

```
In [1]: %matplotlib inline
import warnings
warnings.filterwarnings("ignore")

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

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

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

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

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

READING DATA

```
In [2]: dft = pd.read_csv('train_data.csv',nrows=60000)
dfr = pd.read_csv('resources.csv')
```

```
In [3]: print("Number of data points in train data", dft.shape)
print('-'*50)
print("The attributes of data :", dft.columns.values)
```

Number of data points in train data (60000, 17)

The attributes of data : ['Unnamed: 0' 'id' 'teacher_id' 'teacher_prefix' 'school_state'
 'project_submitted_datetime' 'project_grade_category'
 'project_subject_categories' 'project_subject_subcategories'
 'project_title' 'project_essay_1' 'project_essay_2' 'project_essay_3'
 'project_essay_4' 'project_resource_summary'
 'teacher_number_of_previously_posted_projects' 'project_is_approved']

```
In [4]: print("Number of data points in train data", dfr.shape)
print(dfr.columns.values)
dfr.head(2)
```

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

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

Out[4]:

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

```
In [5]: # 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(dft.columns)]

#sort dataframe based on time pandas python: https://stackoverflow.com/a/49702492/4084039
dft['Date'] = pd.to_datetime(dft['project_submitted_datetime'])
dft.drop('project_submitted_datetime', axis=1, inplace=True)
dft.sort_values(by=['Date'], inplace=True)

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

dft.head(2)
```

Out[5]:

	Unnamed: 0	id	teacher_id	teacher_prefix	school_state	
55660	8393	p205479	2bf07ba08945e5d8b2a3f269b2b3cfe5	Mrs.	CA	200400:27
51140	74477	p189804	4a97f3a390bfe21b99cf5e2b81981c73	Mrs.	CA	200400:46

TEXT PROCESSING

```
In [6]: # merge two column text dataframe:
dft["essay"] = dft["project_essay_1"].map(str) + \
               dft["project_essay_2"].map(str) + \
               dft["project_essay_3"].map(str) + \
               dft["project_essay_4"].map(str)
```

In [7]: dft.head(2)

Out[7]:

Unnamed: 0	id	teacher_id	teacher_prefix	school_state	
55660	8393 p205479	2bf07ba08945e5d8b2a3f269b2b3cfe5	Mrs.	CA	2004-00:27
51140	74477 p189804	4a97f3a390bfe21b99cf5e2b81981c73	Mrs.	CA	2004-00:46

In [8]: [# https://stackoverflow.com/a/47091490/4084039](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 [9]: # we are removing the words from the stop words list: 'no', 'nor', 'not'
stopwords= ['i', 'me', 'my', 'myself', 'we', 'our', 'ours', 'ourselves', 'you'
, "you're", "you've", \
            "you'll", "you'd", 'your', 'yours', 'yourself', 'yourselves', 'he'
, 'him', 'his', 'himself', \
            'she', "she's", 'her', 'hers', 'herself', 'it', "it's", 'its', 'it
self', 'they', 'them', 'their', \
            'theirs', 'themselves', 'what', 'which', 'who', 'whom', 'this', 't
hat', "that'll", 'these', 'those', \
            'am', 'is', 'are', 'was', 'were', 'be', 'been', 'being', 'have',
'has', 'had', 'having', 'do', 'does', \
            'did', 'doing', 'a', 'an', 'the', 'and', 'but', 'if', 'or', 'becau
se', 'as', 'until', 'while', 'of', \
            'at', 'by', 'for', 'with', 'about', 'against', 'between', 'into',
'through', 'during', 'before', 'after', \
            'above', 'below', 'to', 'from', 'up', 'down', 'in', 'out', 'on',
'off', 'over', 'under', 'again', 'further', \
            'then', 'once', 'here', 'there', 'when', 'where', 'why', 'how', 'a
ll', 'any', 'both', 'each', 'few', 'more', \
            'most', 'other', 'some', 'such', 'only', 'own', 'same', 'so', 'tha
n', 'too', 'very', \
            's', 't', 'can', 'will', 'just', 'don', "don't", 'should', "shoul
d've", 'now', 'd', 'll', 'm', 'o', 're', \
            've', 'y', 'ain', 'aren', "aren't", 'couldn', "couldn't", 'didn',
"didn't", 'doesn', "doesn't", 'hadn', \
            "hadn't", 'hasn', "hasn't", 'haven', "haven't", 'isn', "isn't", 'm
a', 'mightn', "mightn't", 'mustn', \
            "mustn't", 'needn', "needn't", 'shan', "shan't", 'shouldn', "shoul
dn't", 'wasn', "wasn't", 'weren', "weren't", \
            'won', "won't", 'wouldn', "wouldn't"]
```

Preprocessing of project_subject_categories

```

In [10]: categories = list(dft['project_subject_categories'].values)

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 & Scienc
e", "Warmth", "Care & Hunger"]
        if 'The' in j.split(): # this will split each of the category based on
space "Math & Science"=> "Math", "&", "Science"
            j=j.replace('The', '') # if we have the words "The" we are going to
replace it with '' (i.e removing 'The')
            j = j.replace(' ', '') # we are placing all the ' '(space) with '' (empty)
ex: "Math & Science"=> "Math&Science"
            temp+=j.strip()+" " # " abc ".strip() will return "abc", remove the tra
iling spaces
            temp = temp.replace('&', '_') # we are replacing the & value into
cat_list.append(temp.strip())

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

from collections import Counter
my_counter = Counter()
for word in dft['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]))

```

Preprocessing of project_subject_subcategories

```

In [11]: sub_categories = list(dft['project_subject_subcategories'].values)
# remove special characters from list of strings python:
#https://stackoverflow.com/a/47301924/4084039

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 & Scienc
e", "Warmth", "Care & Hunger"]
        if 'The' in j.split(): # this will split each of the category based on
space "Math & Science"=> "Math","&", "Science"
            j=j.replace('The','') # if we have the words "The" we are going to
replace it with ''(i.e removing 'The')
            j = j.replace(' ', '') # we are placing all the ' '(space) with ''(empt
y) ex:"Math & Science"=>"Math&Science"
            temp +=j.strip()+" #" abc ".strip() will return "abc", remove the tra
iling spaces
            temp = temp.replace('&','_')
            sub_cat_list.append(temp.strip())

dft['clean_subcategories'] = sub_cat_list
dft.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 dft['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]))

```

```
In [12]: # we have to remove the grades from every row
print(dft['project_grade_category'][:20])
```

```
55660    Grades PreK-2
51140    Grades PreK-2
473      Grades PreK-2
41558      Grades 3-5
29891      Grades 3-5
23374    Grades PreK-2
49228    Grades PreK-2
7176     Grades PreK-2
35006      Grades 3-5
5145      Grades 3-5
48237      Grades 9-12
52282      Grades 9-12
46375      Grades 3-5
36468    Grades PreK-2
36358    Grades PreK-2
39438    Grades PreK-2
2521     Grades PreK-2
58794    Grades PreK-2
40180    Grades PreK-2
53562      Grades 9-12
Name: project_grade_category, dtype: object
```

```
In [13]: d = list(dft['project_grade_category'].values)
# remove special characters from list of strings python:
# https://www.geeksforgeeks.org/removing-stop-words-nltk-python/

grade_cat_list = []
for i in d:
    # consider we have text like this:
    for j in i.split(' '): ## split by space
        j=j.replace('Grades','') # clean grades from the row
        grade_cat_list.append(j.strip())

dft['clean_grade'] = grade_cat_list
dft.drop(['project_grade_category'], axis=1, inplace=True)

my_counter = Counter()
for word in dft['clean_grade'].values:
    my_counter.update(word.split())
project_grade_category_dict= dict(my_counter)
sorted_project_grade_category_dict = dict(sorted(project_grade_category_dict.items(), key=lambda kv: kv[1]))
```

Preparing data for the models

Test - Train Split


```
In [14]: # train test split

from sklearn.model_selection import train_test_split

X_train, X_test, y_train, y_test = train_test_split(dft, dft['project_is_approved'], stratify = dft['project_is_approved'], test_size=0.33)
X_train, X_cv, y_train, y_cv = train_test_split(X_train, y_train, test_size=0.33, stratify=y_train)
```

```
In [15]: X_train, X_cv, y_train, y_cv = train_test_split(X_train, y_train, stratify= y_train, test_size = 0.33)
```

```
In [16]: print(y_train.value_counts())
print(y_test.value_counts())
print(y_cv.value_counts())

1    15295
0     2750
Name: project_is_approved, dtype: int64
1    16782
0     3018
Name: project_is_approved, dtype: int64
1     7535
0     1354
Name: project_is_approved, dtype: int64
```

```
In [17]: #dropping the y labels
#https://stackoverflow.com/questions/13411544/delete-column-from-pandas-dataframe-by-column-name

X_train.drop(["project_is_approved"], axis = 1, inplace = True)

X_test.drop(["project_is_approved"], axis = 1, inplace = True)

X_cv.drop(["project_is_approved"], axis = 1, inplace = True)
```

Text preprocessing

```
In [18]: #Preprocessing for essay
# 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['essay'].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%|███████████| 18045/18045 [00:20<00:00, 760.92it/s]
```

```
In [19]: #Processing for essay
# Combining all the above students
from tqdm import tqdm
preprocessed_essays_test = []
# tqdm is for printing the status bar
for sentence in tqdm(X_test['essay'].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%|██████████| 19800/19800 [00:26<00:00, 749.84it/s]
```

```
In [20]: #Preprocessing for essay
# Combining all the above students
from tqdm import tqdm
preprocessed_essays_cv = []
# tqdm is for printing the status bar
for sentence in tqdm(X_cv['essay'].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%|███████████| ██████████ | 8889/8889 [00:12<00:00, 736.00it/s]
```

```
In [21]: #Preprocessing for essay
# Combining all the above students
from tqdm import tqdm
preprocessed_titles_cv = []
# tqdm is for printing the status bar
for sentence in tqdm(X_cv['project_title'].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_titles_cv.append(sent.lower().strip())
```

```
100%|███████████████████████████████████████████████████  
██████████ | 8889/8889 [00:00<00:00, 16760.57it/s]
```

```
In [22]: #Processing for essay
# Combining all the above students
from tqdm import tqdm
preprocessed_titles_train = []
# tqdm is for printing the status bar
for sentence in tqdm(X_train['project_title'].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_titles_train.append(sent.lower().strip())
```

```
100%|███████████  
█ | 18045/18045 [00:01<00:00, 16870.17it/s]
```

```
In [23]: #Processing for essay
# Combining all the above students
from tqdm import tqdm
preprocessed_titles_test = []
# tqdm is for printing the status bar
for sentence in tqdm(X_test['project_title'].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_titles_test.append(sent.lower().strip())
```

```
100%|███████████████████████████████████████████  
█ | 19800/19800 [00:01<00:00, 16517.69it/s]
```

Encoding numerical, Categorical features

vectorize categorical data

```
In [24]: #project_subject_categories convert categorical to vectors

# convert train,cv and test data of clean_categories into vectors
# we use count vectorizer to convert the values into one
from sklearn.feature_extraction.text import CountVectorizer
vectorizer1 = CountVectorizer(vocabulary=list(sorted_cat_dict.keys()), lowerca
se=False,binary=True)
vectorizer1.fit(X_train['clean_categories'].values)
# firstly convert fit the train data into the vectoriaer then it learn hte voc
ablery
# we use the fitted CountVectorizer to convert the text to vector
X_train_cat = vectorizer1.transform(X_train['clean_categories'].values)
X_cv_cat = vectorizer1.transform(X_cv['clean_categories'].values)
X_test_cat = vectorizer1.transform(X_test['clean_categories'].values)
print(vectorizer1.get_feature_names())
```

```
['Warmth', 'Care_Hunger', 'History_Civics', 'Music_Arts', 'AppliedLearning',
'SpecialNeeds', 'Health_Sports', 'Math_Science', 'Literacy_Language']
```

```
In [25]: f1=vectorizer1.get_feature_names()
print("After vectorizations")
print(X_train_cat.shape, y_train.shape)
print(X_cv_cat.shape, y_cv.shape)
print(X_test_cat.shape, y_test.shape)
print("="*100)
```

```
After vectorizations
(18045, 9) (18045,)
(8889, 9) (8889,)
(19800, 9) (19800,)
```

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

```
In [26]: ##project_subject_subcategories convert categorical to vectors
# convert train,cv and test data of clean_categories into vectors
# we use count vectorizer to convert the values into one
from sklearn.feature_extraction.text import CountVectorizer
vectorizer2 = CountVectorizer(vocabulary=list(sorted_sub_cat_dict.keys()), low
ercase=False, binary=True)
vectorizer2.fit(X_train['clean_subcategories'].values)
# firstly convert fit the train data into the vectoriaer then it learn hte voc
ablery
# we use the fitted CountVectorizer to convert the text to vector
X_train_subcat = vectorizer2.transform(X_train['clean_subcategories'].values)
X_cv_subcat = vectorizer2.transform(X_cv['clean_subcategories'].values)
X_test_subcat = vectorizer2.transform(X_test['clean_subcategories'].values)
print(vectorizer2.get_feature_names())
```

```
['Economics', 'CommunityService', 'FinancialLiteracy', 'ParentInvolvement',
'Extracurricular', 'Civics_Government', 'ForeignLanguages', 'NutritionEducati
on', 'Warmth', 'Care_Hunger', 'SocialSciences', 'PerformingArts', 'CharacterE
ducation', 'TeamSports', 'Other', 'College_CareerPrep', 'History_Geography',
'Music', 'Health_LifeScience', 'EarlyDevelopment', 'ESL', 'Gym_Fitness', 'Env
ironmentalScience', 'VisualArts', 'Health_Wellness', 'AppliedSciences', 'Spec
ialNeeds', 'Literature_Writing', 'Mathematics', 'Literacy']
```

```
In [27]: print("After vectorizations")
print(X_train_subcat.shape, y_train.shape)
print(X_cv_subcat.shape, y_cv.shape)
print(X_test_subcat.shape, y_test.shape)
print("="*100)
```

```
After vectorizations
(18045, 30) (18045,)
(8889, 30) (8889,)
(19800, 30) (19800,)
```

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

```
In [28]: # school_state convert categorical to vectors
# now time to cont the each words
from collections import Counter
my_counter = Counter()
for word in dft['school_state'].values:
    my_counter.update(word.split())# count the words
school_state_dict = dict(my_counter)# store in dictionary
sorted_school_state_dict = dict(sorted(school_state_dict.items(), key=lambda k
v: kv[1]))
print(sorted_school_state_dict)
```

```
{'VT': 40, 'WY': 58, 'ND': 78, 'MT': 120, 'RI': 148, 'NH': 175, 'NE': 176, 'S
D': 177, 'DE': 181, 'AK': 188, 'WV': 252, 'HI': 270, 'ME': 277, 'DC': 294, 'N
M': 295, 'KS': 340, 'IA': 363, 'ID': 371, 'AR': 534, 'CO': 638, 'MN': 671, 'O
R': 676, 'MS': 710, 'KY': 725, 'NV': 774, 'MD': 801, 'CT': 923, 'TN': 935, 'A
L': 944, 'UT': 958, 'WI': 994, 'VA': 1124, 'AZ': 1172, 'NJ': 1235, 'OK': 128
3, 'LA': 1308, 'WA': 1309, 'MA': 1312, 'OH': 1399, 'MO': 1421, 'IN': 1431, 'P
A': 1699, 'MI': 1760, 'SC': 2186, 'GA': 2203, 'IL': 2371, 'NC': 2831, 'FL': 3
444, 'TX': 4010, 'NY': 4039, 'CA': 8377}
```

```
In [29]: # convert train,cv and test data of clean_categories into vectors
# we use count vectorizer to convert the values into one
from sklearn.feature_extraction.text import CountVectorizer
vectorizer3 = CountVectorizer(vocabulary=list(sorted_school_state_dict.keys
()), lowercase=False, binary=True)
vectorizer3.fit(dft['school_state'].values)
# firstly convert fit the train data into the vector then it learn the vocable
ry
# we use the fitted CountVectorizer to convert the text to vector
X_train_school_state = vectorizer3.transform(X_train['school_state'].values)
X_cv_school_state = vectorizer3.transform(X_cv['school_state'].values)
X_test_school_state = vectorizer3.transform(X_test['school_state'].values)
print(vectorizer3.get_feature_names())
```

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

```
In [30]: print("After vectorizations")
print(X_train_school_state .shape, y_train.shape)
print(X_cv_school_state .shape, y_cv.shape)
print(X_test_school_state .shape, y_test.shape)
print("="*100)
```

```
After vectorizations
(18045, 51) (18045,)
(8889, 51) (8889,)
(19800, 51) (19800,)
```

```
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```

```
In [31]: #project_grade_category categorical to vectors
#https://stackoverflow.com/questions/42224700/attributeerror-float-object-has-no-attribute-split
dft['clean_grade']=dft['clean_grade'].fillna("")# fill the null values with space
# convert train,cv and test data of clean_categories into vectors
# we use count vectorizer to convert the values into one
from sklearn.feature_extraction.text import CountVectorizer
vectorizer4 = CountVectorizer(vocabulary=list(sorted_project_grade_category_dict.keys()),lowercase=False, binary=True)
vectorizer4.fit(dft['clean_grade'].values)
# firstly convert fit the train data into the vectoriaer then it learn hte vocabulary
# we use the fitted CountVectorizer to convert the text to vector
X_train_project_grade_category = vectorizer4.transform(X_train['clean_grade'].values)
X_cv_project_grade_category = vectorizer4.transform(X_cv['clean_grade'].values)
X_test_project_grade_category = vectorizer4.transform(X_test['clean_grade'].values)
print(vectorizer4.get_feature_names())

['9-12', '6-8', '3-5', 'PreK-2']
```

```
In [32]: print("After vectorizations")
print(X_train_project_grade_category .shape, y_train.shape)
print(X_cv_project_grade_category .shape, y_cv.shape)
print(X_test_project_grade_category .shape, y_test.shape)
print("="*100)
```

```
After vectorizations
(18045, 4) (18045,)
(8889, 4) (8889,)
(19800, 4) (19800,)
=====
=====
```

```
In [33]: #teacher_prefix categorical to vectors
#https://stackoverflow.com/questions/42224700/attributeerror-float-object-has-no-attribute-split
dft['teacher_prefix']=dft['teacher_prefix'].fillna(" ")# filll the null values with space
my_counter = Counter()
for word in dft['teacher_prefix'].values:
    my_counter.update(word.split())
# dict sort by value python: https://stackoverflow.com/a/613218/4084039
teacher_cat_dict = dict(my_counter)
sorted_teacher_prefix_dict = dict(sorted(teacher_cat_dict.items(), key=lambda kv: kv[1]))
```

```
In [34]: # convert train,cv and test data of clean_categories into vectors
# we use count vectorizer to convert the values into one
from sklearn.feature_extraction.text import CountVectorizer
vectorizer5 = CountVectorizer(vocabulary=list(sorted_teacher_prefix_dict.keys()), lowercase=False, binary=True)
vectorizer5.fit(dft['teacher_prefix'].values.astype('U'))
# firstly convert fit the train data into the vectorizer
# we use the fitted CountVectorizer to convert the text to vector
X_train_teacher_prefix = vectorizer5.transform(X_train['teacher_prefix'].values.astype('U'))
X_cv_teacher_prefix = vectorizer5.transform(X_cv['teacher_prefix'].values.astype('U'))
X_test_teacher_prefix = vectorizer5.transform(X_test['teacher_prefix'].values.astype('U'))
print(vectorizer5.get_feature_names())
```

```
['Dr.', 'Teacher', 'Mr.', 'Ms.', 'Mrs.']
```

```
In [35]: print("After vectorizations")
print(X_train_teacher_prefix .shape, y_train.shape)
print(X_cv_teacher_prefix .shape, y_cv.shape)
print(X_test_teacher_prefix .shape, y_test.shape)
print("="*100)
```

```
After vectorizations
```

```
(18045, 5) (18045,)
```

```
(8889, 5) (8889,)
```

```
(19800, 5) (19800,)
```

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

Encoding essay, and Project_title

```
In [36]: #bow featurization essay

X_train_essay=preprocessed_essays_train
X_cv_essay=preprocessed_essays_cv
X_test_essay=preprocessed_essays_test

X_train_title=preprocessed_titles_train
X_cv_title=preprocessed_titles_cv
X_test_title=preprocessed_titles_test
# We are considering only the words which appeared in at least 10 documents(rows or projects).
vectorizer6 = CountVectorizer(min_df=10,max_features=5000,ngram_range=(1, 2))
vectorizer6.fit(X_train_essay)# that is Learned from trained data

# we use the fitted CountVectorizer to convert the text to vector
X_train_bow = vectorizer6.transform(X_train_essay)
X_cv_bow = vectorizer6.transform(X_cv_essay)
X_test_bow = vectorizer6.transform(X_test_essay)
```



```
In [37]: #bow featurization title
vectorizer7 = CountVectorizer(min_df=10,max_features=5000,ngram_range=(1, 2))
vectorizer7.fit(X_train_title)# that is learned from trained data
# we use the fitted CountVectorizer to convert the text to vector
X_train_bow_title = vectorizer7.transform(X_train_title)
X_cv_bow_title= vectorizer7.transform(X_cv_title)
X_test_bow_title = vectorizer7.transform(X_test_title)
print("After vectorizations")
print(X_train_bow_title.shape, y_train.shape)
print(X_cv_bow_title.shape, y_cv.shape)
print(X_test_bow_title.shape, y_test.shape)
print("="*100)
```

After vectorizations

(18045, 1327) (18045,)

(8889, 1327) (8889,)

(19800, 1327) (19800,)

=====

Tfidf featurization

```
In [38]: #for titles
from sklearn.feature_extraction.text import TfidfVectorizer
# We are considering only the words which appeared in at least 10 documents(ro
ws or projects).
vectorizer8 = TfidfVectorizer(min_df=10,max_features=5000,ngram_range=(1, 2))
vectorizer8.fit(X_train_title)# that is learned from trained data

# we use the fitted CountVectorizer to convert the text to vector
X_train_tf_title = vectorizer8.transform(X_train_title)
X_cv_tf_title= vectorizer8.transform(X_cv_title)
X_test_tf_title = vectorizer8.transform(X_test_title)
print("After vectorizations")
print(X_train_tf_title.shape, y_train.shape)
print(X_cv_tf_title.shape, y_cv.shape)
print(X_test_tf_title.shape, y_test.shape)
print("="*100)
```

After vectorizations

(18045, 1327) (18045,)

(8889, 1327) (8889,)

(19800, 1327) (19800,)

=====

```
In [39]: #for essay
from sklearn.feature_extraction.text import TfidfVectorizer
# We are considering only the words which appeared in at least 10 documents(rows or projects).
vectorizer9 = TfidfVectorizer(min_df=10,max_features=5000,ngram_range=(1, 2))
vectorizer9.fit(X_train_essay)# that is learned from trained data
# we use the fitted CountVectorizer to convert the text to vector
X_train_tf_essay = vectorizer9.transform(X_train_essay)
X_cv_tf_essay= vectorizer9.transform(X_cv_essay)
X_test_tf_essay = vectorizer9.transform(X_test_essay)
print("After vectorizations")
print(X_train_tf_essay.shape, y_train.shape)
print(X_cv_tf_essay.shape, y_cv.shape)
print(X_test_tf_essay.shape, y_test.shape)
print("="*100)
```

```
After vectorizations
(18045, 5000) (18045,)
(8889, 5000) (8889,)
(19800, 5000) (19800,)
```

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

Using Pretrained Models : AVG W2V

```
In [40]: # Reading glove vectors in python: https://stackoverflow.com/a/38230349/4084039

def loadGloveModel(gloveFile):

    print ("Loading Glove Model")

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

    model = {}

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

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

    return model
```

```
In [41]: model = loadGloveModel('glove.42B.300d.txt')
```

```
Loading Glove Model
```

```
1917495it [09:14, 3459.20it/s]
```

```
Done. 1917495 words loaded!
```

```
In [42]: glove_words = set(model.keys())
```

```
In [43]: #for essay  
# average Word2Vec  
# compute average word2vec for each review.  
def func(wordlist):  
  
    train_avg_w2v_vectors = []; # the avg-w2v for each sentence/review is stored  
in this list  
    for sentence in tqdm(wordlist): # for each review/sentence  
        vector = np.zeros(300) # as word vectors are of zero length # we are taking  
the 300dimensions very large  
        cnt_words = 0; # num of words with a valid vector in the sentence/review  
        for word in sentence.split(): # for each word in a review/sentence  
            if word in glove_words:  
                vector += model[word]  
                cnt_words += 1  
        if cnt_words != 0:  
            vector /= cnt_words  
        train_avg_w2v_vectors.append(vector)  
  
    print(len(train_avg_w2v_vectors))  
    print(len(train_avg_w2v_vectors[0]))  
    return train_avg_w2v_vectors
```

```
In [44]: train_avg_w2v_vectors=func(preprocessed_essays_train)
test_avg_w2v_vectors=func(preprocessed_essays_test)
cv_avg_w2v_vectors=func(preprocessed_essays_cv)
#for titles
cv_avg_w2v_vectors_title=func(preprocessed_titles_cv)
test_avg_w2v_vectors_title=func(preprocessed_titles_test)
train_avg_w2v_vectors_title=func(preprocessed_titles_train)
```

```
100%|████████████████████████████████████████████████████████████████████████████████|
██| 18045/18045 [00:09<00:00, 1805.19it/s]

18045
300

100%|████████████████████████████████████████████████████████████████████████████████|
██| 19800/19800 [00:10<00:00, 1884.12it/s]

19800
300

100%|████████████████████████████████████████████████████████████████████████████████|
███| 8889/8889 [00:04<00:00, 1841.96it/s]

8889
300

100%|████████████████████████████████████████████████████████████████████████████████|
███| 8889/8889 [00:00<00:00, 41983.19it/s]

8889
300

100%|████████████████████████████████████████████████████████████████████████████████|
██| 19800/19800 [00:00<00:00, 46198.56it/s]

19800
300

100%|████████████████████████████████████████████████████████████████████████████████|
██| 18045/18045 [00:00<00:00, 41310.19it/s]

18045
300
```

Using Pretrained Models: TFIDF weighted W2V

```
In [45]: tfidf_model = TfidfVectorizer()
tfidf_model.fit(preprocessed_essays_train)
# 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 [46]: # average Word2Vec
# compute average word2vec for each review.
def tf_idf_done(word_list):
    train_title_tfidf_w2v_vectors = []; # the avg-w2v for each sentence/review
    is stored in this list
    for sentence in tqdm(word_list): # 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/r
        eview
        for word in sentence.split():#.split(): # for each word in a review/se
        ntence
            if (word in glove_words) and (word in tfidf_words):
                #vec = model.wv[word]
                vec = model[word] # getting the vector for each word
            # here we are multiplying idf value(dictionary[word]) and the tf value((senten
            ce.count(word)/len(sentence.split())))
            tf_idf = dictionary[word]*(sentence.count(word)/len(sentence.sp
            lit()))
            vector += (vec * tf_idf) # calculating tfidf weighted w2v
            tf_idf_weight += tf_idf
            if tf_idf_weight != 0:
                vector /= tf_idf_weight
            train_title_tfidf_w2v_vectors.append(vector)
    print(len(train_title_tfidf_w2v_vectors))
    print(len(train_title_tfidf_w2v_vectors[0]))
    return train_title_tfidf_w2v_vectors

```

```

In [47]: train_tfidf_w2v_vectors=tf_idf_done(preprocessed_essays_train)
test_tfidf_w2v_vectors=tf_idf_done(preprocessed_essays_test)
cv_tfidf_w2v_vectors=tf_idf_done(preprocessed_essays_cv)

```

```

100%|████████████████████████████████████████████████████████████████████████████████|
████████████████████████████████████████████████████████████████████████████████| 18045/18045 [01:19<00:00, 225.72it/s]

18045
300

100%|████████████████████████████████████████████████████████████████████████████████|
████████████████████████████████████████████████████████████████████████████████| 19800/19800 [01:27<00:00, 225.26it/s]

19800
300

100%|████████████████████████████████████████████████████████████████████████████████|
████████████████████████████████████████████████████████████████████████████████| 8889/8889 [00:39<00:00, 222.26it/s]

8889
300

```

```
In [48]: train_title_tfidf_w2v_vectors=tf_idf_done(preprocessed_titles_train)
test_title_tfidf_w2v_vectors=tf_idf_done(preprocessed_titles_test)
cv_title_tfidf_w2v_vectors=tf_idf_done(preprocessed_titles_cv)
```

```
100%|████████████████████████████████████████████████████████████████████████████████|
18045/18045 [00:00<00:00, 19526.07it/s]
```

```
18045
300
```

```
100%|████████████████████████████████████████████████████████████████████████████████|
19800/19800 [00:01<00:00, 17606.28it/s]
```

```
19800
300
```

```
100%|████████████████████████████████████████████████████████████████████████████████|
8889/8889 [00:00<00:00, 17790.29it/s]
```

```
8889
300
```

Vectorizing Numerical features

```
In [49]: price_data = dfr.groupby('id').agg({'price':'sum', 'quantity':'sum'}).reset_in
dex()
dft = pd.merge(dft, price_data, on='id', how='left')
print(price_data.head(2))
# we also have to do this in tran,test and cv
# so also merge the resource data with the train,cv and test
X_train = pd.merge(X_train, price_data, on = "id", how = "left")
#print(x_train.columns)
X_test = pd.merge(X_test, price_data, on = "id", how = "left")
X_cv = pd.merge(X_cv, price_data, on = "id", how = "left")
```

	id	price	quantity
0	p000001	459.56	7
1	p000002	515.89	21

```
In [50]: #standardization
# check this one: https://www.youtube.com/watch?v=0HOqOcln3Z4&t=530s
# standardization sklearn: https://scikitlearn.org/stable/modules/generated/sk
Learn.preprocessing.StandardScaler.html
from sklearn.preprocessing import StandardScaler
from sklearn.preprocessing import MinMaxScaler
from sklearn import preprocessing

price_scalar = StandardScaler()
price_scalar.fit(X_train['price'].values.reshape(-1,1)) # finding the mean and
standard deviation of this data
#print(f"Mean : {price_scalar.mean_[0]}, Standard deviation : {np.sqrt(price_s
calar.var_[0])}")
# Now standardize the data with above mean and variance.
train_price_standar = price_scalar.transform(X_train['price'].values.reshape(-
1, 1))
# Now standardize the data with above maen and variance.
test_price_standar = price_scalar.transform(X_test['price'].values.reshape(-1,
1))
# Now standardize the data with above maen and variance.
cv_price_standar = price_scalar.transform(X_cv['price'].values.reshape(-1, 1))
```

```
In [51]: # previous_year_projects
price_scalar.fit(X_train['teacher_number_of_previously_posted_projects'].value
s.reshape(-1,1)) # finding the mean and standard deviation of this data
#print(f"Mean : {price_scalar.mean_[0]}, Standard deviation : {np.sqrt(price_s
calar.var_[0])}")
# Now standardize the data with above maen and variance.
train_prev_proj_standar =price_scalar.transform(X_train['teacher_number_of_pre
viously_posted_projects'].values.reshape(-1,1))
# Now standardize the data with above maen and variance.
test_prev_proj_standar =price_scalar.transform(X_test['teacher_number_of_previ
ously_posted_projects'].values.reshape(-1, 1))
# Now standardize the data with above maen and variance.
cv_prev_proj_standar = price_scalar.transform(X_cv['teacher_number_of_previous
ly_posted_projects'].values.reshape(-1, 1))
```

```
In [52]: price_scalar.fit(X_train['quantity'].values.reshape(-1,1)) # finding the mean
and standardddeviation of this data
#print(f"Mean : {price_scalar.mean_[0]}, Standard deviation : {np.sqrt(price_s
calar.var_[0])}")
# Now standardize the data with above maen and variance.
train_qnty_standar = price_scalar.transform(X_train['quantity'].values.reshape
(-1, 1))
# Now standardize the data with above maen and variance.
cv_qnty_standar = price_scalar.transform(X_cv['quantity'].values.reshape(-1, 1
))
# Now standardize the data with above maen and variance.
test_qnty_standar = price_scalar.transform(X_test['quantity'].values.reshape(-
1, 1))
```

merging

```
In [53]: from scipy.sparse import hstack
# with the same hstack function we are concatenating a sparse matrix and a dense matrix
X_set1_train = hstack((X_train_bow_title,X_train_bow,# all bows
                        X_train_teacher_prefix,X_train_cat,X_train_subcat,
                        X_train_project_grade_category,X_train_school_state,
                        train_qnty_standar,train_price_standar,train_prev_proj_
standar))
print(X_set1_train.shape, y_train.shape)

(18045, 6429) (18045,)
```

```
In [54]: from scipy.sparse import hstack
# with the same hstack function we are concatenating a sparse matrix and a dense matrix
X_set1_cv = hstack((X_cv_bow_title,X_cv_bow,
                    X_cv_teacher_prefix,X_cv_cat,X_cv_subcat,
                    X_cv_project_grade_category,X_cv_school_state,
                    cv_qnty_standar,cv_price_standar,cv_prev_proj_standar))
print(X_set1_cv.shape, y_cv.shape)

(8889, 6429) (8889,)
```

```
In [55]: from scipy.sparse import hstack
# with the same hstack function we are concatenating a sparse matrix and a dense matrix
X_set1_test = hstack((X_test_bow_title,X_test_bow,
                      X_test_teacher_prefix,X_test_cat,X_test_subcat,
                      X_test_project_grade_category,X_test_school_state,
                      test_qnty_standar,test_price_standar,test_prev_proj_stan
dar))
print(X_set1_test.shape, y_test.shape)

(19800, 6429) (19800,)
```

```
In [136]: xtr = X_set2_train.tocsr() # Here I have just applied kind of trail and Logic.
It was in coomatrix kada. Coomatrix is not accessible.
```

```
In [137]: xtr
```

```
Out[137]: <18045x6429 sparse matrix of type '<class 'numpy.float64'>'
          with 2359759 stored elements in Compressed Sparse Row format>
```



```
In [56]: from scipy.sparse import hstack
# with the same hstack function we are concatenating a sparse matrix and a dense matrix :)
X_set2_train = hstack((X_train_tf_essay,X_train_tf_title,
                        X_train_teacher_prefix,X_train_cat,X_train_subcat,
                        X_train_project_grade_category,X_train_school_state,
                        train_qnty_standar,train_price_standar,train_prev_proj_
standar))
print(X_set2_train.shape, y_train.shape)
```

(18045, 6429) (18045,)

```
In [57]: from scipy.sparse import hstack
# with the same hstack function we are concatenating a sparse matrix and a dense matrix :)
X_set2_cv = hstack((X_cv_tf_essay,X_cv_tf_title,
                    X_cv_teacher_prefix,X_cv_cat,X_cv_subcat,
                    X_cv_project_grade_category,X_cv_school_state,
                    cv_qnty_standar,cv_price_standar,cv_prev_proj_standar))
print(X_set2_cv.shape, y_cv.shape)
```

(8889, 6429) (8889,)

```
In [144]: # You are getting error coomatrix which is not accessible. For this reason, you
# are getting subscriptable issue.

# Overall, you want things in sparse form. I just checked the type of data and
it is coomatrix type which we don't want actually.

# So we have converted coomatrix type to sparse type using csr

# Here is the coomatrix type.

#type(X_set2_test)  # This is in coomatrix which we don't want and is not accessible.

xte = X_set2_test.tocsr()  # We want in sparse type and so we are converting it to sparse matrix rather than coomatrix type
type(xte)

#Instead of renaming everything just add an extension of .tocsr() wherever there is coomatrix type. Check below how I am doing
```

Out[144]: scipy.sparse.csr.csr_matrix

```
In [145]: from scipy.sparse import hstack
# with the same hstack function we are concatenating a sparse matrix and a dense matrix :)
X_set2_test = hstack((X_test_tf_essay,X_test_tf_title,
                      X_test_teacher_prefix,X_test_cat,X_test_subcat,
                      X_test_project_grade_category,X_test_school_state,
                      test_qnty_standar,test_price_standar,test_prev_proj_stan
dar)).tocsr()
print(X_set2_test.shape, y_test.shape)

(19800, 6429) (19800,)
```

```
In [59]: import numpy
s=numpy.array(train_avg_w2v_vectors)
print(X_train_project_grade_category.shape)
print(s.shape)

(18045, 4)
(18045, 300)
```

```
In [60]: from scipy.sparse import hstack
import numpy
# with the same hstack function we are concatenating a sparse matrix and a dense matrix :)
X_set3_train = hstack((numpy.array(train_avg_w2v_vectors),numpy.array(train_avg_w2v_vectors_title),train_prev_proj_standar,train_price_standar,train_qnty_standar,
                      X_train_teacher_prefix,X_train_cat,X_train_subcat,
                      X_train_project_grade_category,X_train_school_state))
print(X_set3_train.shape, y_train.shape)

(18045, 702) (18045,)
```

```
In [61]: from scipy.sparse import hstack
# with the same hstack function we are concatenating a sparse matrix and a dense matrix :)
X_set3_cv =hstack((cv_avg_w2v_vectors,cv_avg_w2v_vectors_title,cv_prev_proj_standar,cv_price_standar,cv_qnty_standar,
                  X_cv_teacher_prefix,X_cv_cat,X_cv_subcat,
                  X_cv_project_grade_category,X_cv_school_state))
print(X_set3_cv.shape, y_cv.shape)

(8889, 702) (8889,)
```

```
In [62]: from scipy.sparse import hstack
# with the same hstack function we are concatenating a sparse matrix and a dense matrix :)
X_set3_test =hstack((test_avg_w2v_vectors,test_avg_w2v_vectors_title,test_prev_proj_standar,test_price_standar,
                    test_qnty_standar,
                    X_test_teacher_prefix,X_test_cat,X_test_subcat,
                    X_test_project_grade_category,X_test_school_state))
print(X_set3_test.shape, y_test.shape)

(19800, 702) (19800,)
```

```
In [63]: import numpy
s=numpy.array(train_tfidf_w2v_vectors)
print(X_train_project_grade_category.shape)
print(s.shape)
```

```
(18045, 4)
(18045, 300)
```

```
In [64]: from scipy.sparse import hstack
# with the same hstack function we are concatenating a sparse matrix and a dense matrix :)
X_set4_train =hstack((train_tfidf_w2v_vectors, train_title_tfidf_w2v_vectors,train_prev_proj_standar,
                      train_price_standar,train_qnty_standar,
                      X_train_teacher_prefix,X_train_cat,X_train_subcat,
                      X_train_project_grade_category,X_train_school_state))

print(X_set4_train.shape, y_train.shape)
```

```
(18045, 702) (18045,)
```

```
In [65]: from scipy.sparse import hstack
# with the same hstack function we are concatenating a sparse matrix and a dense matrix :)
X_set4_cv =hstack((cv_tfidf_w2v_vectors,cv_title_tfidf_w2v_vectors,cv_prev_proj_standar,
                  cv_price_standar,cv_qnty_standar,
                  X_cv_teacher_prefix,X_cv_cat,X_cv_subcat,
                  X_cv_project_grade_category,X_cv_school_state))

print(X_set4_cv.shape, y_cv.shape)
```

```
(8889, 702) (8889,)
```

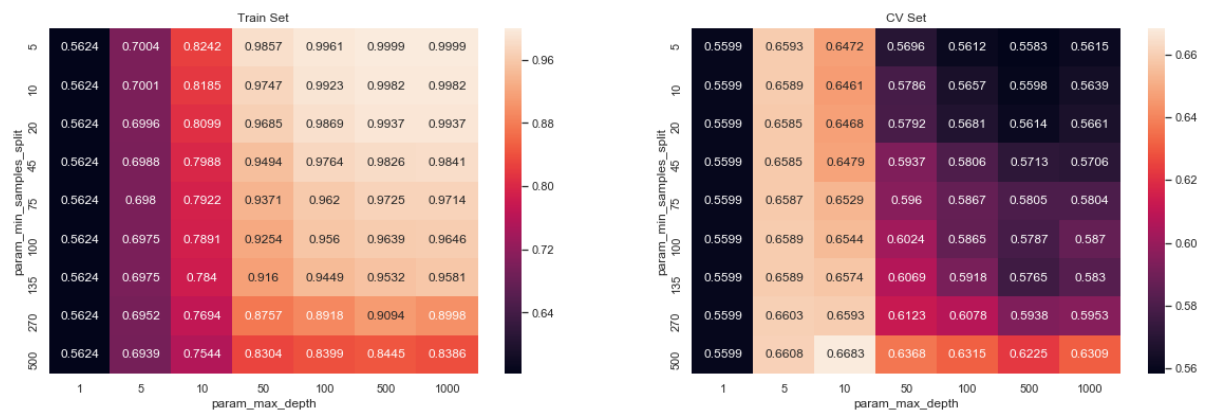
```
In [66]: from scipy.sparse import hstack
# with the same hstack function we are concatenating a sparse matrix and a dense matrix :)
X_set4_test = hstack((test_title_tfidf_w2v_vectors,test_tfidf_w2v_vectors,test_prev_proj_standar,test_price_standar,test_qnty_standar,X_test_teacher_prefix,
                    X_test_cat,X_test_subcat,
                    X_test_project_grade_category,X_test_school_state))
print(X_set4_test.shape, y_test.shape)
```

```
(19800, 702) (19800,)
```

Decison trees on BOW

```
In [67]: from sklearn.metrics import roc_auc_score
import matplotlib.pyplot as plt
from sklearn.model_selection import train_test_split
from sklearn.model_selection import GridSearchCV
from sklearn.model_selection import cross_val_score
from sklearn.tree import DecisionTreeClassifier
dt1 = DecisionTreeClassifier(class_weight = 'balanced')
parameters = {'max_depth': [1, 5, 10, 50, 100, 500, 1000], 'min_samples_split': [5, 10, 20, 45, 75, 100, 135, 270, 500]}
clf1 = GridSearchCV(dt1, parameters, cv=3, scoring='roc_auc', return_train_score=True)
se1 = clf1.fit(X_set1_train, y_train)
```

```
In [68]: import seaborn as sns; sns.set()
max_scores1 = pd.DataFrame(clf1.cv_results_).groupby(['param_min_samples_split', 'param_max_depth']).max().unstack()['mean_test_score', 'mean_train_score']
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()
```



Best Estimator and Best tune parameters

```
In [69]: print(clf1.best_estimator_)
#Mean cross-validated score of the best_estimator
print(clf1.score(X_set1_train,y_train))
print(clf1.score(X_set1_test,y_test))
```

```
DecisionTreeClassifier(class_weight='balanced', criterion='gini', max_depth=1
0,
                        max_features=None, max_leaf_nodes=None,
                        min_impurity_decrease=0.0, min_impurity_split=None,
                        min_samples_leaf=1, min_samples_split=500,
                        min_weight_fraction_leaf=0.0, presort=False,
                        random_state=None, splitter='best')
0.7511722190852625
0.6719906398813649
```

```
In [159]: # Best tune parameters
best_tune_parameters=[{ 'max_depth':[10], 'min_samples_split':[500] } ]
```

```
In [160]: clf1.get_params().keys()
```

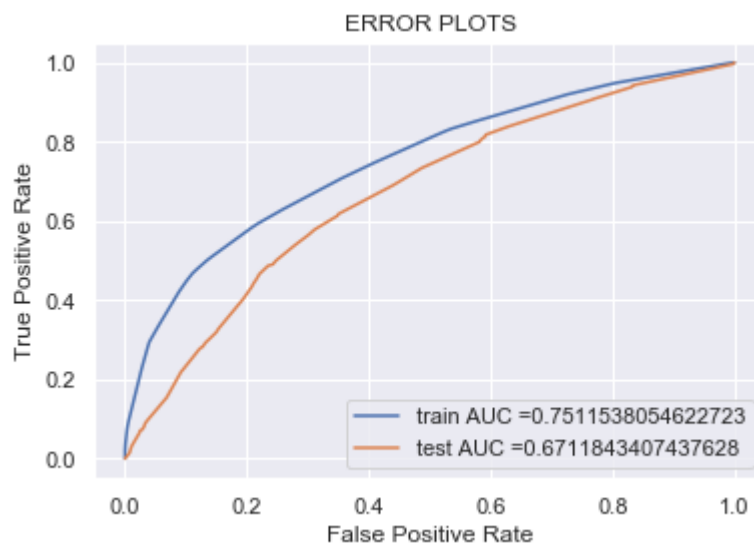
```
Out[160]: dict_keys(['cv', 'error_score', 'estimator__class_weight', 'estimator__criter
ion', 'estimator__max_depth', 'estimator__max_features', 'estimator__max_leaf
_nodes', 'estimator__min_impurity_decrease', 'estimator__min_impurity_split',
'estimator__min_samples_leaf', 'estimator__min_samples_split', 'estimator__mi
n_weight_fraction_leaf', 'estimator__presort', 'estimator__random_state', 'es
timator__splitter', 'estimator', 'iid', 'n_jobs', 'param_grid', 'pre_dispatc
h', 'refit', 'return_train_score', 'scoring', 'verbose'])
```

**Fitting Model to Hyper-Parameter Curve -> Best
Max_depth-> 10 , Best Min_sample_split-> 100**

```

In [205]: # https://scikitlearn.org/stable/modules/generated/sklearn.metrics.roc_curve.html#sklearn.metrics.roc_curve
from sklearn.metrics import roc_curve, auc
from sklearn.tree import DecisionTreeClassifier
from sklearn.metrics import roc_curve, auc
clf11= GridSearchCV( DecisionTreeClassifier(class_weight = 'balanced'),best_tune_parameters)
clfV1=DecisionTreeClassifier (class_weight = 'balanced',max_depth=10,min_samples_split=500)
clf11.fit(X_set1_train, y_train)
# for visulation
clfV1.fit(X_set1_train, y_train)
#https://scikitlearn.org/stable/modules/generated/sklearn.linear_model.SGDClassifier.html#sklearn.linear_model.SGDClassifier.decision_function
y_train_pred1 = clf11.predict_proba(X_set1_train)[:,1]
y_test_pred1 = clf11.predict_proba(X_set1_test)[:,1]
train_fpr1, train_tpr1, tr_thresholds1 = roc_curve(y_train, y_train_pred1)
test_fpr1, test_tpr1, te_thresholds1 = roc_curve(y_test, y_test_pred1)
plt.plot(train_fpr1, train_tpr1, label="train AUC =" +str(auc(train_fpr1, train_tpr1)))
plt.plot(test_fpr1, test_tpr1, label="test AUC =" +str(auc(test_fpr1, test_tpr1)))
plt.legend()
plt.xlabel("False Positive Rate")
plt.ylabel("True Positive Rate")
plt.title("ERROR PLOTS")
plt.grid(True)
plt.show()

```



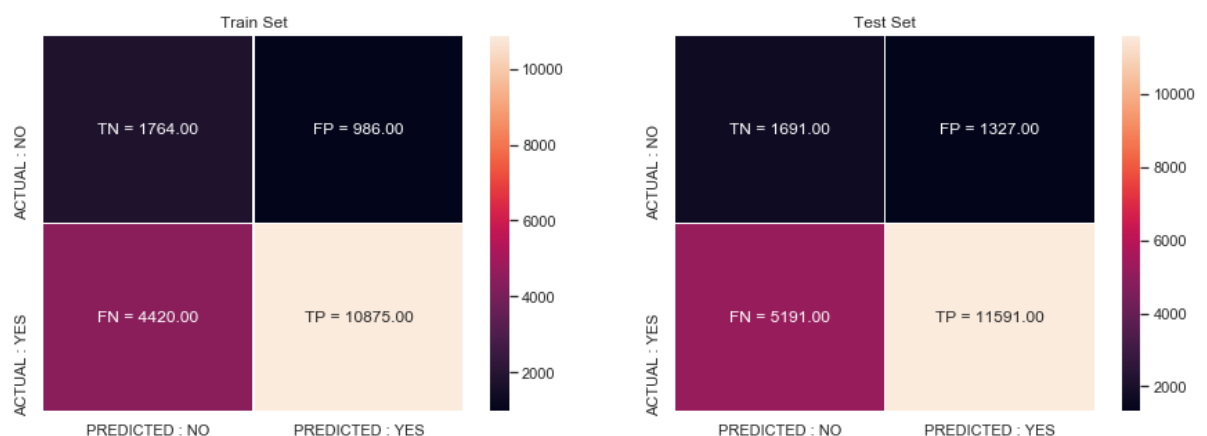
Confusion Matrix

```
In [162]: def predict(proba, threshold, fpr, tpr):
    t = threshold[np.argmax(fpr*(1-tpr))]
    print("the maximum value of tpr*(1-fpr)", np.round(max(tpr*(1-fpr)),2) ,
    "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
```

```
In [163]: #https://www.quantinsti.com/blog/creating-heatmap-using-python-seaborn
import seaborn as sns; sns.set()
con_m_train = confusion_matrix(y_train, predict(y_train_pred1, tr_thresholds1,
train_fpr1, train_tpr1))
con_m_test = confusion_matrix(y_test, predict(y_test_pred1, te_thresholds1, te
st_fpr1, test_tpr1))
key = (np.asarray(['TN', 'FP'], ['FN', 'TP'])))
fig, ax = plt.subplots(1,2, figsize=(15,5))
labels_train = (np.asarray(["{0} = {1:.2f}" .format(key, value) for key, value
in zip(key.flatten(), con_m_train.flatten())])).reshape(2,2)
labels_test = (np.asarray(["{0} = {1:.2f}" .format(key, value) for key, value
in zip(key.flatten(), con_m_test.flatten())])).reshape(2,2)
sns.heatmap(con_m_train, linewidths=.5, xticklabels=['PREDICTED : NO', 'PREDICT
ED : YES'], yticklabels=['ACTUAL : NO', 'ACTUAL : YES'], annot = labels_train,
fmt = '', ax=ax[0])
sns.heatmap(con_m_test, linewidths=.5, xticklabels=['PREDICTED : NO', 'PREDICT
ED : YES'], yticklabels=['ACTUAL : NO', 'ACTUAL : YES'], annot = labels_test, f
mt = '', ax=ax[1])
ax[0].set_title('Train Set')
ax[1].set_title('Test Set')
plt.show()
```

the maximum value of $tpr*(1-fpr)$ 0.47 for threshold 0.44

the maximum value of $tpr*(1-fpr)$ 0.4 for threshold 0.45



Visualizing Decision Tree

```
In [164]: #Feature aggregation
f1=vectorizer1.get_feature_names()
f2=vectorizer2.get_feature_names()
f3=vectorizer3.get_feature_names()
f4=vectorizer4.get_feature_names()
f5=vectorizer5.get_feature_names()
fb=vectorizer6.get_feature_names()
ft=vectorizer7.get_feature_names()
fb1=vectorizer8.get_feature_names()
ft1=vectorizer9.get_feature_names()

feature_agg_bow = f1 + f2 + f3 + f4 + f5 + fb + ft
feature_agg_tfidf = f1 + f2 + f3 + f4 + f5 + fb1 + ft1
# 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')
```

```
In [165]: pip install pydotplus
```

```
Requirement already satisfied: pydotplus in c:\users\hp\anaconda3\lib\site-pa
ckages (2.0.2)
Requirement already satisfied: pyparsing>=2.0.1 in c:\users\hp\anaconda3\lib
\site-packages (from pydotplus) (2.3.1)
Note: you may need to restart the kernel to use updated packages.
```

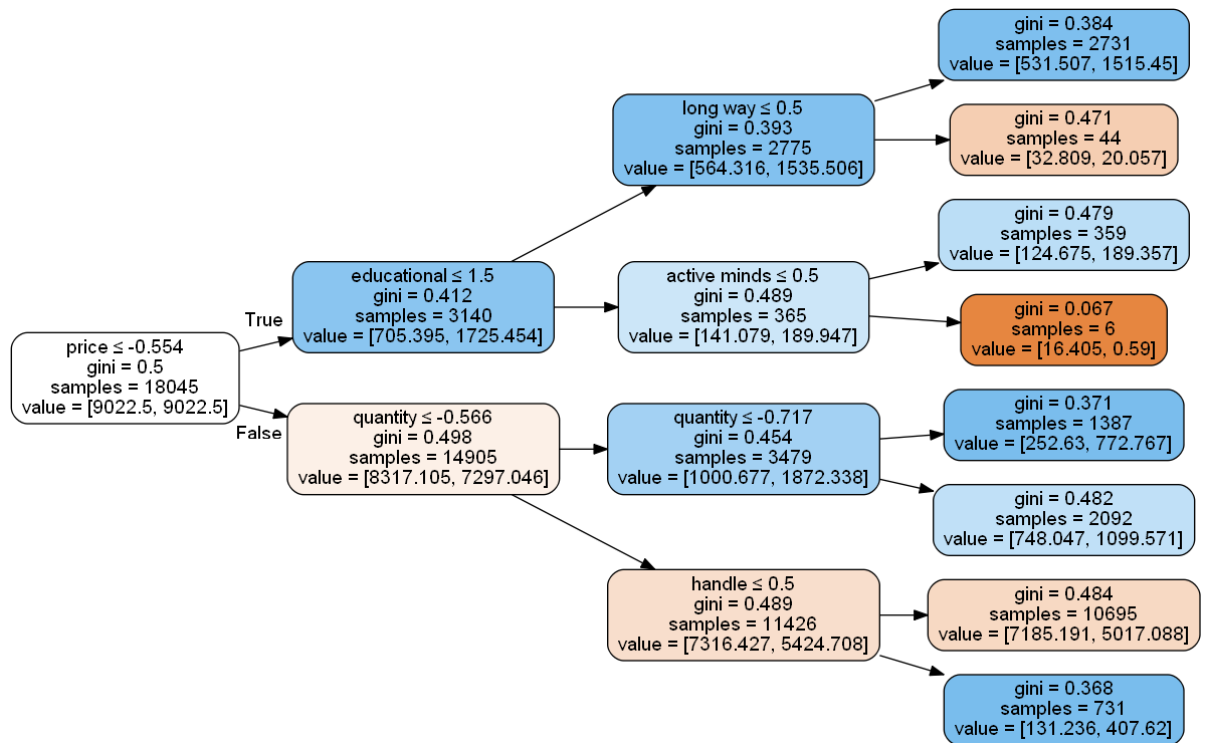


```

In [166]: 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_c
characters=True, feature_names=feature_agg_bow, rotate=True)
graph = pydotplus.graph_from_dot_data(dot_data.getvalue())
Image(graph.create_png())

```

Out[166]:



Analysis on the False positives

In [167]: *#Get the False positives datapoints*
`X_test['essay'].values[1]`

Out[167]: "My classroom is filled with fun-loving special education students that are happy, active, and ready to learn. Their ages range between 3-5 years. All of my students have special needs, including autism, speech and language impairments, and intellectual disabilities, but we don't let that slow us down! We are part of a low income school district on an elementary school campus. Many of my students are nonverbal and need lots of visual and physical supports. We only get a few dollars a year for paper and crayons. We are in desperate need of enrichment supplies! I have a very busy group of students in my preschool special education class. We are so excited, we just can't sit still. We need some special stools to help us move while sitting at the table. This way we can sit with our friends but still keep moving. \\r\\nThe scooter boards will let us twist and shout when we need a break from sitting. The science materials are hands-on to help us move and get kinesthetic input when learning difficult topics. The ball toss will help us develop our gross motor skills and encourage our desire to move in an appropriate and fun way. Help us Move it, Move it!nannan"

In [168]: *#https://www.google.com/search?q=geeks+for+geeks+false+positive&rlz=1C1SQJL_enIN849IN849&oq=geeks+for+geeks+false+positive&aqs=chrome..69i57j33l5.6431j0j7&sourceid=chrome&ie=UTF-8*
#https://github.com/pskadasi/DecisionTrees_DonorsChoose/blob/master/Copy_of_8_DonorsChoose_DT_(1).ipynb
`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['essay'].values[i])`

WORD CLOUD OF ESSAY: Word Cloud is a data visualization technique used for representing text data in which the size of each word indicates its frequency or importance. Significant textual data points can be highlighted using a word cloud. Word clouds are widely used for analyzing data from social network websites.

```
In [169]: 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 [170]: # 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)

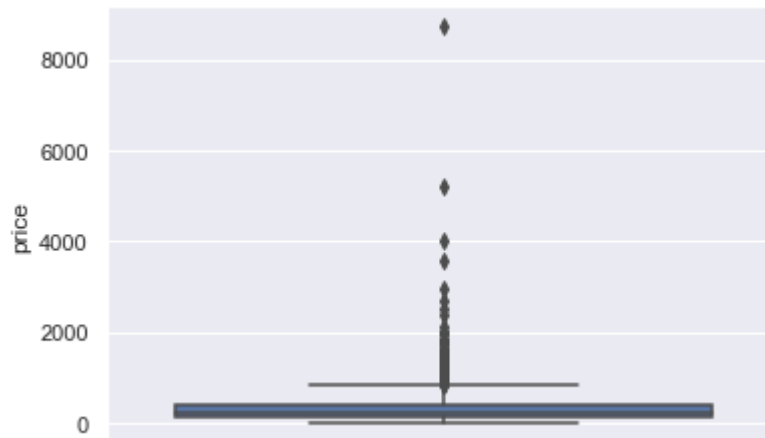
    X_test_falsePos1 = X_test_falsePos1.append(X_test.filter(items=[i], axis=0))

X_test_falsePos1.head(1)
len(X_test_falsePos1)
```

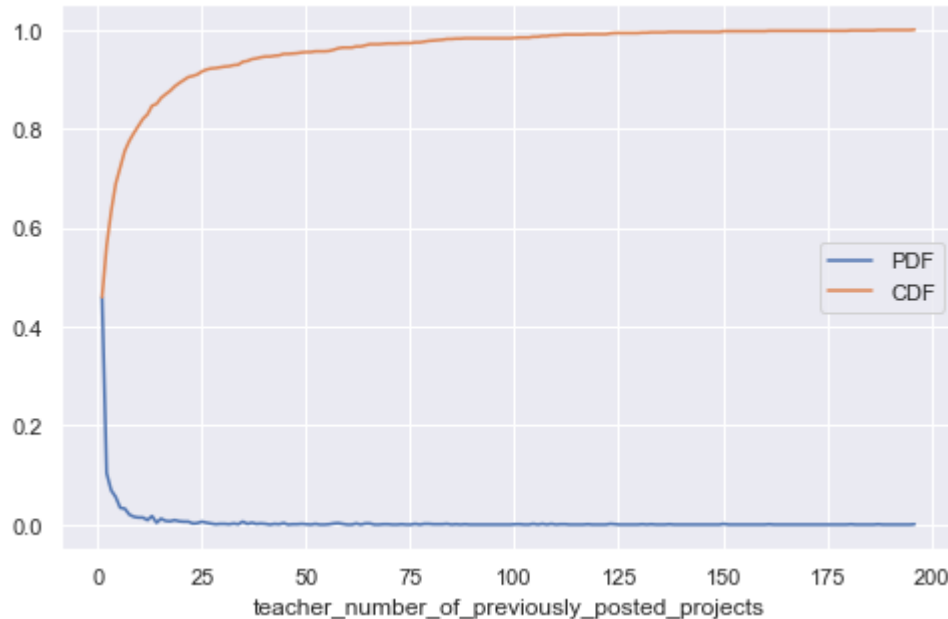
Out[170]: 1327

```
In [171]: ##Box Plot (FP 'price')
sns.boxplot(y='price', data=X_test_falsePos1)
```

Out[171]: <matplotlib.axes._subplots.AxesSubplot at 0x16d2c8e7588>



```
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
y_posted_projects'],
bins='auto', density=True)
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()
```

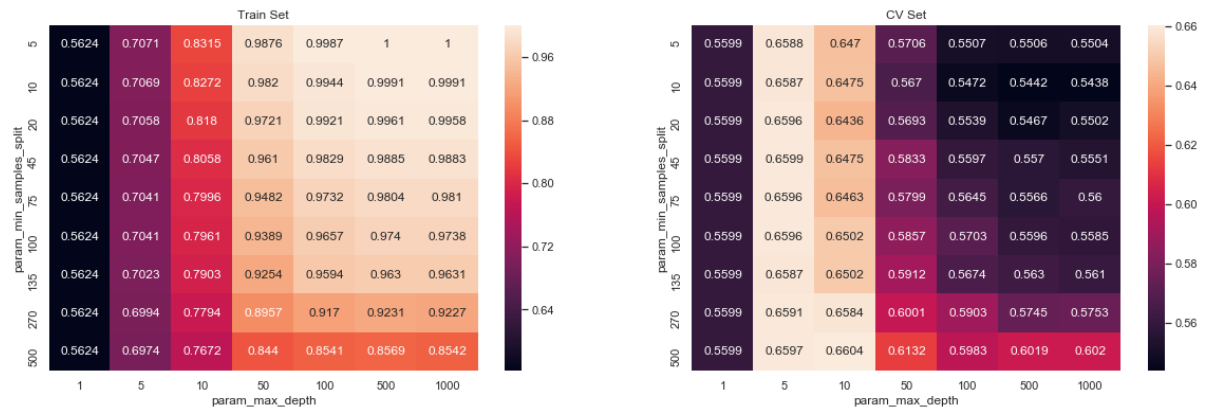


Applying Decision trees on TFIDF

```
In [84]: from sklearn.metrics import roc_auc_score
import matplotlib.pyplot as plt
from sklearn.model_selection import train_test_split
from sklearn.model_selection import GridSearchCV
from sklearn.model_selection import cross_val_score
from sklearn.tree import DecisionTreeClassifier
dt2 = DecisionTreeClassifier(class_weight = 'balanced')
parameters = {'max_depth': [1, 5, 10, 50, 100, 500, 1000], 'min_samples_split'
: [5, 10, 20, 45, 75, 100, 135, 270, 500]}
clf2 = GridSearchCV(dt2, parameters, cv=3, scoring='roc_auc', return_train_score=True)
se2 = clf2.fit(X_set2_train, y_train)
```

```
In [85]: import seaborn as sns; sns.set()
max_scores1 = pd.DataFrame(clf2.cv_results_).groupby(['param_min_samples_split', 'param_max_depth']).max().unstack()['mean_test_score', 'mean_train_score']

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()
```



Best Estimator and Best tune parameters

```
In [86]: print(clf2.best_estimator_)
#Mean cross-validated score of the best_estimator
print(clf2.score(X_set2_train,y_train))
print(clf2.score(X_set2_test,y_test))
```

```
DecisionTreeClassifier(class_weight='balanced', criterion='gini', max_depth=1
0,
```

```
max_features=None, max_leaf_nodes=None,
min_impurity_decrease=0.0, min_impurity_split=None,
min_samples_leaf=1, min_samples_split=500,
min_weight_fraction_leaf=0.0, presort=False,
random_state=None, splitter='best')
```

```
0.7628998900413088
0.6658739514606636
```

```
In [173]: # Best tune parameters
best_tune_parameters=[{'max_depth':[10], 'min_samples_split':[500] }]
```

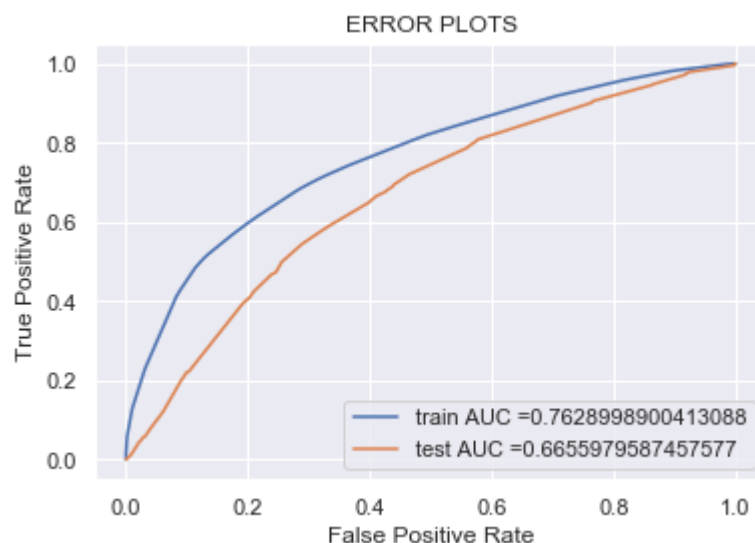
```

In [206]: ##Fitting Model to Hyper-Parameter Curve
# https://scikitlearn.org/stable/modules/generated/sklearn.metrics.roc_curve.html#sklearn.metrics.roc_curve
from sklearn.metrics import roc_curve, auc
from sklearn.tree import DecisionTreeClassifier
from sklearn.metrics import roc_curve, auc
clf11= GridSearchCV( DecisionTreeClassifier(class_weight = 'balanced'),best_tu
ne_parameters)
clfV1=DecisionTreeClassifier (class_weight = 'balanced',max_depth=10,min_sampl
es_split=500)
clf11.fit(X_set2_train, y_train)
# for visulation
clfV1.fit(X_set2_train, y_train)
#https://scikitlearn.org/stable/modules/generated/sklearn.linear_model.SGDClassifier.html#sklearn.linear_model.SGDClassifier.decision_function
y_train_pred1 = clf11.predict_proba(X_set2_train)[:,1]
y_test_pred1 = clf11.predict_proba(X_set2_test)[:,1]

train_fpr1, train_tpr1, tr_thresholds1 = roc_curve(y_train, y_train_pred1)
test_fpr1, test_tpr1, te_thresholds1 = roc_curve(y_test, y_test_pred1)

plt.plot(train_fpr1, train_tpr1, label="train AUC =" +str(auc(train_fpr1, train
_tpr1)))
plt.plot(test_fpr1, test_tpr1, label="test AUC =" +str(auc(test_fpr1, test_tpr1
)))
plt.legend()
plt.xlabel("False Positive Rate")
plt.ylabel("True Positive Rate")
plt.title("ERROR PLOTS")
plt.grid(True)
plt.show()

```



Confusion matrix

```
In [175]: import seaborn as sns; sns.set()
con_m_train = confusion_matrix(y_train, predict(y_train_pred1, tr_thresholds1,
train_fpr1, train_tpr1))
con_m_test = confusion_matrix(y_test, predict(y_test_pred1, te_thresholds1, te
st_fpr1, test_tpr1))
key = (np.asarray(['TN', 'FP'], ['FN', 'TP'])))
fig, ax = plt.subplots(1,2, figsize=(15,5))
labels_train = (np.asarray(["{0} = {1:.2f}" .format(key, value) for key, value
in zip(key.flatten(), con_m_train.flatten())])).reshape(2,2)
labels_test = (np.asarray(["{0} = {1:.2f}" .format(key, value) for key, value
in zip(key.flatten(),
con_m_test.flatten())])).reshape(2,2)
sns.heatmap(con_m_train, linewidths=.5, xticklabels=['PREDICTED : NO', 'PREDIC
TED : YES'], yticklabels=['ACTUAL : NO', 'ACTUAL : YES'], annot = labels_train,
fmt = '', ax=ax[0])
sns.heatmap(con_m_test, linewidths=.5, xticklabels=['PREDICTED : NO', 'PREDICT
ED : YES'], yticklabels=['ACTUAL : NO', 'ACTUAL : YES'], annot = labels_test, f
mt = '', ax=ax[1])
ax[0].set_title('Train Set')
ax[1].set_title('Test Set')
plt.show()
```

the maximum value of $tpr*(1-fpr)$ 0.49 for threshold 0.39

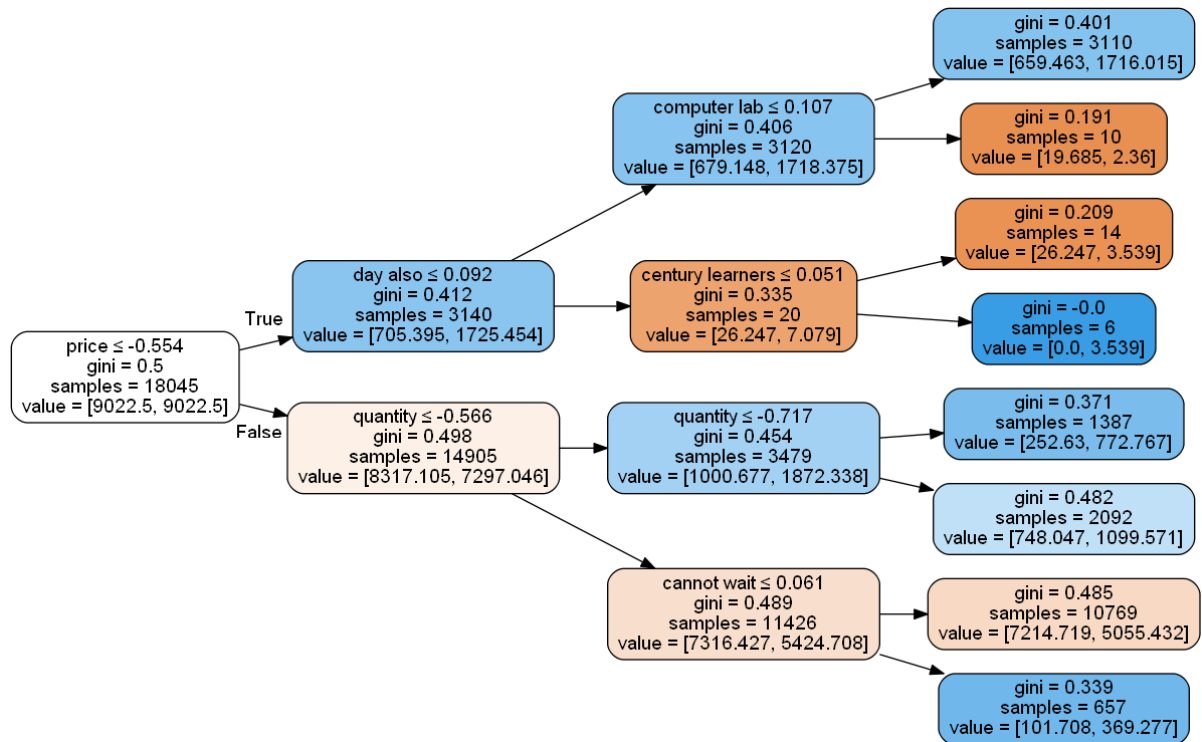
the maximum value of $tpr*(1-fpr)$ 0.39 for threshold 0.5



Visualizing Decision Tree


```
In [176]: 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, feature_names=feature_agg_bow, rotate=True)
graph = pydotplus.graph_from_dot_data(dot_data.getvalue())
Image(graph.create_png())
```

Out[176]:



Analysis on the False positives

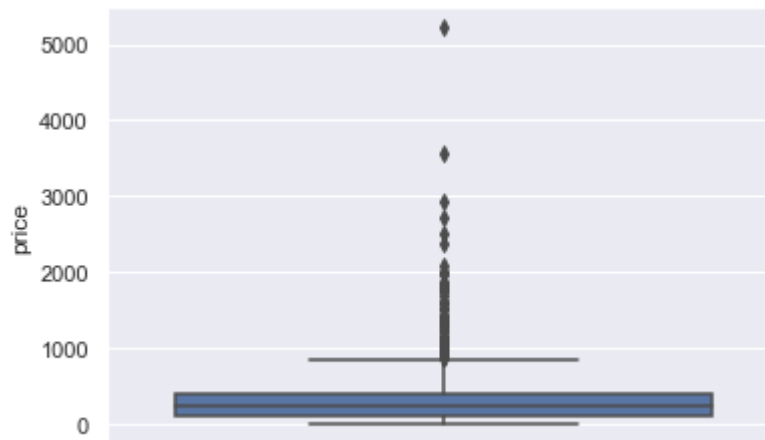
```
In [177]: #https://www.google.com/search?q=geeks+for+geeks+false+positive&rlz=1C1SQJL_en
IN849IN849&oq=geeks+for+geeks+false+positive&aqs=chrome..69i57j33l5.6431j0j7&s
ourceid=chrome&ie=UTF-8
#https://github.com/pskadasi/DecisionTrees_DonorsChoose/blob/master/Copy_of_8_
DonorsChoose_DT_(1).ipynb
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['essay'].values[i])
```



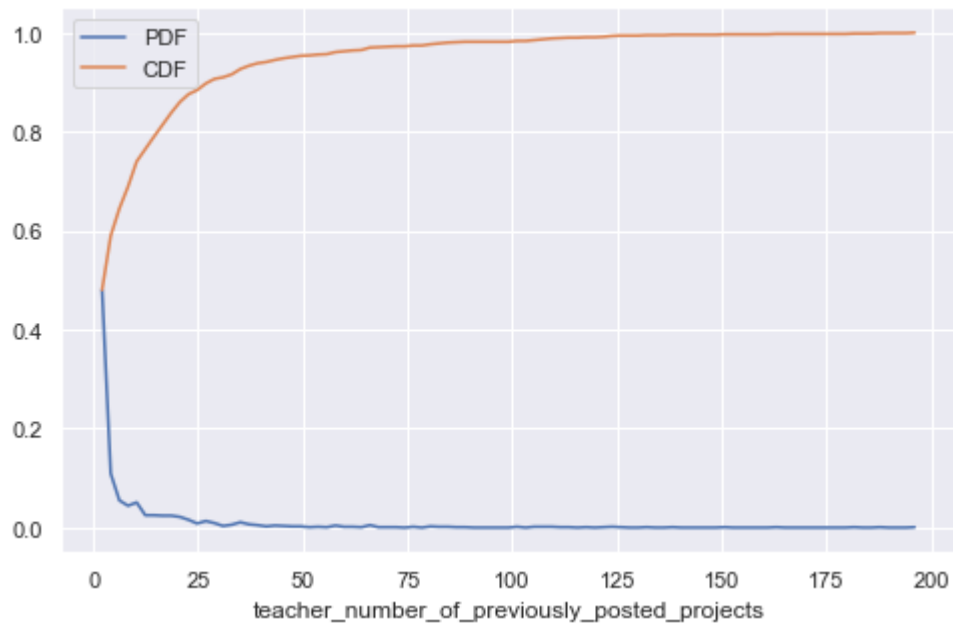
```
In [179]: # first get the columns:
cols = X_test.columns
X_test_falsePos1 = pd.DataFrame(columns=cols)
# get the data of the false pisitives
for i in fpi : # (in fpi all the false positives data points indexes)
    X_test_falsePos1 = X_test_falsePos1.append(X_test.filter(items=[i], axis=0))
```

```
In [180]: #Box Plot (FP 'price')
sns.boxplot(y='price', data=X_test_falsePos1)
```

Out[180]: <matplotlib.axes._subplots.AxesSubplot at 0x16bc09070b8>



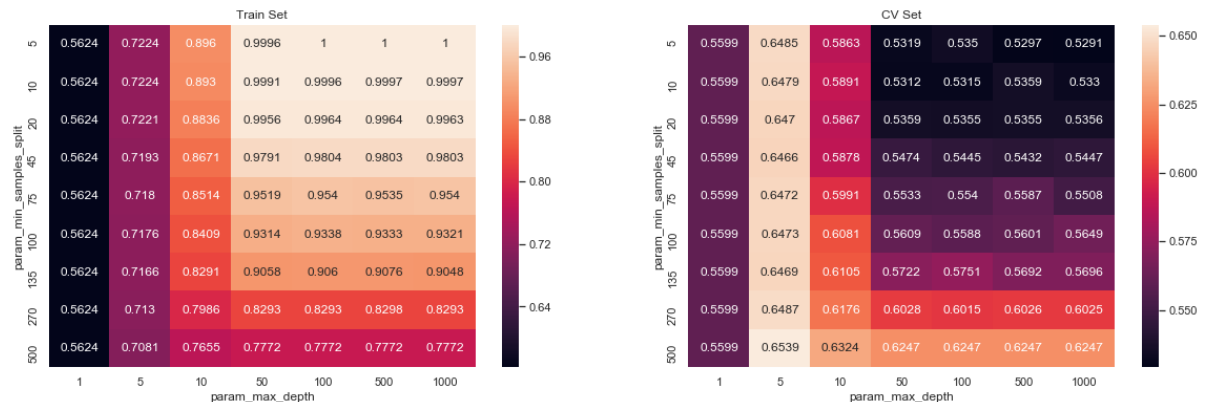
```
In [181]: #PDF (FP ,teacher_number_of_previously_posted_projects)
plt.figure(figsize=(8,5))
counts, bin_edges = np.histogram(X_test_falsePos1['teacher_number_of_previously_
y_posted_projects'],bins='auto', density=True)
pdf = counts/sum(counts)
cdf = np.cumsum(pdf)
pdfP, = plt.plot(bin_edges[1:], pdf)
cdfP, = plt.plot(bin_edges[1:], cdf)
plt.legend([pdfP, cdfP], ["PDF", "CDF"])
plt.xlabel('teacher_number_of_previously_posted_projects')
plt.show()
```



Applying Decision trees on AVG W2V

```
In [96]: from sklearn.metrics import roc_auc_score
import matplotlib.pyplot as plt
from sklearn.model_selection import train_test_split
from sklearn.model_selection import GridSearchCV
from sklearn.model_selection import cross_val_score
from sklearn.tree import DecisionTreeClassifier
dt3= DecisionTreeClassifier(class_weight = 'balanced')
parameters = {'max_depth': [1, 5, 10, 50, 100, 500, 1000], 'min_samples_split'
: [5, 10, 20, 45, 75, 100, 135, 270, 500]}
clf3 = GridSearchCV(dt3, parameters, cv=3, scoring='roc_auc',n_jobs=4,return_t
rain_score=True)
se3 = clf3.fit(X_set3_train, y_train)

import seaborn as sns; sns.set()
max_scores1 = pd.DataFrame(clf3.cv_results_).groupby(['param_min_samples_spli
t', 'param_max_depth']).max().unstack()['mean_test_score', 'mean_train_score'
]]
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()
```



```
In [97]: #Best Estimator and Best tune parameters
print(clf3.best_estimator_)
#Mean cross-validated score of the best estimator
print(clf3.score(X_set3_train,y_train))
print(clf3.score(X_set3_test,y_test))
```

```
DecisionTreeClassifier(class_weight='balanced', criterion='gini', max_depth=
5,
```

```
max_features=None, max_leaf_nodes=None,
min_impurity_decrease=0.0, min_impurity_split=None,
min_samples_leaf=1, min_samples_split=500,
min_weight_fraction_leaf=0.0, presort=False,
random_state=None, splitter='best')
```

```
0.6965402121905555
```

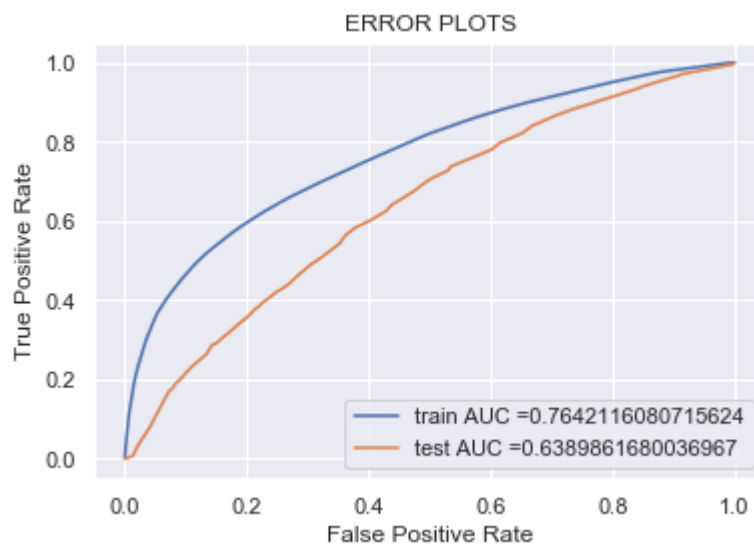
```
0.6437607521359746
```

```
In [182]: # Best tune parameters
best_tune_parameters=[{'max_depth':[5], 'min_samples_split':[500] } ]
```

```

In [207]: #Fitting Model to Hyper-Parameter Curve
# https://scikitlearn.org/stable/modules/generated/sklearn.metrics.roc_curve.html#sklearn.metrics.roc_curve
from sklearn.metrics import roc_curve, auc
from sklearn.tree import DecisionTreeClassifier
from sklearn.metrics import roc_curve, auc
clf11= GridSearchCV( DecisionTreeClassifier(class_weight = 'balanced'),best_tune_parameters)
clfV1=DecisionTreeClassifier (class_weight = 'balanced',max_depth=5,min_samples_split=500)
clf11.fit(X_set3_train, y_train)
# for visulation
clfV1.fit(X_set3_train, y_train)
#https://scikitlearn.org/stable/modules/generated/sklearn.linear_model.SGDClassifier.html#sklearn.linear_model.SGDClassifier.decision_function
y_train_pred1 = clf11.predict_proba(X_set3_train)[:,1]
y_test_pred1 = clf11.predict_proba(X_set3_test)[:,1]
train_fpr1, train_tpr1, tr_thresholds1 = roc_curve(y_train, y_train_pred1)
test_fpr1, test_tpr1, te_thresholds1 = roc_curve(y_test, y_test_pred1)
plt.plot(train_fpr1, train_tpr1, label="train AUC =" +str(auc(train_fpr1, train_tpr1)))
plt.plot(test_fpr1, test_tpr1, label="test AUC =" +str(auc(test_fpr1, test_tpr1)))
plt.legend()
plt.xlabel("False Positive Rate")
plt.ylabel("True Positive Rate")
plt.title("ERROR PLOTS")
plt.grid(True)
plt.show()

```



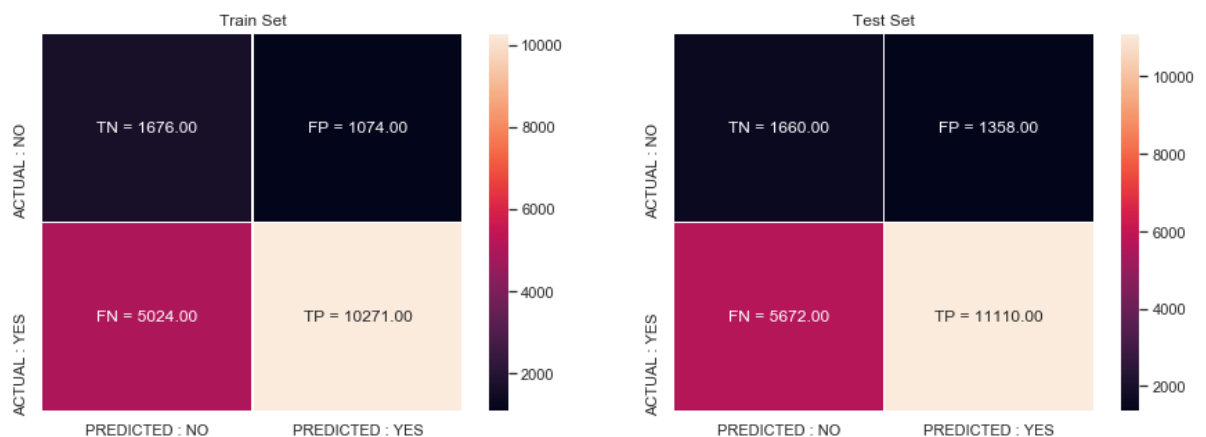
```

In [184]: #confusion matrix test data
#https://www.quantinsti.com/blog/creating-heatmap-using-python-seaborn
import seaborn as sns; sns.set()
con_m_train = confusion_matrix(y_train, predict(y_train_pred1, tr_thresholds1,
train_fpr1, train_tpr1))
con_m_test = confusion_matrix(y_test, predict(y_test_pred1, te_thresholds1, te
st_fpr1, test_tpr1))
key = (np.asarray(['TN', 'FP'], ['FN', 'TP'])))
fig, ax = plt.subplots(1,2, figsize=(15,5))
labels_train = (np.asarray(["{0} = {1:.2f}" .format(key, value) for key, value
in zip(key.flatten(), con_m_train.flatten())])).reshape(2,2)
labels_test = (np.asarray(["{0} = {1:.2f}" .format(key, value) for key, value
in zip(key.flatten(), con_m_test.flatten())])).reshape(2,2)
sns.heatmap(con_m_train, linewidths=.5, xticklabels=['PREDICTED : NO', 'PREDICT
ED : YES'],
yticklabels=['ACTUAL : NO', 'ACTUAL : YES'], annot = labels_train, fmt = '', a
x=ax[0])
sns.heatmap(con_m_test, linewidths=.5, xticklabels=['PREDICTED : NO', 'PREDICT
ED : YES'],
yticklabels=['ACTUAL : NO', 'ACTUAL : YES'], annot = labels_test, fmt = '', ax
=ax[1])
ax[0].set_title('Train Set')
ax[1].set_title('Test Set')
plt.show()

```

the maximum value of $tpr \cdot (1 - fpr)$ 0.41 for threshold 0.47

the maximum value of $tpr \cdot (1 - fpr)$ 0.37 for threshold 0.47



```

In [185]: ##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['essay'].values[i])

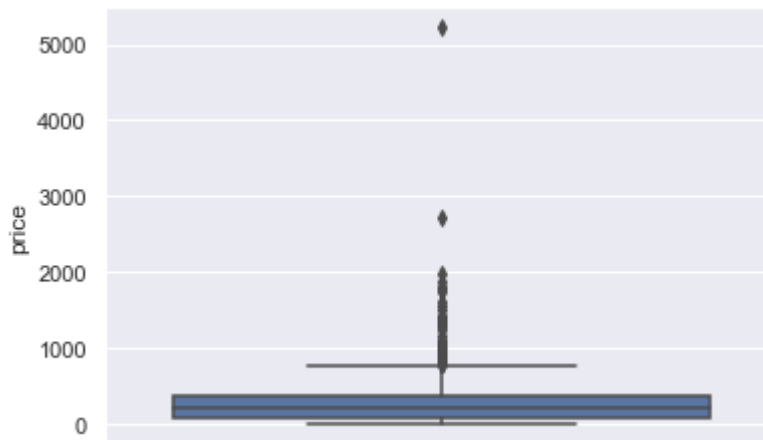
```



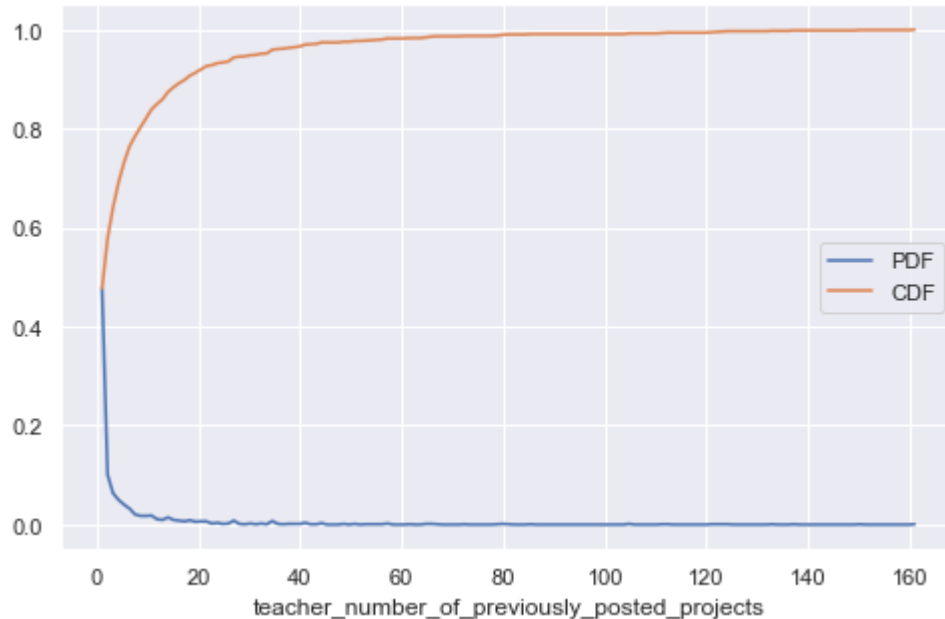
```
In [188]: #DataFrame of False Positives
# first get the columns:
cols = X_test.columns
X_test_falsePos1 = pd.DataFrame(columns=cols)
# get the data of the false positives
for i in fpi : # (in fpi all the false positives data points indexes)
    X_test_falsePos1 = X_test_falsePos1.append(X_test.filter(items=[i], axis=0))
```

```
In [189]: #Box Plot (FP 'price')
sns.boxplot(y='price', data=X_test_falsePos1)
```

Out[189]: <matplotlib.axes._subplots.AxesSubplot at 0x16bb87b9240>



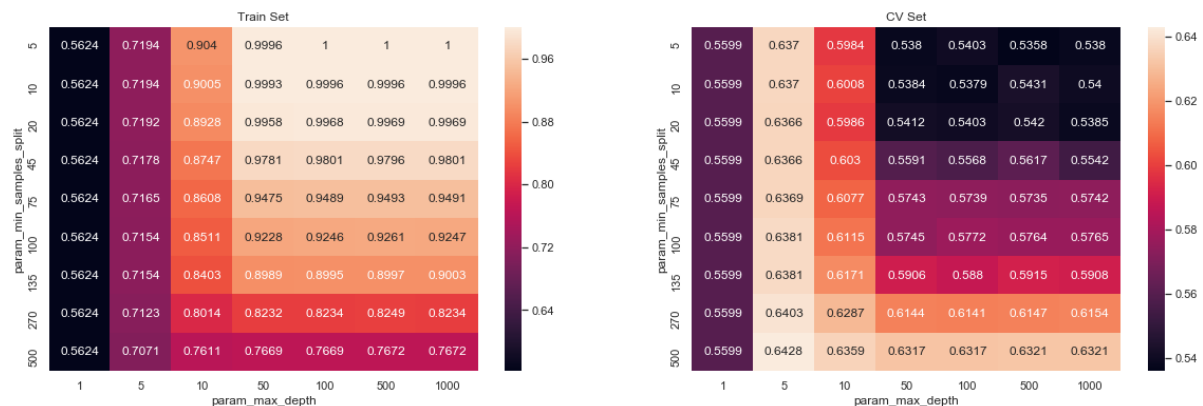
```
In [190]: #PDF (FP ,teacher_number_of_previously_posted_projects)
plt.figure(figsize=(8,5))
counts, bin_edges = np.histogram(X_test_falsePos1['teacher_number_of_previously
y_posted_projects'],
bins='auto', density=True)
pdf = counts/sum(counts)
cdf = np.cumsum(pdf)
pdfP, = plt.plot(bin_edges[1:], pdf)
cdfP, = plt.plot(bin_edges[1:], cdf)
plt.legend([pdfP, cdfP], ["PDF", "CDF"])
plt.xlabel('teacher_number_of_previously_posted_projects')
plt.show()
```



Applying Decision trees on td_idf W2V

```
In [107]: from sklearn.metrics import roc_auc_score
import matplotlib.pyplot as plt
from sklearn.model_selection import train_test_split
from sklearn.model_selection import GridSearchCV
from sklearn.model_selection import cross_val_score
from sklearn.tree import DecisionTreeClassifier
dt4= DecisionTreeClassifier(class_weight = 'balanced')
parameters = {'max_depth': [1, 5, 10, 50, 100, 500, 1000], 'min_samples_split':
: [5, 10, 20, 45, 75, 100, 135, 270, 500]}
clf4 = GridSearchCV(dt4, parameters, cv=3, scoring='roc_auc',return_train_score=True)
set4= clf4.fit(X_set4_train, y_train)

import seaborn as sns; sns.set()
max_scores1 = pd.DataFrame(clf4.cv_results_).groupby(['param_min_samples_split', 'param_max_depth']).max().unstack()[['mean_test_score', 'mean_train_score']]
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()
```



```
In [108]: #Best Estimator and Best tune parameters
print(clf4.best_estimator_)
#Mean cross-validated score of the best_estimator
print(clf4.score(X_set4_train,y_train))
print(clf4.score(X_set4_test,y_test))
```

```
DecisionTreeClassifier(class_weight='balanced', criterion='gini', max_depth=5,
```

```
max_features=None, max_leaf_nodes=None,
min_impurity_decrease=0.0, min_impurity_split=None,
min_samples_leaf=1, min_samples_split=500,
min_weight_fraction_leaf=0.0, presort=False,
random_state=None, splitter='best')
```

```
0.6969748759249904
```

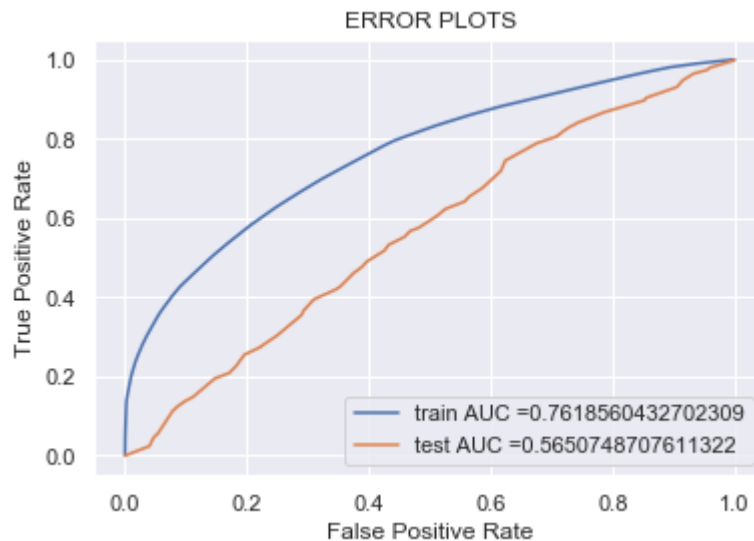
```
0.6138512151971972
```

```
In [191]: best_tune_parameters= [{'max_depth': [5], 'min_samples_split':[500] }]
```

```

In [208]: #Fitting Model to Hyper-Parameter Curve
from sklearn.metrics import roc_curve, auc
from sklearn.tree import DecisionTreeClassifier
from sklearn.metrics import roc_curve, auc
clf11= GridSearchCV( DecisionTreeClassifier(class_weight = 'balanced'),best_tu
ne_parameters)
clfV1=DecisionTreeClassifier (class_weight = 'balanced',max_depth=5,min_sample
s_split=500)
clf11.fit(X_set4_train, y_train)
# for visulation
clfV1.fit(X_set4_train, y_train)
#https://scikitlearn.org/stable/modules/generated/sklearn.linear_model.SGDClas
sifier.html#sklearn.linear_model.SGDClassifier.decision_function
y_train_pred1 = clf11.predict_proba(X_set4_train)[:,1]
y_test_pred1 = clf11.predict_proba(X_set4_test)[:,1]
train_fpr1, train_tpr1, tr_thresholds1 = roc_curve(y_train, y_train_pred1)
test_fpr1, test_tpr1, te_thresholds1 = roc_curve(y_test, y_test_pred1)
plt.plot(train_fpr1, train_tpr1, label="train AUC =" +str(auc(train_fpr1, train
_tpr1)))
plt.plot(test_fpr1, test_tpr1, label="test AUC =" +str(auc(test_fpr1, test_tpr1
)))
plt.legend()
plt.xlabel("False Positive Rate")
plt.ylabel("True Positive Rate")
plt.title("ERROR PLOTS")
plt.grid(True)
plt.show()

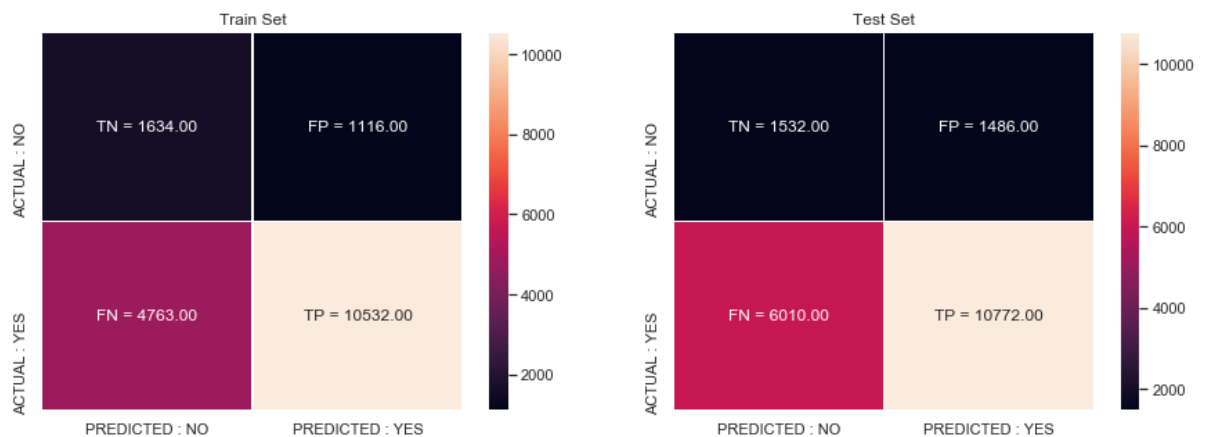
```



```
In [193]: #CONFUSION MATRIX
#https://www.quantinsti.com/blog/creating-heatmap-using-python-seaborn
import seaborn as sns; sns.set()
con_m_train = confusion_matrix(y_train, predict(y_train_pred1, tr_thresholds1,
train_fpr1, train_tpr1))
con_m_test = confusion_matrix(y_test, predict(y_test_pred1, te_thresholds1, te
st_fpr1, test_tpr1))
key = (np.asarray(['TN', 'FP'], ['FN', 'TP'])))
fig, ax = plt.subplots(1,2, figsize=(15,5))
labels_train = (np.asarray(["{0} = {1:.2f}" .format(key, value) for key, value
in zip(key.flatten(), con_m_train.flatten())])).reshape(2,2)
labels_test = (np.asarray(["{0} = {1:.2f}" .format(key, value) for key, value
in zip(key.flatten(), con_m_test.flatten())])).reshape(2,2)
sns.heatmap(con_m_train, linewidths=.5, xticklabels=['PREDICTED : NO', 'PREDIC
TED : YES'], yticklabels=['ACTUAL : NO', 'ACTUAL : YES'], annot = labels_train,
fmt = '', ax=ax[0])
sns.heatmap(con_m_test, linewidths=.5, xticklabels=['PREDICTED : NO', 'PREDICT
ED : YES'], yticklabels=['ACTUAL : NO', 'ACTUAL : YES'], annot = labels_test, f
mt = '', ax=ax[1])
ax[0].set_title('Train Set')
ax[1].set_title('Test Set')
plt.show()
```

the maximum value of $tpr*(1-fpr)$ 0.42 for threshold 0.45

the maximum value of $tpr*(1-fpr)$ 0.33 for threshold 0.5

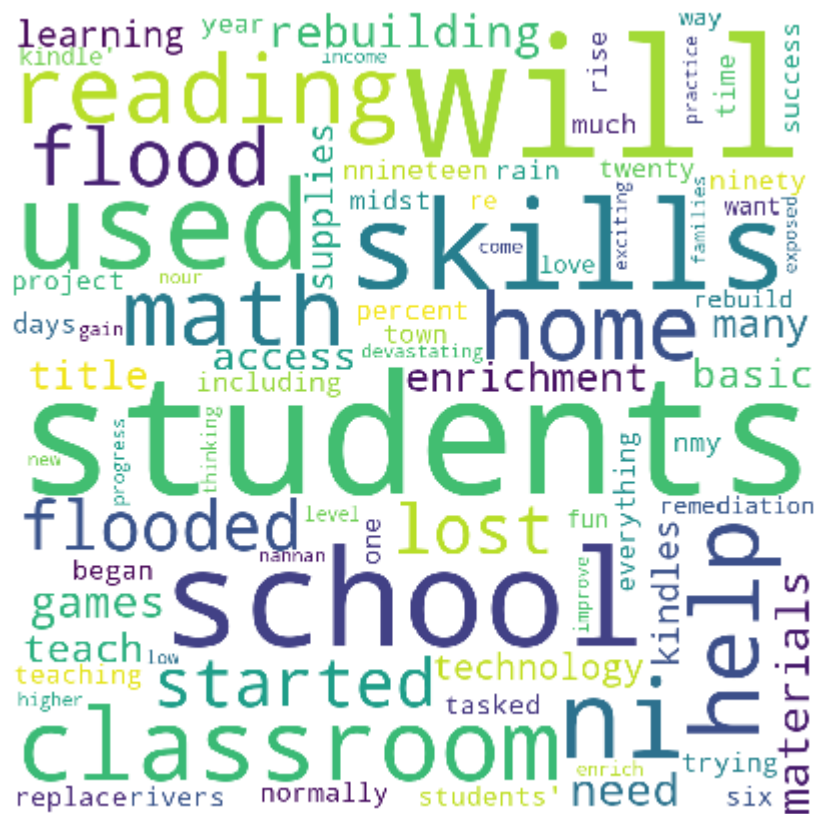


```
In [194]: #Analysis on the False positives
fpi = []
for i in range(len(y_test)) :
    if (y_test.values[i] == 0) & (predictions1[i] == 1) :
        fpi.append(i)
fp_essay1 = []
for i in fpi :
    fp_essay1.append(X_test['essay'].values[i])
```

```
#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 + ' '

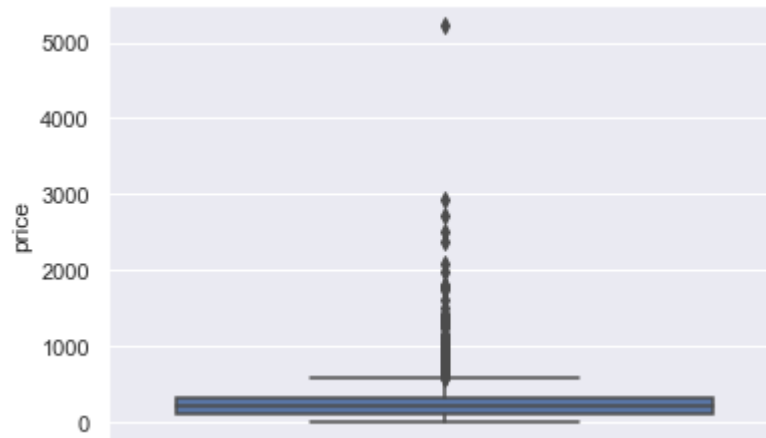
wordcloud = WordCloud(width = 800, height = 800, background_color ='white', stopwords = stopwords,
min_font_size = 10).generate(comment_words)

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

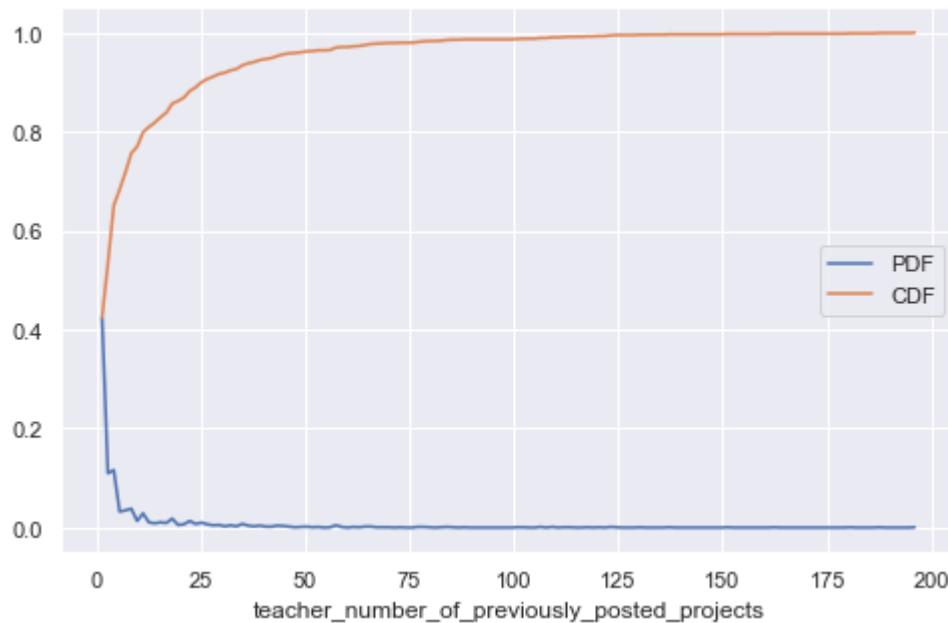


```
In [196]: #Box Plot (FP 'price')
# first get the columns:
cols = X_test.columns
X_test_falsePos1 = pd.DataFrame(columns=cols)
# get the data of the false pisitives
for i in fpi : # (in fpi all the false positives data points indexes)
    X_test_falsePos1 = X_test_falsePos1.append(X_test.filter(items=[i], axis=0))
sns.boxplot(y='price', data=X_test_falsePos1)
```

Out[196]: <matplotlib.axes._subplots.AxesSubplot at 0x16d281599b0>



```
In [197]: #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='auto', density=True)
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()
```



Select 5k best features from features of Set 2 using feature_importances, 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

```
In [130]: #https://stackoverflow.com/questions/47111434/randomforestregressor-and-feature-importances-error
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 [146]: # for tf-idf set 2
X_set5_train = selectKImportance(clf2, xtr, 5000)
X_set5_test = selectKImportance(clf2, X_set2_test, 5000)

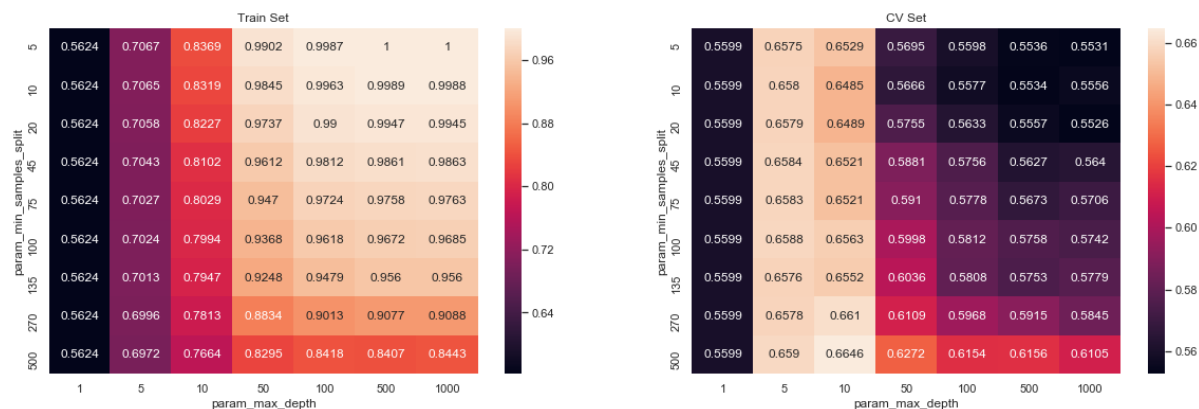
print(X_set5_train.shape)
print(X_set5_test.shape)
```

```
(18045, 5000)
(19800, 5000)
```


Decision tree on Important features

```
In [147]: from sklearn.metrics import roc_auc_score
import matplotlib.pyplot as plt
from sklearn.model_selection import train_test_split
from sklearn.model_selection import GridSearchCV
from sklearn.model_selection import cross_val_score
from sklearn.tree import DecisionTreeClassifier
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, 270, 500]}
clf5 = GridSearchCV(dt5, parameters, cv=3, scoring='roc_auc', return_train_score=True)
set5= clf5.fit(X_set5_train, y_train)

import seaborn as sns; sns.set()
max_scores1 = pd.DataFrame(clf5.cv_results_).groupby(['param_min_samples_split', 'param_max_depth']).max().unstack()['mean_test_score', 'mean_train_score']
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()
```



```
In [148]: #Best Estimator and Best tune parameters  
print(clf5.best_estimator_)  
#Mean cross-validated score of the best_estimator  
print(clf5.score(X_set5_train,y_train))  
print(clf5.score(X_set5_test,y_test))
```

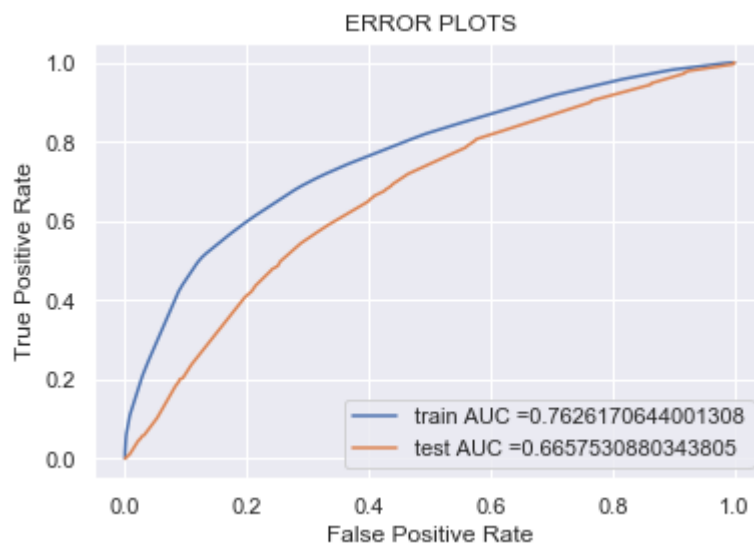
```
DecisionTreeClassifier(class_weight='balanced', criterion='gini', max_depth=1  
0,  
                        max_features=None, max_leaf_nodes=None,  
                        min_impurity_decrease=0.0, min_impurity_split=None,  
                        min_samples_leaf=1, min_samples_split=500,  
                        min_weight_fraction_leaf=0.0, presort=False,  
                        random_state=None, splitter='best')  
0.7628998900413088  
0.6659494725920092
```

```
In [198]: # Best tune parameters  
best_tune_parameters=[{ 'max_depth': [10], 'min_samples_split':[500] } ]
```

```

In [209]: # train with best hyperparameter
# https://scikitlearn.org/stable/modules/generated/sklearn.metrics.roc_curve.html#sklearn.metrics.roc_curve
from sklearn.metrics import roc_curve, auc
from sklearn.tree import DecisionTreeClassifier
from sklearn.metrics import roc_curve, auc
clf11= GridSearchCV( DecisionTreeClassifier(class_weight = 'balanced'),best_tune_parameters)
clfV1=DecisionTreeClassifier (class_weight = 'balanced',max_depth=10,min_samples_split=500)
clf11.fit(X_set5_train, y_train)
# for visulation
clfV1.fit(X_set5_train, y_train)
#https://scikitlearn.org/stable/modules/generated/sklearn.linear_model.SGDClassifier.html#sklearn.linear_model.SGDClassifier.decision_function
y_train_pred1 = clf11.predict_proba(X_set5_train)[:,1]
y_test_pred1 = clf11.predict_proba(X_set5_test)[:,1]
train_fpr1, train_tpr1, tr_thresholds1 = roc_curve(y_train, y_train_pred1)
test_fpr1, test_tpr1, te_thresholds1 = roc_curve(y_test, y_test_pred1)
plt.plot(train_fpr1, train_tpr1, label="train AUC =" +str(auc(train_fpr1, train_tpr1)))
plt.plot(test_fpr1, test_tpr1, label="test AUC =" +str(auc(test_fpr1, test_tpr1)))
plt.legend()
plt.xlabel("False Positive Rate")
plt.ylabel("True Positive Rate")
plt.title("ERROR PLOTS")
plt.grid(True)
plt.show()

```



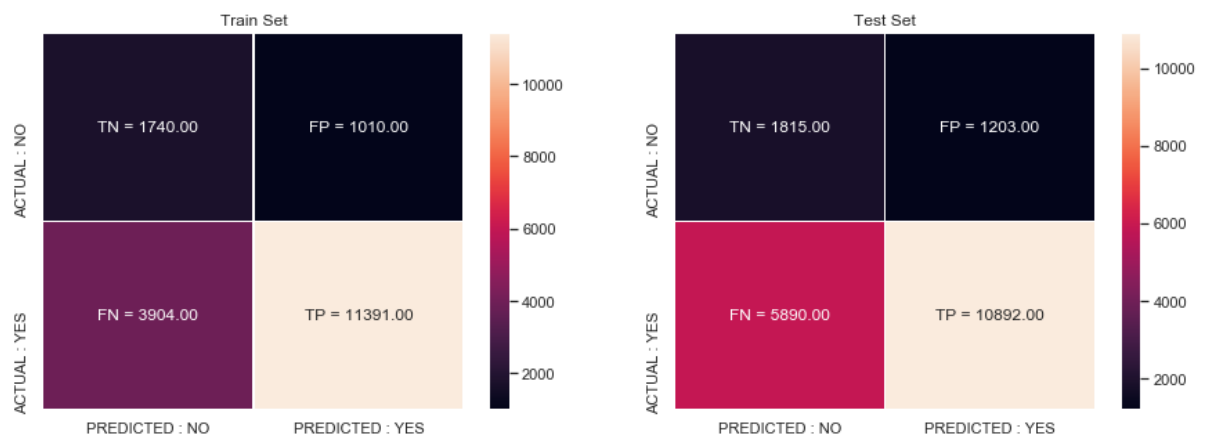
```

In [200]: #CONFUSION MATRIX
#https://www.quantinsti.com/blog/creating-heatmap-using-python-seaborn
import seaborn as sns; sns.set()
con_m_train = confusion_matrix(y_train, predict(y_train_pred1, tr_thresholds1,
train_fpr1, train_tpr1))
con_m_test = confusion_matrix(y_test, predict(y_test_pred1, te_thresholds1, te
st_fpr1, test_tpr1))
key = (np.asarray(['TN', 'FP'], ['FN', 'TP'])))
fig, ax = plt.subplots(1,2, figsize=(15,5))
labels_train = (np.asarray(["{0} = {1:.2f}" .format(key, value) for key, value
in zip(key.flatten(), con_m_train.flatten())])).reshape(2,2)
labels_test = (np.asarray(["{0} = {1:.2f}" .format(key, value) for key, value
in zip(key.flatten(),
con_m_test.flatten())])).reshape(2,2)
sns.heatmap(con_m_train, linewidths=.5, xticklabels=['PREDICTED : NO', 'PREDIC
TED : YES'], yticklabels=['ACTUAL : NO', 'ACTUAL : YES'], annot = labels_train,
fmt = '', ax=ax[0])
sns.heatmap(con_m_test, linewidths=.5, xticklabels=['PREDICTED : NO', 'PREDICT
ED : YES'], yticklabels=['ACTUAL : NO', 'ACTUAL : YES'], annot = labels_test, f
mt = '', ax=ax[1])
ax[0].set_title('Train Set')
ax[1].set_title('Test Set')
plt.show()

```

the maximum value of $tpr*(1-fpr)$ 0.49 for threshold 0.39

the maximum value of $tpr*(1-fpr)$ 0.39 for threshold 0.5



```

In [201]: #Analysis on the False positives

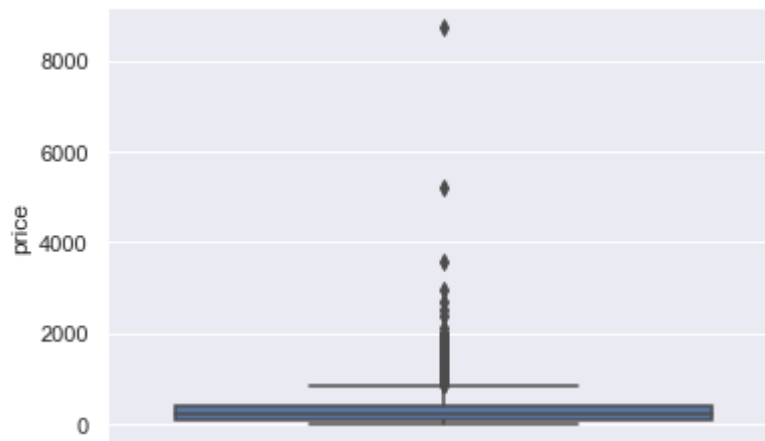
fpi = []
for i in range(len(y_test)) :
    if (y_test.values[i] == 0) & (predictions1[i] == 1) :
        fpi.append(i)
fp_essay1 = []
for i in fpi :
    fp_essay1.append(X_test['essay'].values[i])

```

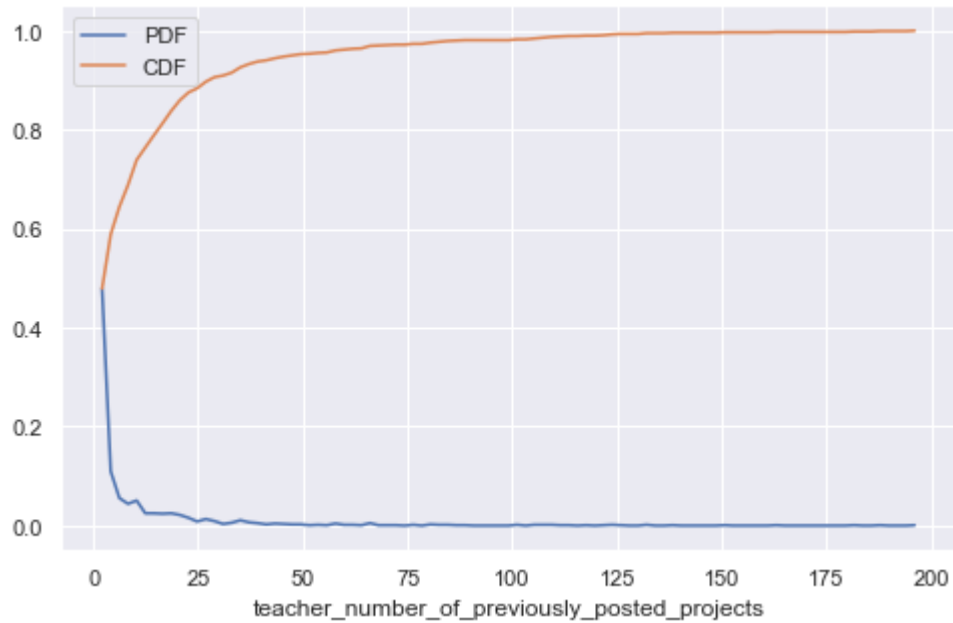


```
In [203]: #Box Plot (FP 'price')
# first get the columns:
cols = X_test.columns
X_test_falsePos1 = pd.DataFrame(columns=cols)
# get the data of the false pisitives
for i in fpi : # (in fpi all the false positives data points indexes)
    X_test_falsePos1 = X_test_falsePos1.append(X_test.filter(items=[i], axis=0))
sns.boxplot(y='price', data=X_test_falsePos1)
```

Out[203]: <matplotlib.axes._subplots.AxesSubplot at 0x16bc088ad68>



```
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_
y_posted_projects'],bins='auto', density=True)
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()
```



Conclusions

```
In [210]: # Please compare all your models using Prettytable Library
#how to use pretty table http://zetcode.com/python/prettytable/
from prettytable import PrettyTable
tb = PrettyTable()
tb.field_names= ( " Vectorizer ", " Max_depth ", " Min_sample_split ", " Test -AUC ")
tb.add_row([ " BOW ", 10, 500, 67])
tb.add_row([ " Tf - Idf", 10 , 500 ,66.5 ])
tb.add_row([ " AVG-W2V", 5, 500,63.8 ])
tb.add_row([ "A VG - Tf - Idf", 5 , 500 ,56.5])
tb.add_row([ "Top 5000 Features", 10, 500 ,66.5 ])
print(tb.get_string(titles = "Decision trees- Observations"))
```

Vectorizer	Max_depth	Min_sample_split	Test -AUC
BOW	10	500	67
Tf - Idf	10	500	66.5
AVG-W2V	5	500	63.8
A VG - Tf - Idf	5	500	56.5
Top 5000 Features	10	500	66.5