y_test.shape)
```

```
Final Data matrix
(49041, 13398) (49041,)
(24155, 13398) (24155,)
(36052, 13398) (36052,)
```

## GridSearch CV

In [137]:

```
dt2 = DecisionTreeClassifier(class_weight = 'balanced')
parameters = {'max_depth': [1, 5, 10, 50, 100, 500, 1000], 'min_samples_split': [5, 10, 20,

clf2 = GridSearchCV(dt2, parameters, cv= 2, scoring='roc_auc', n_jobs=-1, return_train_score=
clf2.fit(X_tr2, y_train)
```

Out[137]:

```
GridSearchCV(cv=2, error_score='raise-deprecating',
             estimator=DecisionTreeClassifier(class_weight='balanced',
                                              criterion='gini', max_depth=No
ne,
                                              max_features=None,
                                              max_leaf_nodes=None,
                                              min_impurity_decrease=0.0,
                                              min_impurity_split=None,
                                              min_samples_leaf=1,
                                              min_samples_split=2,
                                              min_weight_fraction_leaf=0.0,
                                              presort=False, random_state=No
ne,
                                              splitter='best'),
             iid='warn', n_jobs=-1,
             param_grid={'max_depth': [1, 5, 10, 50, 100, 500, 1000],
                         'min_samples_split': [5, 10, 20, 45, 75, 100, 135,
270,
                                              500]},
             pre_dispatch='2*n_jobs', refit=True, return_train_score=True,
             scoring='roc_auc', verbose=0)
```

In [138]:

```
import seaborn as sns; sns.set()

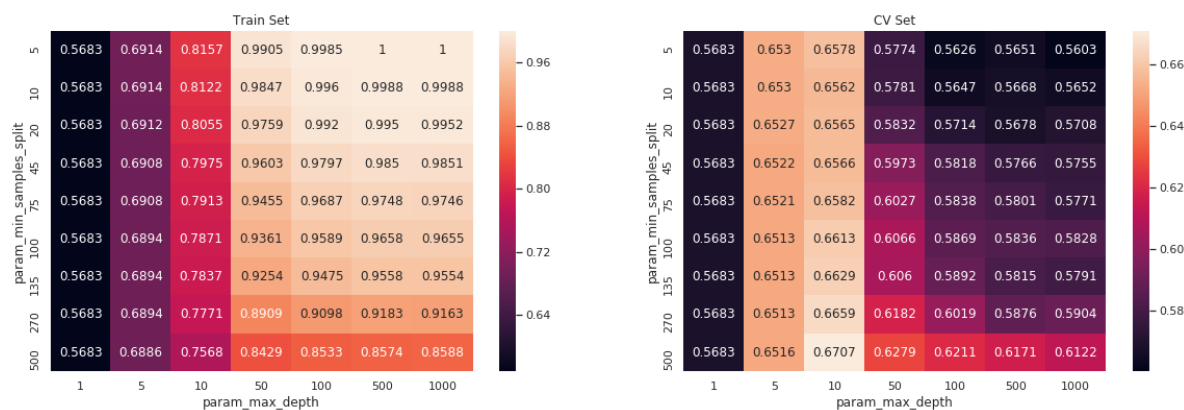
max_scores2 = pd.DataFrame(clf2.cv_results_).groupby(['param_min_samples_split', 'param_max_depth'])

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

sns.heatmap(max_scores2.mean_train_score, annot = True, fmt='.4g', ax=ax[0])
sns.heatmap(max_scores2.mean_test_score, annot = True, fmt='.4g', ax=ax[1])

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

plt.show()
```

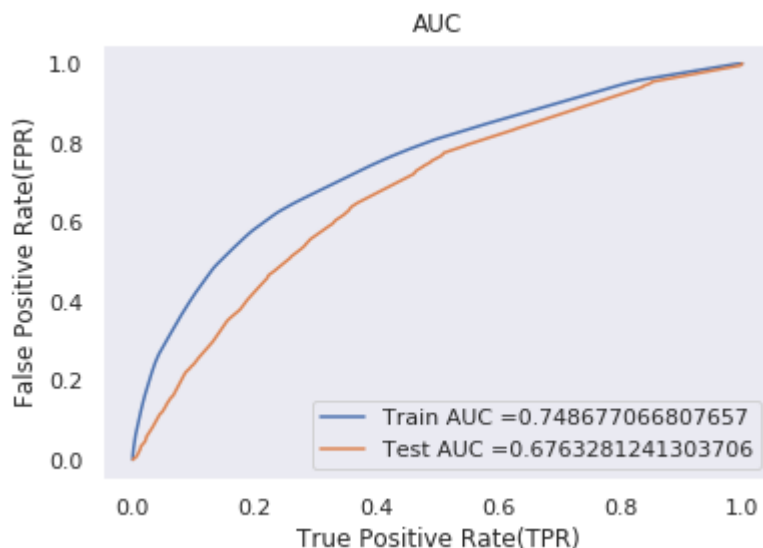


## Train model using the best hyper-parameter value

In [139]:

```
# https://scikitlearn.org/stable/modules/generated/sklearn.metrics.roc_curve.html#sklearn.m
from sklearn.metrics import roc_curve, auc
model = DecisionTreeClassifier(max_depth = 10, min_samples_split = 500, class_weight = 'balanced')
model.fit(X_tr2, y_train)
clfV2 = DecisionTreeClassifier(class_weight = 'balanced', max_depth=3, min_samples_split=500)
# for visulation
clfV2.fit(X_tr2, y_train)
# roc_auc_score(y_true, y_score) the 2nd parameter should be probability estimates of the p
# not the predicted outputs
y_train_pred = batch_predict(model, X_tr2)
y_test_pred = batch_predict(model, X_te2)

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()
plt.show()
```



## Confusion Matrix -Train data

In [140]:

```
print("Train confusion matrix")
print(confusion_matrix(y_train, predict(y_train_pred, tr_thresholds, train_fpr, train_tpr)))
```

Train confusion matrix

the maximum value of  $tpr \cdot (1 - fpr)$  0.2499886634136951 for threshold 0.329

```
[[ 3738  3688]
 [ 7959 33656]]
```

In [141]:

```
conf_matr_df_train_3 = pd.DataFrame(confusion_matrix(y_train, predict(y_train_pred, tr_thre
```

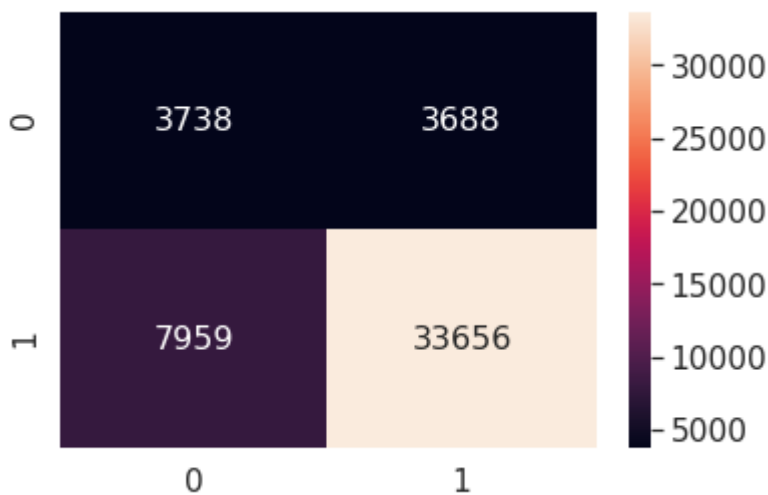
the maximum value of  $\text{tpr} \cdot (1 - \text{fpr})$  0.2499886634136951 for threshold 0.329

In [142]:

```
sns.set(font_scale=1.4)#for label size
sns.heatmap(conf_matr_df_train_3, annot=True,annot_kws={"size": 16}, fmt='g')
```

Out[142]:

<matplotlib.axes.\_subplots.AxesSubplot at 0x7f5e144124e0>



## Test Data

In [143]:

```
print("Test confusion matrix")
print(confusion_matrix(y_test, predict(y_test_pred, tr_thresholds, test_fpr, test_fpr)))
```

Test confusion matrix

the maximum value of  $\text{tpr} \cdot (1 - \text{fpr})$  0.24995026120376243 for threshold 0.384

```
[[ 2691  2768]
 [ 7099 23494]]
```

In [144]:

```
conf_matr_df_test_4 = pd.DataFrame(confusion_matrix(y_test, predict(y_test_pred, tr_threshc
```

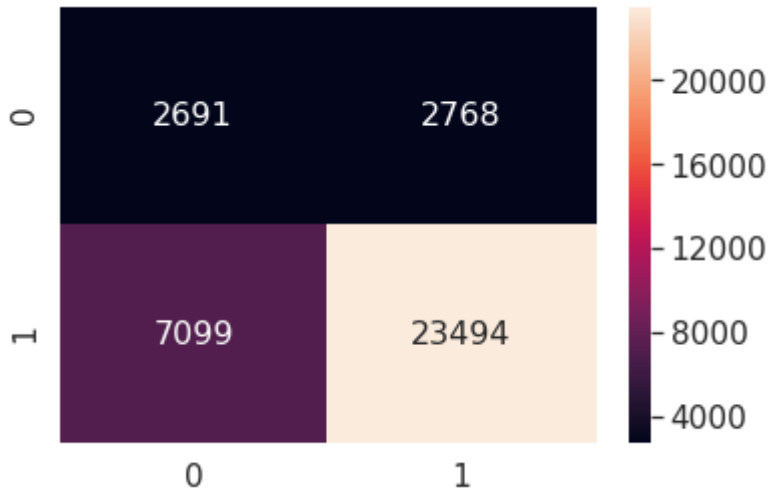
the maximum value of  $\text{tpr} \cdot (1 - \text{fpr})$  0.24995026120376243 for threshold 0.384

In [145]:

```
sns.set(font_scale=1.4)#for label size
sns.heatmap(conf_matr_df_test_4, annot=True,annot_kws={"size": 16}, fmt='g')
```

Out[145]:

&lt;matplotlib.axes.\_subplots.AxesSubplot at 0x7f5e138a45f8&gt;



## Visualizing Decision Tree

In [146]:

```
from sklearn.externals.six import StringIO
from IPython.display import Image
from sklearn.tree import export_graphviz
import pydotplus
```

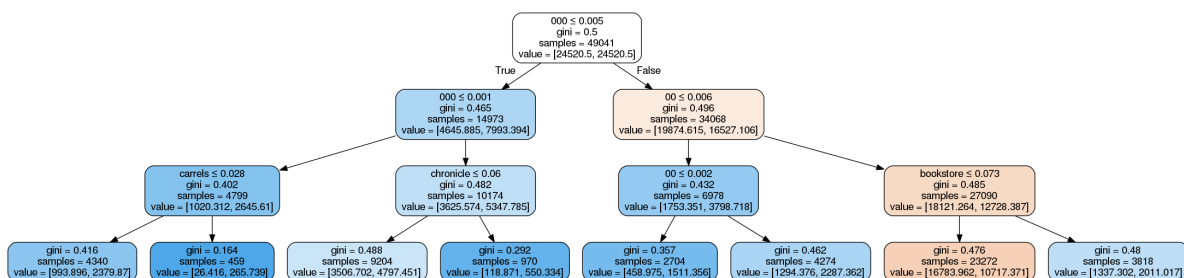
In [147]:

```
dot_data = StringIO()

#dt_feat_names = list(X_test.columns)
#dt_target_names = [str(s) for s in [0,1]]
export_graphviz(clfV2, out_file=dot_data, filled=True, rounded=True, special_characters=True)

graph = pydotplus.graph_from_dot_data(dot_data.getvalue())
Image(graph.create_png())
```

Out[147]:



## False Positives Retrieval

In [148]:

```
fpi = []

for i in range(len(y_test)) :
    if (y_test.values[i] == 0) & (predictions1[i] == 1) :
        fpi.append(i)

fp_essay1 = []
for i in fpi :
    fp_essay1.append(X_test['clean_essays'].values[i])
```

In [149]:

```
len(fp_essay1)
```

Out[149]:

2768

## Word Cloud (False positives essay)



In [150]:

```
from wordcloud import WordCloud, STOPWORDS

comment_words = ' '
stopwords = set(STOPWORDS)

for val in fp_essay1 :
    val = str(val)
    tokens = val.split()
    for i in range(len(tokens)):
        tokens[i] = tokens[i].lower()
    for words in tokens :
        comment_words = comment_words + words + ' '

wordcloud = WordCloud(width = 800, height = 800, background_color ='white', stopwords = stopwords, min_font_size = 10)

plt.figure(figsize = (6, 6), facecolor = None)
plt.imshow(wordcloud)
plt.axis("off")
plt.tight_layout(pad = 0)

plt.show()
```

In [151]:

```
cols = X_test.columns  
X_test_falsePos1 = pd.DataFrame(columns=cols)
```

In [152]:

```
for i in fpi :  
    X_test_falsePos1 = X_test_falsePos1.append(X_test.filter(items=[i], axis=0))
```

In [153]:

```
len(X_test_falsePos1)
```

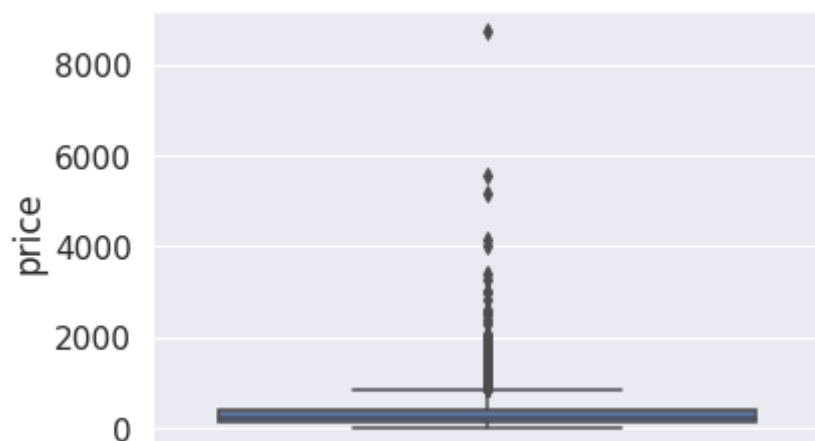
Out[153]:

2768

## box plot for Price

In [154]:

```
ax = sns.boxplot(y='price', data=X_test_falsePos1)
```

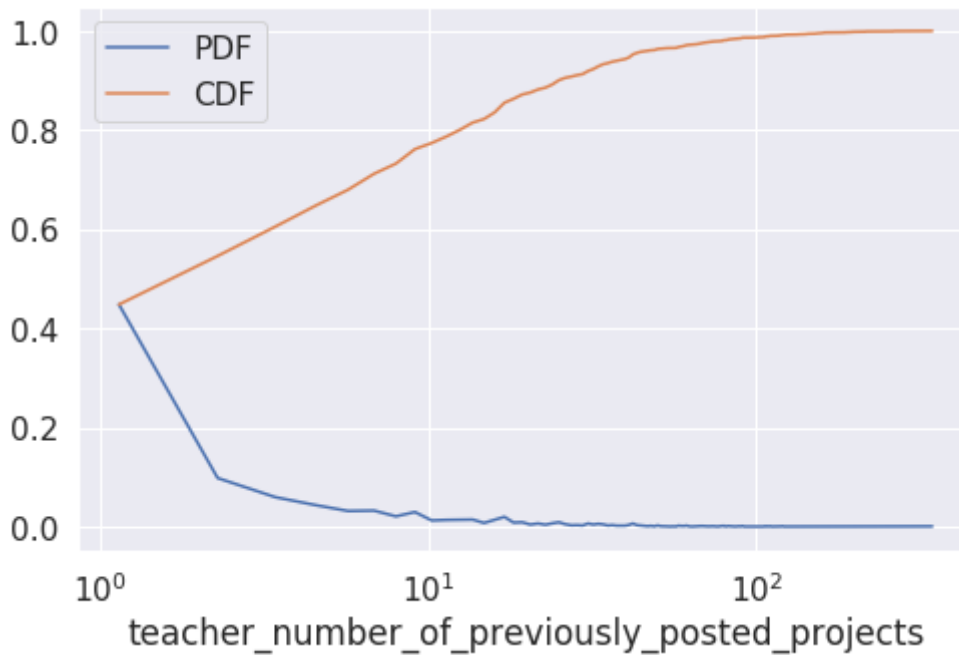


## PDF on teacher\_number\_of\_previously\_posted\_projects

In [155]:

```
plt.figure(figsize=(8,5))

counts, bin_edges = np.histogram(X_test_falsePos1['teacher_number_of_previously_posted_proj
pdf = counts/sum(counts)
cdf = np.cumsum(pdf)
pdfP, = plt.plot(bin_edges[1:], pdf)
cdfP, = plt.plot(bin_edges[1:], cdf)
plt.legend([pdfP, cdfP], ["PDF", "CDF"])
plt.xscale('log')
plt.xlabel('teacher_number_of_previously_posted_projects')
plt.show()
```



### Set 3 : Categorical, Numerical features + Project\_title(AVG W2V) + Preprocessed\_essay (AVG W2V)

In [156]:

```
# merge two sparse matrices: https://stackoverflow.com/a/19710648/4084039
from scipy.sparse import hstack

X_tr = hstack((categories_one_hot_train, sub_categories_one_hot_train, school_state_categori
X_te = hstack((categories_one_hot_test, sub_categories_one_hot_test, school_state_categories
X_cr = hstack((categories_one_hot_cv, sub_categories_one_hot_cv, school_state_categories_one
```

In [157]:

```
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)
```

```
Final Data matrix
(49041, 703) (49041,)
(24155, 703) (24155,)
(36052, 703) (36052,)
```

## Gridsearch CV

In [158]:

```
dt3 = DecisionTreeClassifier(class_weight = 'balanced')
parameters = {'max_depth': [1, 5, 10, 50, 100, 500, 1000], 'min_samples_split': [5, 10, 20,

clf3 = GridSearchCV(dt3, parameters, cv= 2, scoring='roc_auc',n_jobs=-1,return_train_score=
clf3.fit(X_tr, y_train)
```

Out[158]:

```
GridSearchCV(cv=2, error_score='raise-deprecating',
              estimator=DecisionTreeClassifier(class_weight='balanced',
                                                criterion='gini', max_depth=No
ne,
                                                max_features=None,
                                                max_leaf_nodes=None,
                                                min_impurity_decrease=0.0,
                                                min_impurity_split=None,
                                                min_samples_leaf=1,
                                                min_samples_split=2,
                                                min_weight_fraction_leaf=0.0,
                                                presort=False, random_state=No
ne,
                                                splitter='best'),
              iid='warn', n_jobs=-1,
              param_grid={'max_depth': [1, 5, 10, 50, 100, 500, 1000],
                           'min_samples_split': [5, 10, 20, 45, 75, 100, 135,
270,
                           500]},
              pre_dispatch='2*n_jobs', refit=True, return_train_score=True,
              scoring='roc_auc', verbose=0)
```

In [159]:

```
import seaborn as sns; sns.set()

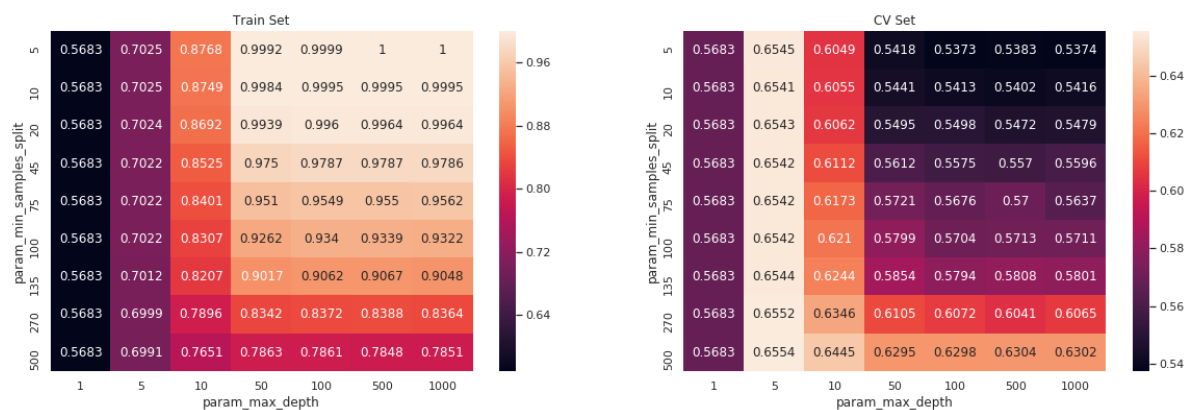
max_scores3 = pd.DataFrame(clf3.cv_results_).groupby(['param_min_samples_split', 'param_max_depth'])

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

sns.heatmap(max_scores3.mean_train_score, annot = True, fmt='.4g', ax=ax[0])
sns.heatmap(max_scores3.mean_test_score, annot = True, fmt='.4g', ax=ax[1])

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

plt.show()
```

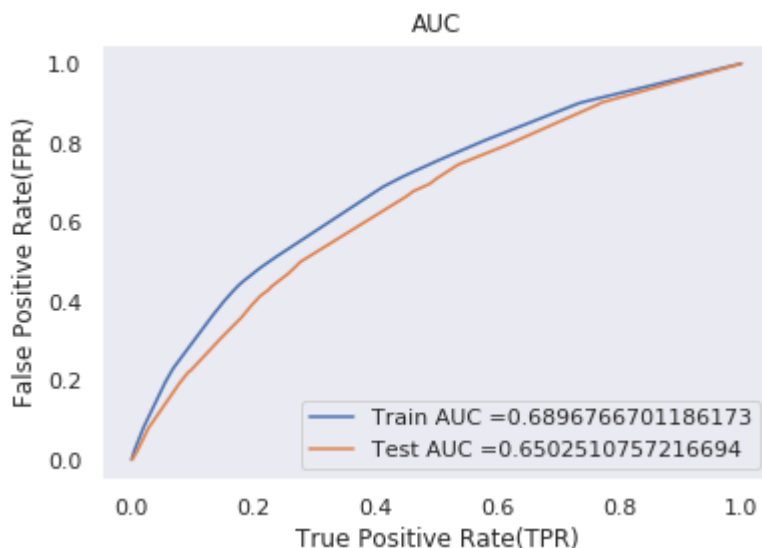


## B) Train the model using the best hyper parameter value

In [160]:

```
# https://scikitlearn.org/stable/modules/generated/sklearn.metrics.roc_curve.html#sklearn.m
from sklearn.metrics import roc_curve, auc
model = DecisionTreeClassifier(max_depth = 5, min_samples_split = 270, class_weight = 'balanced')
model.fit(X_tr, y_train)
clfV1 = DecisionTreeClassifier(class_weight = 'balanced', max_depth=2, min_samples_split=500)
# for visulation
clfV1.fit(X_tr, y_train)
# roc_auc_score(y_true, y_score) the 2nd parameter should be probability estimates of the p
# not the predicted outputs
y_train_pred = batch_predict(model, X_tr)
y_test_pred = batch_predict(model, X_te)

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()
plt.show()
```



## C) Confusion Matrix

### Train data

In [161]:

```
from sklearn.metrics import confusion_matrix
print("Train confusion matrix")
print(confusion_matrix(y_train, predict(y_train_pred, tr_thresholds, train_fpr, train_tpr)))
```

Train confusion matrix  
the maximum value of  $tpr \cdot (1 - fpr)$  0.2499999818661462 for threshold 0.411  
[[ 3714 3712]  
 [10258 31357]]

In [162]:

```
conf_matr_df_train_5 = pd.DataFrame(confusion_matrix(y_train, predict(y_train_pred, tr_thre
```

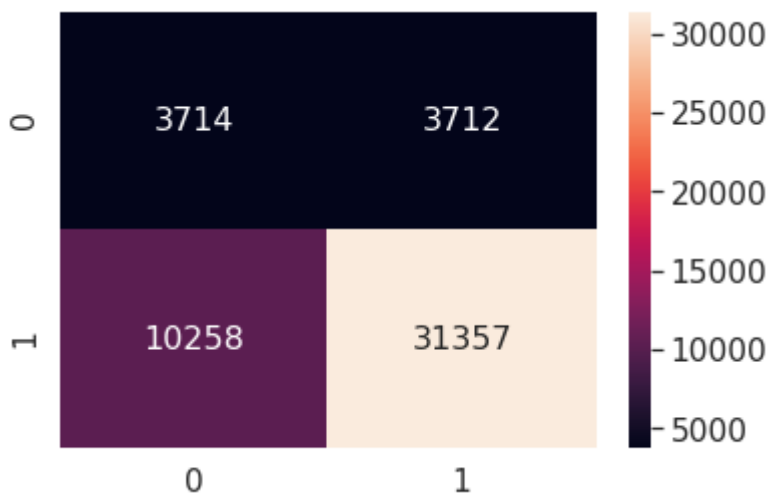
the maximum value of  $tpr \cdot (1-fpr)$  0.2499999818661462 for threshold 0.411

In [163]:

```
sns.set(font_scale=1.4)#for label size
sns.heatmap(conf_matr_df_train_5, annot=True,annot_kws={"size": 16}, fmt='g')
```

Out[163]:

<matplotlib.axes.\_subplots.AxesSubplot at 0x7f5e139735f8>



## Test data

In [164]:

```
print("Test confusion matrix")
print(confusion_matrix(y_test, predict(y_test_pred, tr_thresholds, test_fpr, test_fpr)))
```

Test confusion matrix

the maximum value of  $tpr \cdot (1-fpr)$  0.24999999161092998 for threshold 0.42

```
[[ 2729  2730]
 [ 8880 21713]]
```

In [165]:

```
conf_matr_df_test_6 = pd.DataFrame(confusion_matrix(y_test, predict(y_test_pred, tr_threshc
```

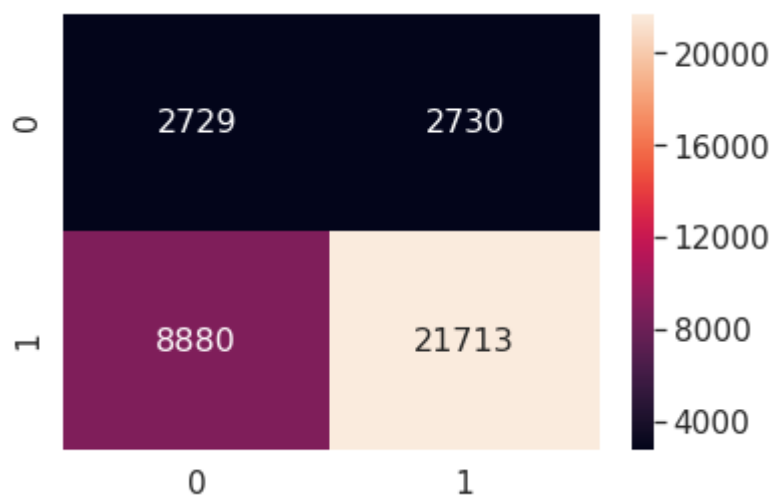
the maximum value of  $tpr \cdot (1-fpr)$  0.24999999161092998 for threshold 0.42

In [166]:

```
sns.set(font_scale=1.4)#for label size
sns.heatmap(conf_matr_df_test_6, annot=True,annot_kws={"size": 16}, fmt='g')
```

Out[166]:

<matplotlib.axes.\_subplots.AxesSubplot at 0x7f5e14349b38>



In [167]:

```
fpi = []

for i in range(len(y_test)) :
    if (y_test.values[i] == 0) & (predictions1[i] == 1) :
        fpi.append(i)

fp_essay1 = []
for i in fpi :
    fp_essay1.append(X_test['clean_essays'].values[i])
```

In [168]:

```
len(fp_essay1)
```

Out[168]:

2730



In [169]:

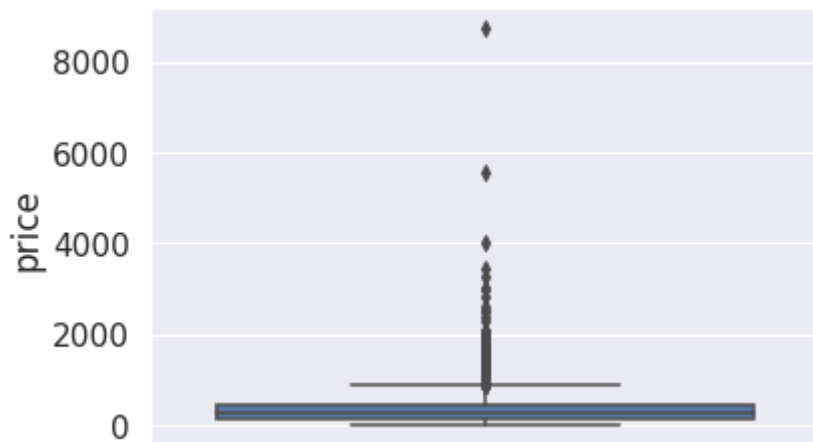
```
#Word cloud of essay
from wordcloud import WordCloud, STOPWORDS
comment_words = ' '
stopwords = set(STOPWORDS)
for val in fp_essay1 :
    val = str(val)
    tokens = val.split()
    for i in range(len(tokens)):
        tokens[i] = tokens[i].lower()
    for words in tokens :
        comment_words = comment_words + words + ' '
wordcloud = WordCloud(width = 800, height = 800, background_color = 'white', stopwords =
stopwords,min_font_size = 10).generate(comment_words)
plt.figure(figsize = (6, 6), facecolor = None)
plt.imshow(wordcloud)
plt.axis("off")
plt.tight_layout(pad = 0)
plt.show()
```

In [171]:

```
sns.boxplot(y='price', data=X_test_falsePos1)
```

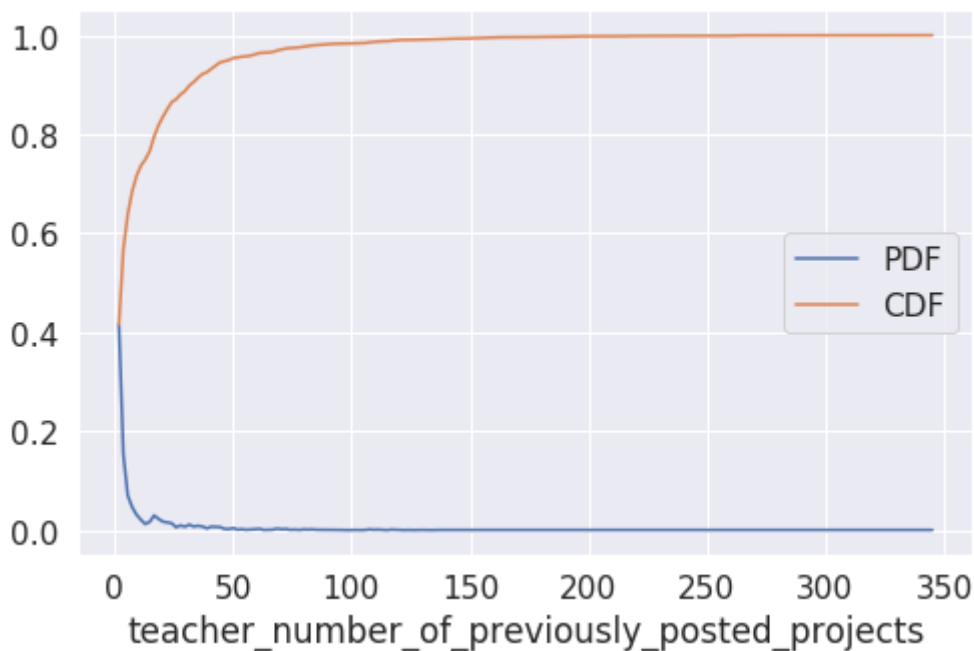
Out[171]:

```
<matplotlib.axes._subplots.AxesSubplot at 0x7f5e13ff49e8>
```



In [172]:

```
#PDF (FP ,teacher_number_of_previously_posted_projects)
plt.figure(figsize=(8,5))
counts, bin_edges = np.histogram(X_test_falsePos1['teacher_number_of_previously_posted_projects'], bins=10)
pdf = counts/sum(counts)
cdf = np.cumsum(pdf)
pdfP, = plt.plot(bin_edges[1:], pdf)
cdfP, = plt.plot(bin_edges[1:], cdf)
plt.legend([pdfP, cdfP], ["PDF", "CDF"])
plt.xlabel('teacher_number_of_previously_posted_projects')
plt.show()
```



In [ ]:

## Set 4 : Categorical, Numerical features + Project\_title(TFIDF W2V) + Preprocessed\_essay (TFIDF W2V)

In [173]:

```
# merge two sparse matrices: https://stackoverflow.com/a/19710648/4084039
```

```
from scipy.sparse import hstack
```

```
X_tr = hstack((categories_one_hot_train, sub_categories_one_hot_train, school_state_categories
```

```
X_te = hstack((categories_one_hot_test, sub_categories_one_hot_test, school_state_categories
```

```
X_cr = hstack((categories_one_hot_cv, sub_categories_one_hot_cv, school_state_categories_one
```

In [174]:

```
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)
```

```
Final Data matrix
(49041, 703) (49041,)
(24155, 703) (24155,)
(36052, 703) (36052,)
```

## GridSearchCV

In [175]:

```
dt4 = DecisionTreeClassifier(class_weight = 'balanced')
parameters = {'max_depth': [1, 5, 10, 50, 100, 500, 1000], 'min_samples_split': [5, 10, 20,
clf4 = GridSearchCV(dt4, parameters, cv=2, scoring='roc_auc', n_jobs=-1, return_train_score=
clf4.fit(X_tr, y_train)
```

Out[175]:

```
GridSearchCV(cv=2, error_score='raise-deprecating',
             estimator=DecisionTreeClassifier(class_weight='balanced',
                                              criterion='gini', max_depth=No
ne,
                                              max_features=None,
                                              max_leaf_nodes=None,
                                              min_impurity_decrease=0.0,
                                              min_impurity_split=None,
                                              min_samples_leaf=1,
                                              min_samples_split=2,
                                              min_weight_fraction_leaf=0.0,
                                              presort=False, random_state=No
ne,
                                              splitter='best'),
             iid='warn', n_jobs=-1,
             param_grid={'max_depth': [1, 5, 10, 50, 100, 500, 1000],
                         'min_samples_split': [5, 10, 20, 45, 75, 100, 135,
270,
                                              500]}},
             pre_dispatch='2*n_jobs', refit=True, return_train_score=True,
             scoring='roc_auc', verbose=0)
```

In [176]:

```
import seaborn as sns; sns.set()

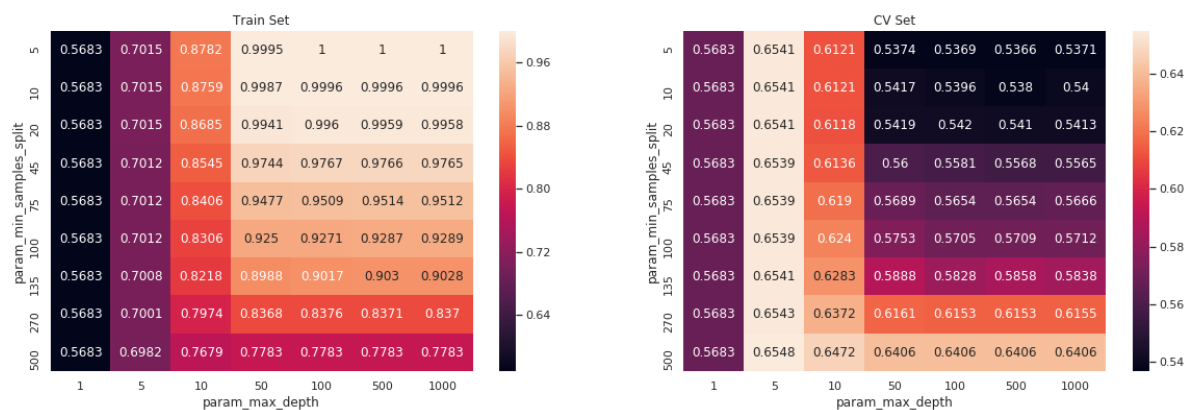
max_scores4 = pd.DataFrame(clf4.cv_results_).groupby(['param_min_samples_split', 'param_max_depth'])

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

sns.heatmap(max_scores4.mean_train_score, annot = True, fmt='.4g', ax=ax[0])
sns.heatmap(max_scores4.mean_test_score, annot = True, fmt='.4g', ax=ax[1])

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

plt.show()
```

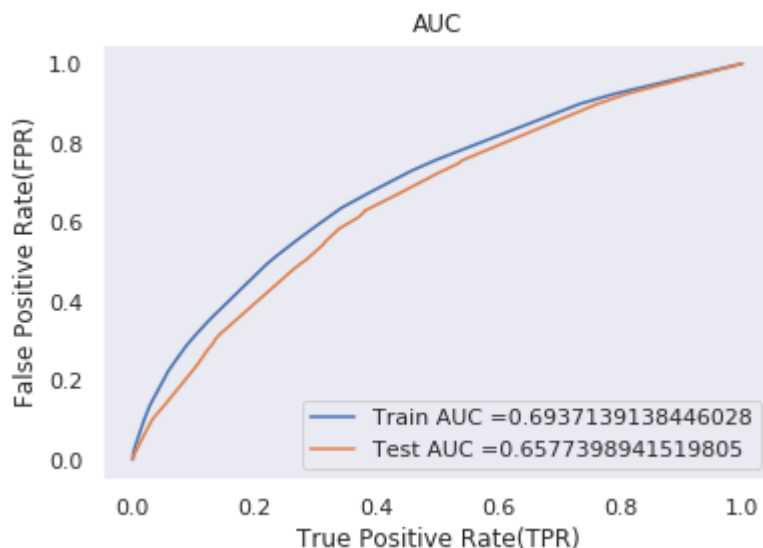


## Train the model using the best hyper parameter value

In [177]:

```
# https://scikitlearn.org/stable/modules/generated/sklearn.metrics.roc_curve.html#sklearn.m
from sklearn.metrics import roc_curve, auc
model = DecisionTreeClassifier(max_depth = 5, min_samples_split = 500, class_weight = 'balanced')
model.fit(X_tr, y_train)
clfV1=DecisionTreeClassifier (class_weight = 'balanced', max_depth=5, min_samples_split=500)
# for visulation
clfV1.fit(X_tr, y_train)
# roc_auc_score(y_true, y_score) the 2nd parameter should be probability estimates of the p
# not the predicted outputs
y_train_pred = batch_predict(model, X_tr)
y_test_pred = batch_predict(model, X_te)

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()
plt.show()
```



## Confusion Matrix

In [178]:

```
from sklearn.metrics import confusion_matrix
print("Train confusion matrix")
print(confusion_matrix(y_train, predict(y_train_pred, tr_thresholds, train_fpr, train_tpr)))
```

Train confusion matrix  
the maximum value of  $tpr \cdot (1 - fpr)$  0.24999412463136592 for threshold 0.403  
[[ 3731 3695]  
 [10129 31486]]

In [179]:

```
conf_matr_df_train_7 = pd.DataFrame(confusion_matrix(y_train, predict(y_train_pred, tr_thre
```

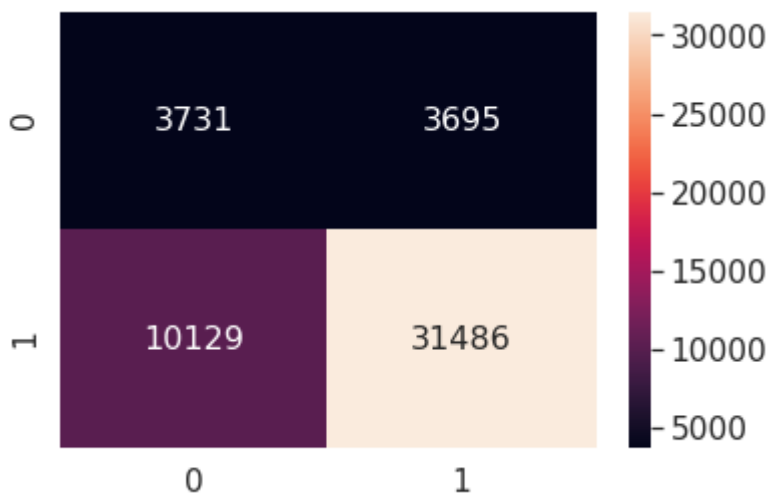
the maximum value of  $tpr*(1-fpr)$  0.24999412463136592 for threshold 0.403

In [180]:

```
sns.set(font_scale=1.4)#for label size
sns.heatmap(conf_matr_df_train_7, annot=True,annot_kws={"size": 16}, fmt='g')
```

Out[180]:

<matplotlib.axes.\_subplots.AxesSubplot at 0x7f5e13e1dcf8>



In [181]:

```
print("Test confusion matrix")
print(confusion_matrix(y_test, predict(y_test_pred, tr_thresholds, test_fpr, test_fpr)))
```

Test confusion matrix

the maximum value of  $tpr*(1-fpr)$  0.24999858224716406 for threshold 0.444

```
[[ 2736  2723]
 [ 8510 22083]]
```

In [182]:

```
conf_matr_df_test_8 = pd.DataFrame(confusion_matrix(y_test, predict(y_test_pred, tr_threshc
```

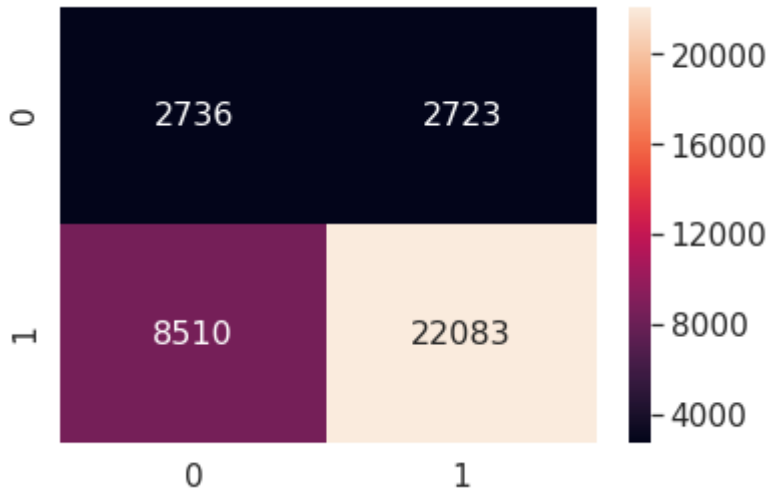
the maximum value of  $tpr*(1-fpr)$  0.24999858224716406 for threshold 0.444

In [183]:

```
sns.set(font_scale=1.4)#for label size
sns.heatmap(conf_matr_df_test_8, annot=True,annot_kws={"size": 16}, fmt='g')
```

Out[183]:

<matplotlib.axes.\_subplots.AxesSubplot at 0x7f5e14310048>



In [184]:

```
#Analysis on the False positives
fpi = []
for i in range(len(y_test)) :
    if (y_test.values[i] == 0) & (predictions1[i] == 1) :
        fpi.append(i)
fp_essay1 = []
for i in fpi :
    fp_essay1.append(X_test['clean_essays'].values[i])
```

In [185]:

```
len(fp_essay1)
```

Out[185]:

2723

In [186]:

```
#WORD CLOUD OF ESSAY
```

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

```
comment_words = ''
```

```
stopwords = set(STOPWORDS)
```

```
for val in fp_essay1 :
```

```
    val = str(val)
```

```
    tokens = val.split()
```

```
for i in range(len(tokens)):
```

```
    tokens[i] = tokens[i].lower()
```

```
for words in tokens :
```

```
    comment_words = comment_words + words + ' '
```

```
wordcloud = WordCloud(width = 800, height = 800, background_color = 'white', stopwords = stop
```

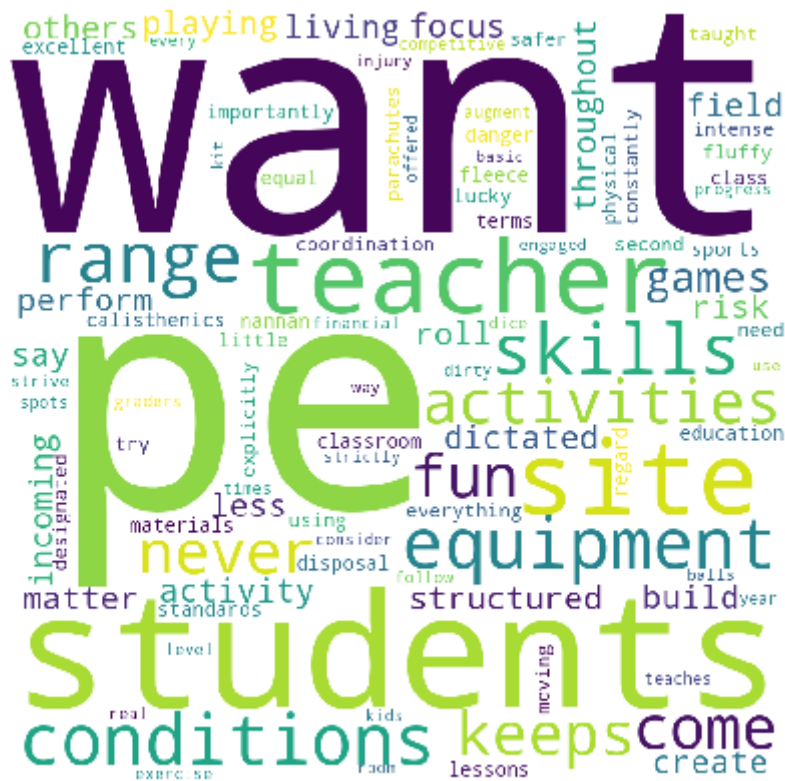
```
plt.figure(figsize = (6, 6), facecolor = None)
```

```
plt.imshow(wordcloud)
```

```
plt.axis("off")
```

```
plt.tight_layout(pad = 0)
```

```
plt.show()
```



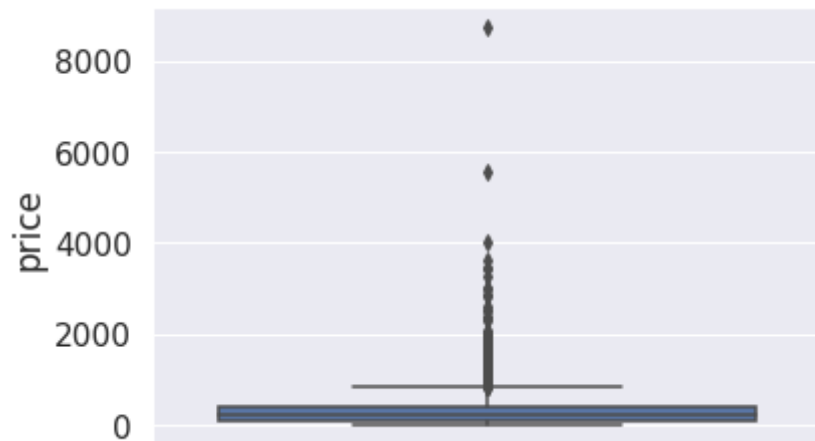


In [187]:

```
#Box Plot (FP 'price')
# first get the columns:
cols = X_test.columns
X_test_falsePos1 = pd.DataFrame(columns=cols)
# get the data of the false pisitives
for i in fpi : # (in fpi all the false positives data points indexes)
    X_test_falsePos1 = X_test_falsePos1.append(X_test.filter(items=[i], axis=0))
sns.boxplot(y='price', data=X_test_falsePos1)
```

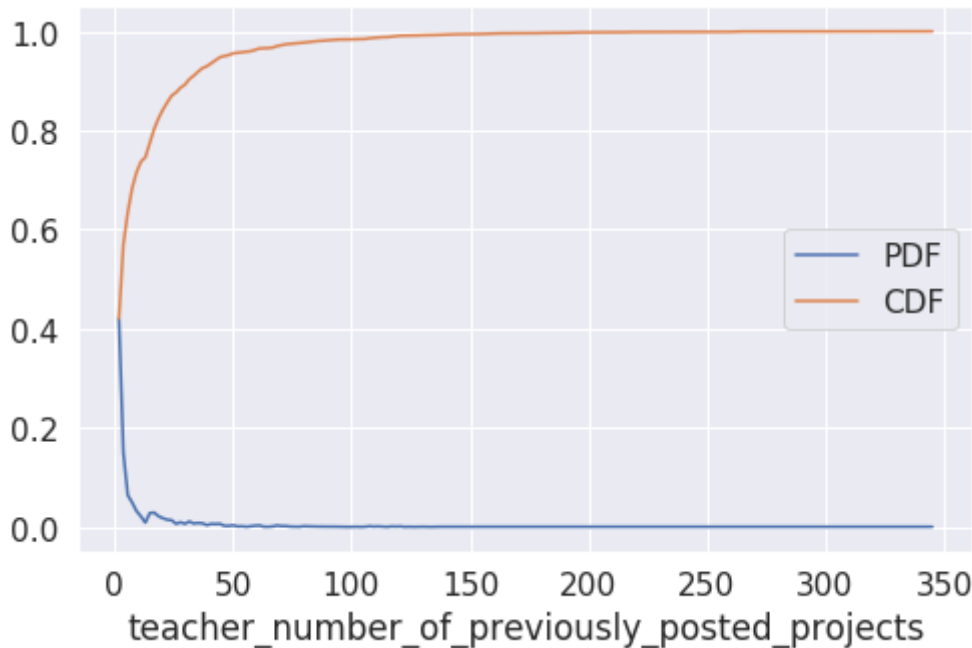
Out[187]:

<matplotlib.axes.\_subplots.AxesSubplot at 0x7f5e13c5cf98>



In [188]:

```
#PDF (FP ,teacher_number_of_previously_posted_projects)
plt.figure(figsize=(8,5))
counts, bin_edges = np.histogram(X_test_falsePos1['teacher_number_of_previously_posted_proj
pdf = counts/sum(counts)
cdf = np.cumsum(pdf)
pdfP, = plt.plot(bin_edges[1:], pdf)
cdfP, = plt.plot(bin_edges[1:], cdf)
plt.legend([pdfP, cdfP], ["PDF", "CDF"])
plt.xlabel('teacher_number_of_previously_posted_projects')
plt.show()
```



In [ ]:

## Set 5 Select 5k best features from features of Set 2 using feature\_importances

In [189]:

```
#https://stackoverflow.com/questions/47111434/randomforestregressor-and-feature-importances
from sklearn.ensemble import RandomForestClassifier
from sklearn.ensemble import RandomForestRegressor
from sklearn.model_selection import GridSearchCV
def selectKImportance(model, X, k=5):
    return X[:,model.best_estimator_.feature_importances_.argsort()[::-1][:k]]
```

In [190]:

```
X_train5 = selectKImportance(clf2, X_tr2, 5000)
X_test5 = selectKImportance(clf2, X_te2, 5000)
print(X_train5.shape)
print(X_test5.shape)
```

(49041, 5000)

(36052, 5000)

In [191]:

```
dt5 = DecisionTreeClassifier(class_weight = 'balanced')
parameters = {'max_depth': [1, 5, 10, 50, 100, 500, 1000], 'min_samples_split': [5, 10, 20, 45, 75, 100, 135, 170, 200, 250, 300, 350, 400, 450, 500]},

clf5 = GridSearchCV(dt5, parameters, cv= 2, scoring='roc_auc', n_jobs=-1, return_train_score=True)
clf5.fit(X_tr2, y_train)
```

Out[191]:

```
GridSearchCV(cv=2, error_score='raise-deprecating',
             estimator=DecisionTreeClassifier(class_weight='balanced',
                                              criterion='gini', max_depth=None,
                                              max_features=None,
                                              max_leaf_nodes=None,
                                              min_impurity_decrease=0.0,
                                              min_impurity_split=None,
                                              min_samples_leaf=1,
                                              min_samples_split=2,
                                              min_weight_fraction_leaf=0.0,
                                              presort=False, random_state=None,
                                              splitter='best'),
             iid='warn', n_jobs=-1,
             param_grid={'max_depth': [1, 5, 10, 50, 100, 500, 1000],
                        'min_samples_split': [5, 10, 20, 45, 75, 100, 135, 170, 200, 250, 300, 350, 400, 450, 500]}},
             pre_dispatch='2*n_jobs', refit=True, return_train_score=True,
             scoring='roc_auc', verbose=0)
```

In [192]:

```
import seaborn as sns; sns.set()

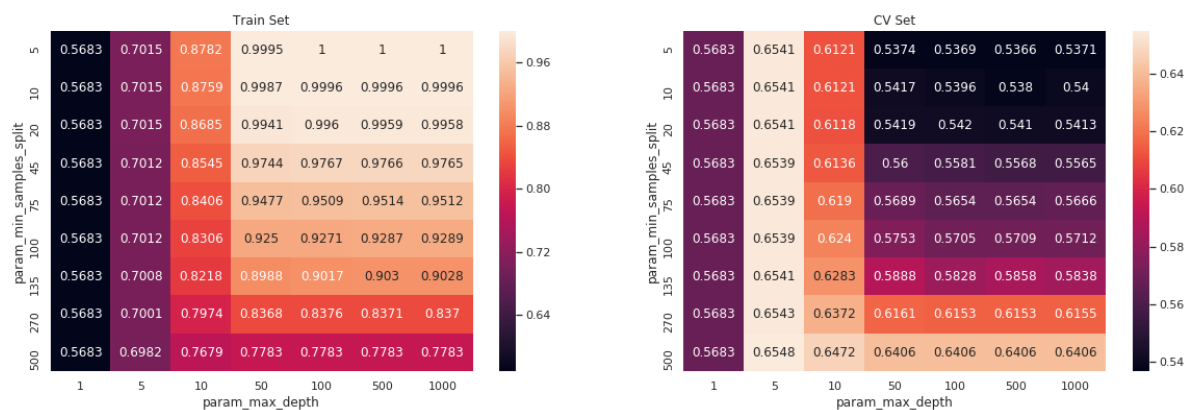
max_scores5 = pd.DataFrame(clf4.cv_results_).groupby(['param_min_samples_split', 'param_max_depth'])

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

sns.heatmap(max_scores5.mean_train_score, annot = True, fmt='.4g', ax=ax[0])
sns.heatmap(max_scores5.mean_test_score, annot = True, fmt='.4g', ax=ax[1])

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

plt.show()
```

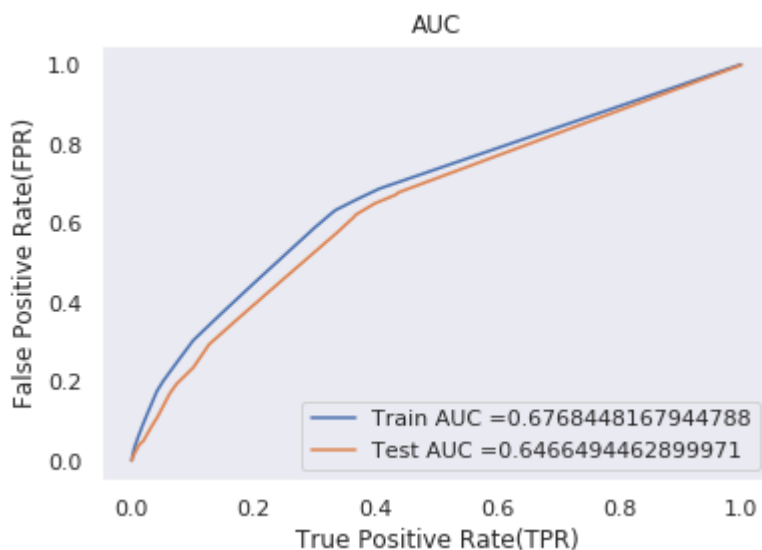


## Train Model using best Hyperparameter Value

In [193]:

```
# https://scikitlearn.org/stable/modules/generated/sklearn.metrics.roc_curve.html#sklearn.m
from sklearn.metrics import roc_curve, auc
model = DecisionTreeClassifier(max_depth = 5, min_samples_split = 500, class_weight = 'balanced')
model.fit(X_tr2, y_train)
clfV1=DecisionTreeClassifier (class_weight = 'balanced', max_depth=5, min_samples_split=500)
# for visulation
clfV1.fit(X_tr2, y_train)
# roc_auc_score(y_true, y_score) the 2nd parameter should be probability estimates of the p
# not the predicted outputs
y_train_pred = batch_predict(model, X_tr2)
y_test_pred = batch_predict(model, X_te2)

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()
plt.show()
```



## Confusion Matrix

In [194]:

```
from sklearn.metrics import confusion_matrix
print("Train confusion matrix")
print(confusion_matrix(y_train, predict(y_train_pred, tr_thresholds, train_fpr, train_tpr)))
```

Train confusion matrix  
the maximum value of  $tpr * (1 - fpr)$  0.24210089328089218 for threshold 0.365  
[[ 4373 3053]  
 [12930 28685]]

In [195]:

```
conf_matr_df_train_9 = pd.DataFrame(confusion_matrix(y_train, predict(y_train_pred, tr_thre
```

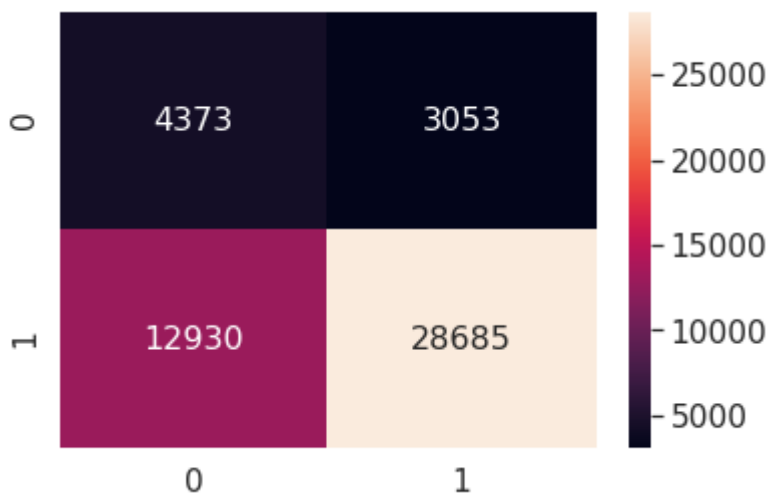
the maximum value of  $tpr \cdot (1-fpr)$  0.24210089328089218 for threshold 0.365

In [196]:

```
sns.set(font_scale=1.4)#for label size
sns.heatmap(conf_matr_df_train_9, annot=True,annot_kws={"size": 16}, fmt='g')
```

Out[196]:

<matplotlib.axes.\_subplots.AxesSubplot at 0x7f5e13c579e8>



In [197]:

```
print("Test confusion matrix")
print(confusion_matrix(y_test, predict(y_test_pred, tr_thresholds, test_fpr, test_fpr)))
```

Test confusion matrix

the maximum value of  $tpr \cdot (1-fpr)$  0.2460636386128223 for threshold 0.365

```
[[ 3072  2387]
 [ 9879 20714]]
```

In [198]:

```
conf_matr_df_test_10 = pd.DataFrame(confusion_matrix(y_test, predict(y_test_pred, tr_thresh
```

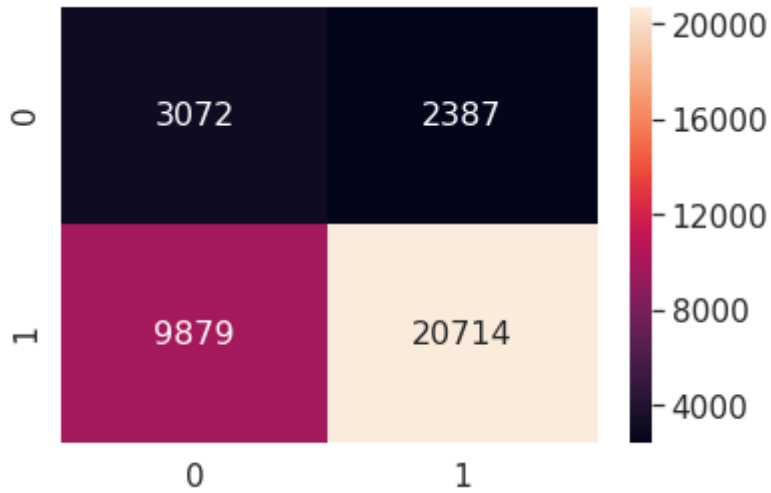
the maximum value of  $tpr \cdot (1-fpr)$  0.2460636386128223 for threshold 0.365

In [199]:

```
sns.set(font_scale=1.4)#for label size
sns.heatmap(conf_matr_df_test_10, annot=True,annot_kws={"size": 16}, fmt='g')
```

Out[199]:

<matplotlib.axes.\_subplots.AxesSubplot at 0x7f5e143ea0f0>



In [200]:

```
#Analysis on the False positives
fpi = []
for i in range(len(y_test)) :
    if (y_test.values[i] == 0) & (predictions1[i] == 1) :
        fpi.append(i)
fp_essay1 = []
for i in fpi :
    fp_essay1.append(X_test['clean_essays'].values[i])
```

In [201]:

```
#WORD CLOUD OF ESSAY
from wordcloud import WordCloud, STOPWORDS
comment_words = ' '
stopwords = set(STOPWORDS)
for val in fp_essay1 :
    val = str(val)
    tokens = val.split()
    for i in range(len(tokens)):
        tokens[i] = tokens[i].lower()
    for words in tokens :
        comment_words = comment_words + words + ' '

wordcloud = WordCloud(width = 800, height = 800, background_color ='white', stopwords =stop
plt.figure(figsize = (6, 6), facecolor = None)
plt.imshow(wordcloud)
plt.axis("off")
plt.tight_layout(pad = 0)
plt.show()
```

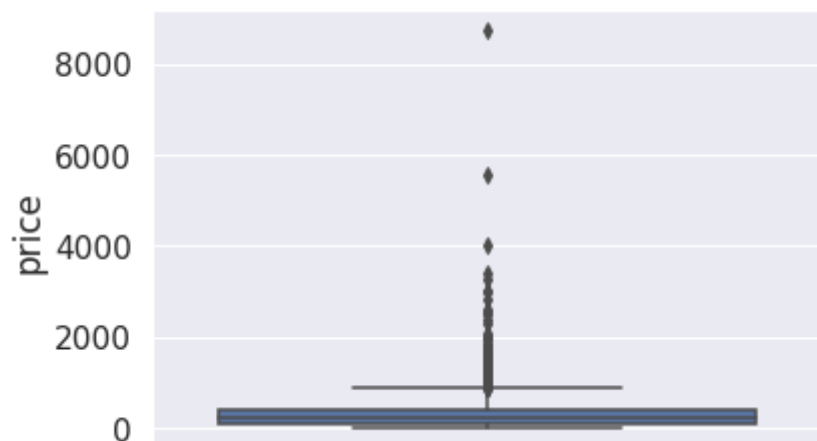


In [202]:

```
#Box Plot (FP 'price')
# first get the columns:
cols = X_test.columns
X_test_falsePos1 = pd.DataFrame(columns=cols)
# get the data of the false pisitives
for i in fpi : # (in fpi all the false positives data points indexes)
    X_test_falsePos1 = X_test_falsePos1.append(X_test.filter(items=[i], axis=0))
sns.boxplot(y='price', data=X_test_falsePos1)
```

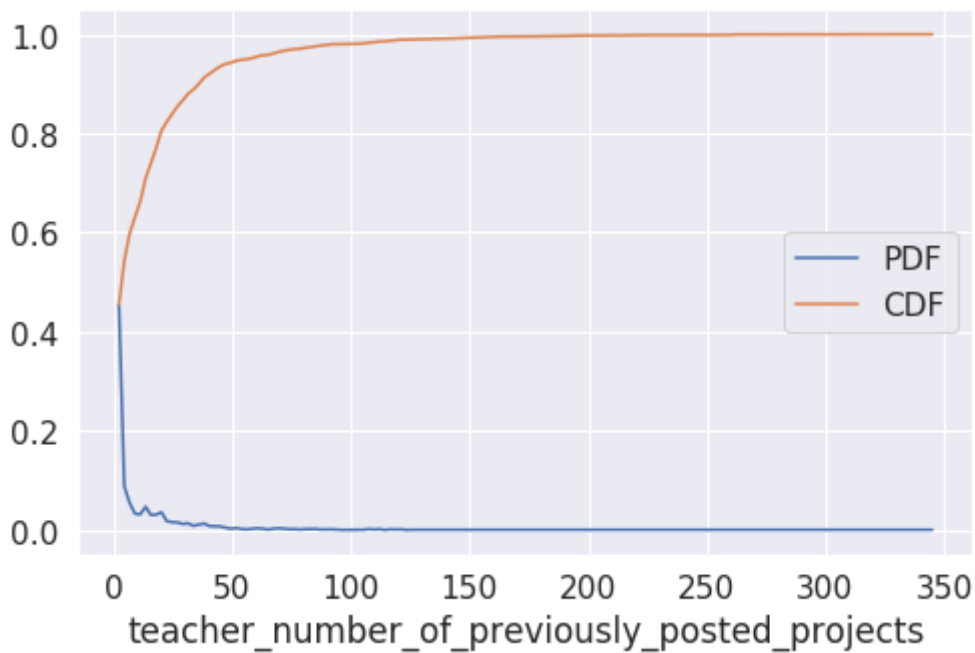
Out[202]:

<matplotlib.axes.\_subplots.AxesSubplot at 0x7f5e138b4630>



In [204]:

```
#PDF (FP ,teacher_number_of_previously_posted_projects)
plt.figure(figsize=(8,5))
counts, bin_edges = np.histogram(X_test_falsePos1['teacher_number_of_previously_posted_proj
pdf = counts/sum(counts)
cdf = np.cumsum(pdf)
pdfP, = plt.plot(bin_edges[1:], pdf)
cdfP, = plt.plot(bin_edges[1:], cdf)
plt.legend([pdfP, cdfP], ["PDF", "CDF"])
plt.xlabel('teacher_number_of_previously_posted_projects')
plt.show()
```



In [ ]:

In [205]:

```

from prettytable import PrettyTable

#If you get a ModuleNotFoundError error , install prettytable using: pip3 install prettytab

x = PrettyTable()
x.field_names = ["Vectorizer", "Max depth", "min samples Split", "AUC"]

x.add_row(["BOW", 100, 500, 0.674])
x.add_row(["TFIDF", 10, 500, 0.676])
x.add_row(["AVG W2V", 10, 270, 0.650])
x.add_row(["TFIDF W2V", 5, 500, 0.657])
x.add_row(["5k best features", 5, 500, 0.646])

print(x)

```

Vectorizer	Max depth	min samples Split	AUC
BOW	100	500	0.674
TFIDF	10	500	0.676
AVG W2V	10	270	0.65
TFIDF W2V	5	500	0.657
5k best features	5	500	0.646

In [ ]: