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

| • | |
|--|-----------------------------------|
| A unique identifier for the proposed project. Example: p036502 | project_id |
| Title of the project. Examples | |
| Art Will Make You Happy! First Grade Fun | project_title |
| Grade level of students for which the project is targeted. One of the following enumerated values: | |
| • Grades PreK-2 • Grades 3-5 | <pre>project_grade_category</pre> |
| • Grades 6-8 | |
| • Grades 9-12 | |

Description

Feature

| | _ coonplicat |
|--|--|
| | One or more (comma-separated) subject categories for the project from the following enumerated list of values: |
| | Applied Learning Care & Hunger |
| | Health & Sports |
| | History & Civics History & Language |
| | Literacy & LanguageMath & Science |
| <pre>project_subject_categories</pre> | • Music & The Arts |
| | Special Needs Warmth |
| | • Wariii.Cii |
| | Examples: |
| | • Music & The Arts |
| | Literacy & Language, Math & Science |
| school_state | State where school is located (Two-letter U.S. postal code |
| _ | (https://en.wikipedia.org/wiki/List_of_U.Sstate_abbreviations#Postal_codes)). Example: WY |
| | One or more (comma-separated) subject subcategories for the project. Examples: |
| <pre>project_subject_subcategories</pre> | • Literacy |
| | Literature & Writing, Social Sciences |
| | An explanation of the resources needed for the project. Example: |
| <pre>project_resource_summary</pre> | • My students need hands on literacy materials to manage sensory needs! |
| project_essay_1 | First application essay* |
| project_essay_2 | Second application essay* |
| project_essay_3 | Third application essay* |
| project_essay_4 | Fourth application essay* |
| <pre>project_submitted_datetime</pre> | Datetime when project application was submitted. Example: 2016-04-28 12:43:56.245 |
| | A unique identifier for the teacher of the proposed project. Example: |
| teacher_id | bdf8baa8fedef6bfeec7ae4ff1c15c56 |
| | |

Description

| Feature | Description |
|---------|-------------|
|---------|-------------|

Teacher's title. One of the following enumerated values:

teacher_prefix •
•
•
•
•
•
•

nan

Dr.

Mr. Mrs.

Ms.

Teacher.

Number of project applications previously submitted by the same teacher. **Example:** 2

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:

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

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

Project_is_approved

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

Notes on the Essay Data

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

- project essay 1: "Introduce us to your classroom"
- __project_essay_2:__ "Tell us more about your students"
- __project_essay_3:__ "Describe how your students will use the materials you're requesting"

teacher_number_of_previously_posted_projects

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

- __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.

```
%matplotlib inline
In [1]:
        import warnings
        warnings.filterwarnings("ignore")
        import dill
        import salite3
        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
        from wordcloud import WordCloud
```

```
g chunkize to chunkize_serial
  warnings.warn("detected Windows; aliasing chunkize to chunkize_serial")
```

1.1 Reading Data

```
In [2]: project_data = pd.read_csv('train_data.csv')
    resource_data = pd.read_csv('resources.csv')
```

Taking only 50K points as Training runs slower with many points

```
In [3]:
        import random
        project data = project data.loc[random.sample(list(project data.index), 50000)]
        print("Number of data points in train data", project data.shape)
In [4]:
        print('-'*50)
        print("The attributes of data :", project data.columns.values)
        Number of data points in train data (50000, 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']
        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 |

1.2 preprocessing of project_subject_categories

```
In [6]: catogories = list(project data['project subject categories'].values)
        # remove special characters from list of strings python: https://stackoverflow.com/a/47301924/4084039
        # https://www.geeksforgeeks.org/removing-stop-words-nltk-python/
        # https://stackoverflow.com/questions/23669024/how-to-strip-a-specific-word-from-a-string
        # https://stackoverflow.com/questions/8270092/remove-all-whitespace-in-a-string-in-python
        cat list = []
        for i in catogories:
            temp = ""
            # consider we have text like this "Math & Science, Warmth, Care & Hunger"
            for j in i.split(','): # it will split it in three parts ["Math & Science", "Warmth", "Care & Hunger"]
                if 'The' in j.split(): # this will split each of the catogory based on space "Math & Science"=> "Math",
                    j=j.replace('The','') # if we have the words "The" we are going to replace it with ''(i.e removing
                j = j.replace(' ','') # we are placeing all the ' '(space) with ''(empty) ex:"Math & Science"=>"Math&Sci
                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 [7]:
        sub catogories = list(project data['project subject subcategories'].values)
        # remove special characters from list of strings python: https://stackoverflow.com/a/47301924/4084039
        # https://www.geeksforgeeks.org/removing-stop-words-nltk-python/
        # https://stackoverflow.com/questions/23669024/how-to-strip-a-specific-word-from-a-string
        # https://stackoverflow.com/questions/8270092/remove-all-whitespace-in-a-string-in-python
        sub cat list = []
        for i in sub catogories:
            temp = ""
            # consider we have text like this "Math & Science, Warmth, Care & Hunger"
            for j in i.split(','): # it will split it in three parts ["Math & Science", "Warmth", "Care & Hunger"]
                if 'The' in j.split(): # this will split each of the catogory based on space "Math & Science"=> "Math",
                    i=i.replace('The','') # if we have the words "The" we are going to replace it with ''(i.e removing
                j = j.replace(' ','') # we are placeing all the ' '(space) with ''(empty) ex: "Math & Science" => "Math&Sci
                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]))
```

1.3 Text preprocessing

project_data.head(2) In [9]: Out[9]: Unnamed: id teacher_id teacher_prefix school_state project_submitted_datetime project_grade_ca 0 69722 Mrs. UT 79480 p045266 a3cd759de4924a1a7dbfb16cd59f444d 2016-09-01 08:15:55 Grac 38481 135951 p060970 7125cd2ad94bbdbc44c3c98eab53808b Mrs. IN 2016-07-06 22:17:29 Grac In [10]: #### 1.4.2.3 Using Pretrained Models: TFIDF weighted W2V

```
In [11]: # printing some random reviews
    print(project_data['essay'].values[0])
    print("="*50)
    print(project_data['essay'].values[150])
    print(project_data['essay'].values[1000])
    print("="*50)
    print(project_data['essay'].values[20000])
    print("="*50)
```

As a teacher in a low-income/high poverty school district, many students are faced with several challenges bot h in and out of the classroom. Despite the challenges they face, students come into the classroom eager to lea rn. I am looking to provide my students with creative and meaningful learning experiences through the power of choice, self determination, inquiry, and teamwork.\r\n\r\nMy students are hardworking, positive, and eager lea rners. They like to move; they love to read, explore, and love lots of positive attention. Many student famili es face hardships. Most families receive a free lunch based on their socioeconomic status. These things may pr event them from getting ahead, but no matter their challenges, students are eager to learn.\r\n\r\nFrom the mi nute students walk in the door of our classroom, I focus on their potential and growth while they are with me. By doing this in a creative and positive way, I am hopeful to inspire students to understand themselves as lea rners strive to be life long learners and positive community members. How can simple machines change the lives of those who are disabled? Help the students in my class and the rest of the third grade become inventors thro ugh self exploration. Students will use models of simple machines and books to first understand what a simple machine is. Afterwards, they will use this information to then find a way to use simple machines to help those who are disabled. \r\n\r\nWith your support students will have materials they need to explore without the limi t of materials such as rope, pulleys, carabiners, pails, and chipboard.\r\n\r\nThere are three third grade cla ssrooms who will use the materials for this project. My goal is for students to not only understand how hard i t is for some people to live, but for students to find a new appreciation for inventing with simple machines t o help those who are coming home from war, born disabled or had a crippling accident that left them in a wheel chair.nannan

My students get 20 minutes a day of recess. Some days it's raining and they can't get outside, or its winter a nd we have no access to exercise for the day. My students want to get up and move. I sat down with my kids and asked what kind of indoor recess toys they would like. They all want to do similar things they can do when we have outdoor recess. They want to move, move, move! They suggested games with balls or items they can move wit h. The majority of my students don't want to color on a rainy day. \r\n\r\n86% of our students receive free or reduced lunch. Our students come from a very diverse family life. Our students are part of a Title 1 school. We are in our third year of an extended school day, which means they are in school for an extra 100 minutes a day. \r\n\r\nMy students want to have fun during indoor recess instead of your typical board games or colorin g. My students are requesting items that they can get up and move during recess. We sat down and talked about different types of indoor recess activities that can get them up and moving. They used their Chromebooks to find me items they would like to play with inside on a rainy day or in the winter.\r\nThe basketball hoop, foam balls, pogo sticks, hopscotch mat, EZ steppers, and spooner boards will keep them active and moving like they are outside. The toss games and other board games will keep them having fun while getting healthy. \r\nMy stud

ents are in school for an extra 100 minutes a day compared to other schools. I want them up and moving so they have the energy to continue with their long day. When I asked them what they want to do besides learn their re sponse was \"We want to have fun in school\", \" My favorite part of school is recess.\". This inspired me to find activities that they can excel in school and get the exercise they need to stay healthy.nannan

My Kindergarten students are diverse, lively learners who are always interested in 'what happens next?'. They are young and impressionable with big hearts and growing minds. \r\n\r\nThey want to know more and love learning new ways to get there. \r\n\r\nThey are discoverers of new information and enjoy sharing their learning with others. My school is a low income, Title 1 school with a very high non English speaking population. Most fam ilies are struggling to provide the basics at home and a book is rarely seen. My students enter with limited e xposure to literacy so they do not yet know the joys that it brings. Our school does not have the proper funds to replace/repair material at our school. My students are going without basic learning materials, such as a functioning listening center. These CD/Cassette Players will give my students the necessary tools to hear stories, as they are read out loud, with expression and excitement, without losing the information because they are struggling to read the words themselves.\r\n\r\nReaders are born in the laps of their laps of their parents. S ince, most of my students lack that experience, the materials in this project will help in filling that gap. B eing able to enrich the students' learning significantly while closing their gap on the lack of spoken words w ill significantly boost their chances at the academic success that I know they are capable of!nannan

Our high school is located in a 100 + year old elementary school, but the students never let this deter them. They come to class every day prepared to work hard for what they want. These students are some of the most har d-working, determined people I have ever had the pleasure of knowing.\r\nMy student are used to working to ove rcome adversity in their lives; I would love to help make one aspect of their lives less of a struggle. I woul d like to help them do so by providing them with the resources they need. We may have not all of the resource s that we would like, but we work with what we have. In today's world it is so important that my students lear n how to use technology they do not have at home, and that is what I want to help with.\t\r\nCurrently when we have a presentation, project, or paper assigned, the students have to take turns trying to do this on the eigh t chromebooks we have). We also try to borrow a few spare Chromebooks from other teachers if they are free for the day. My students need new Chromebooks for our classroom.\r\nThey will be thrilled by the opportunity to re search their projects and papers on new computers rather than trying to share the 8 chromebooks we currently h ave. That amount does not even cover a third of my students in each class; the rest of the students must use t heir personal phones to research and write, which is a huge hassle.\r\nThe students will be able to research i nformation and learn how to be more technologically literate at the same time.\r\nProviding more computers for the students to access the wealth of knowledge that surrounds them will not only better enable them to learn f acts, it will also give me opportunities to teach digital literacy and citizenship. These computers will also give a morale boost to students that are not used to having nice, new things provided to them.\r\nnannan

```
In [12]: | # 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 [13]: sent = decontracted(project_data['essay'].values[20000])
    print(sent)
    print("="*50)
```

Our high school is located in a 100 + year old elementary school, but the students never let this deter them. They come to class every day prepared to work hard for what they want. These students are some of the most har d-working, determined people I have ever had the pleasure of knowing.\r\nMy student are used to working to ove rcome adversity in their lives; I would love to help make one aspect of their lives less of a struggle. I woul d like to help them do so by providing them with the resources they need. We may have not all of the resource s that we would like, but we work with what we have. In today is world it is so important that my students lea rn how to use technology they do not have at home, and that is what I want to help with.\t\r\nCurrently when w e have a presentation, project, or paper assigned, the students have to take turns trying to do this on the ei ght chromebooks we have). We also try to borrow a few spare Chromebooks from other teachers if they are free f or the day. My students need new Chromebooks for our classroom.\r\nThey will be thrilled by the opportunity to research their projects and papers on new computers rather than trying to share the 8 chromebooks we currently have. That amount does not even cover a third of my students in each class; the rest of the students must use their personal phones to research and write, which is a huge hassle.\r\nThe students will be able to research information and learn how to be more technologically literate at the same time.\r\nProviding more computers fo r the students to access the wealth of knowledge that surrounds them will not only better enable them to learn facts, it will also give me opportunities to teach digital literacy and citizenship. These computers will also give a morale boost to students that are not used to having nice, new things provided to them.\r\nnannan ______

```
In [14]: # \r \n \t remove from string python: http://texthandler.com/info/remove-line-breaks-python/
    sent = sent.replace('\\r', ' ')
    sent = sent.replace('\\"', ' ')
    print(sent)
```

Our high school is located in a 100 + year old elementary school, but the students never let this deter them. They come to class every day prepared to work hard for what they want. These students are some of the most har d-working, determined people I have ever had the pleasure of knowing. My student are used to working to overc ome adversity in their lives; I would love to help make one aspect of their lives less of a struggle. I would like to help them do so by providing them with the resources they need. We may have not all of the resources that we would like, but we work with what we have. In today is world it is so important that my students learn how to use technology they do not have at home, and that is what I want to help with.\t Currently when we hav e a presentation, project, or paper assigned, the students have to take turns trying to do this on the eight c hromebooks we have). We also try to borrow a few spare Chromebooks from other teachers if they are free for th e day. My students need new Chromebooks for our classroom. They will be thrilled by the opportunity to resear ch their projects and papers on new computers rather than trying to share the 8 chromebooks we currently have. That amount does not even cover a third of my students in each class; the rest of the students must use their personal phones to research and write, which is a huge hassle. The students will be able to research informat ion and learn how to be more technologically literate at the same time. Providing more computers for the stud ents to access the wealth of knowledge that surrounds them will not only better enable them to learn facts, it will also give me opportunities to teach digital literacy and citizenship. These computers will also give a mo rale boost to students that are not used to having nice, new things provided to them. nannan

```
In [15]: #remove spacial character: https://stackoverflow.com/a/5843547/4084039
sent = re.sub('[^A-Za-z0-9]+', ' ', sent)
print(sent)
```

Our high school is located in a 100 year old elementary school but the students never let this deter them They come to class every day prepared to work hard for what they want These students are some of the most hard work ing determined people I have ever had the pleasure of knowing My student are used to working to overcome adver sity in their lives I would love to help make one aspect of their lives less of a struggle I would like to hel p them do so by providing them with the resources they need We may have not all of the resources that we would like but we work with what we have In today is world it is so important that my students learn how to use tech nology they do not have at home and that is what I want to help with t Currently when we have a presentation p roject or paper assigned the students have to take turns trying to do this on the eight chromebooks we have We also try to borrow a few spare Chromebooks from other teachers if they are free for the day My students need n ew Chromebooks for our classroom They will be thrilled by the opportunity to research their projects and paper s on new computers rather than trying to share the 8 chromebooks we currently have That amount does not even c over a third of my students in each class the rest of the students must use their personal phones to research and write which is a huge hassle The students will be able to research information and learn how to be more te chnologically literate at the same time Providing more computers for the students to access the wealth of know ledge that surrounds them will not only better enable them to learn facts it will also give me opportunities t o teach digital literacy and citizenship These computers will also give a morale boost to students that are no t used to having nice new things provided to them nannan

```
In [16]:
         # https://gist.github.com/sebleier/554280
         # we are removing the words from the stop words list: 'no', 'nor', 'not'
         stopwords= ['i', 'me', 'my', 'myself', 'we', 'our', 'ours', 'ourselves', 'you', "you're", "you've",\
                      "you'll", "you'd", 'your', 'yours', 'yourself', 'yourselves', 'he', 'him', 'his', 'himself', \setminus
                      'she', "she's", 'her', 'hers', 'herself', 'it', "it's", 'its', 'itself', 'they', 'them', 'their',\
                      'theirs', 'themselves', 'what', 'which', 'who', 'whom', 'this', 'that', "that'll", 'these', 'those',
                      'am', 'is', 'are', 'was', 'were', 'be', 'been', 'being', 'have', 'has', 'had', 'having', 'do', 'does
                      'did', 'doing', 'a', 'an', 'the', 'and', 'but', 'if', 'or', 'because', 'as', 'until', 'while', 'of'
                      'at', 'by', 'for', 'with', 'about', 'against', 'between', 'into', 'through', 'during', 'before', 'ad
                      'above', 'below', 'to', 'from', 'up', 'down', 'in', 'out', 'on', 'off', 'over', 'under', 'again', '-
                      'then', 'once', 'here', 'there', 'when', 'where', 'why', 'how', 'all', 'any', 'both', 'each', 'few'
                      'most', 'other', 'some', 'such', 'only', 'own', 'same', 'so', 'than', 'too', 'very', \
                      's', 't', 'can', 'will', 'just', 'don', "don't", 'should', "should've", 'now', 'd', 'll', 'm', 'o',
                      've', 'y', 'ain', 'aren', "aren't", 'couldn', "couldn't", 'didn', "didn't", 'doesn', "doesn't", 'had
                      "hadn't", 'hasn', "hasn't", 'haven', "haven't", 'isn', "isn't", 'ma', 'mightn', "mightn't", 'mustn'
                      "mustn't", 'needn', "needn't", 'shan', "shan't", 'shouldn', "shouldn't", 'wasn', "wasn't", 'weren',
                      'won', "won't", 'wouldn', "wouldn't", "nannan", "nan"]
```

```
In [17]: # Combining all the above stundents
from tqdm import tqdm
preprocessed_essays = []
# tqdm is for printing the status bar
for sentance in tqdm(project_data['essay'].values):
    sent = decontracted(sentance)
    sent = sent.replace('\\r', ' ')
    sent = sent.replace('\\", ' ')
    sent = sent.replace('\\", ' ')
    sent = re.sub('[^A-Za-z0-9]+', ' ', sent)
# https://gist.github.com/sebleier/554280
sent = ' '.join(e for e in sent.split() if e not in stopwords)
    preprocessed_essays.append(sent.lower().strip())
```

100%| 50000/50000 [00:57<00:00, 87]

```
In [18]: # after preprocesing
preprocessed_essays[20000]
```

Out[18]: 'our high school located 100 year old elementary school students never let deter they come class every day pre pared work hard want these students hard working determined people i ever pleasure knowing my student used wor king overcome adversity lives i would love help make one aspect lives less struggle i would like help providin g resources need we may not resources would like work in today world important students learn use technology n ot home i want help currently presentation project paper assigned students take turns trying eight chromebooks we also try borrow spare chromebooks teachers free day my students need new chromebooks classroom they thrille d opportunity research projects papers new computers rather trying share 8 chromebooks currently that amount n ot even cover third students class rest students must use personal phones research write huge hassle the stude nts able research information learn technologically literate time providing computers students access wealth k nowledge surrounds not better enable learn facts also give opportunities teach digital literacy citizenship th ese computers also give morale boost students not used nice new things provided'

1.4 Preprocessing of `project_title`

Following Code blocks provided by me.

```
In [19]: # Code took from original code provided.
         # Also function used from original code.
         preprocessed titles = []
         for sent in tqdm(project data['project title'].values):
             sent = decontracted(sent)
             sent = sent.replace('\\r', ' ')
             sent = sent.replace('\\"', ' ')
             sent = sent.replace('\\n', ' ')
             sent = re.sub('[^A-Za-z0-9]+', '', sent)
             sent = ' '.join(e.lower() for e in sent.split() if e.lower() not in stopwords)
             preprocessed titles.append(sent.lower().strip())
         100%
                                                                                         50000/50000 [00:02<00:00, 2078
         8.14it/s]
In [20]: preprocessed titles[20000]
Out[20]: 'research ready chromebooks'
```

Following Code blocks present in original notebook.

1.5 Preparing data for models

```
- 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 (optinal)
- quantity : numerical (optinal)
- 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 [22]: # we use count vectorizer to convert the values into one
    from sklearn.feature_extraction.text import CountVectorizer
    vectorizer = CountVectorizer(vocabulary=list(sorted_cat_dict.keys()), lowercase=False, binary=True)
    categories_one_hot = vectorizer.fit_transform(project_data['clean_categories'].values)
    print(vectorizer.get_feature_names())
    print("Shape of matrix after one hot encodig ",categories_one_hot.shape)

['Warmth', 'Care_Hunger', 'History_Civics', 'Music_Arts', 'AppliedLearning', 'SpecialNeeds', 'Health_Sports',
    'Math_Science', 'Literacy_Language']
    Shape of matrix after one hot encodig (50000, 9)
```

```
In [23]: # we use count vectorizer to convert the values into one
    vectorizer = CountVectorizer(vocabulary=list(sorted_sub_cat_dict.keys()), lowercase=False, binary=True)
    sub_categories_one_hot = vectorizer.fit_transform(project_data['clean_subcategories'].values)
    print(vectorizer.get_feature_names())
    print("Shape of matrix after one hot encodig ",sub_categories_one_hot.shape)

['Economics', 'CommunityService', 'FinancialLiteracy', 'ParentInvolvement', 'Extracurricular', 'Civics_Governm
    ent', 'ForeignLanguages', 'Warmth', 'Care_Hunger', 'NutritionEducation', 'SocialSciences', 'PerformingArts',
    'CharacterEducation', 'TeamSports', 'Other', 'College_CareerPrep', 'History_Geography', 'Music', 'Health_LifeS
    cience', 'EarlyDevelopment', 'ESL', 'Gym_Fitness', 'EnvironmentalScience', 'VisualArts', 'Health_Wellness', 'A
    ppliedSciences', 'SpecialNeeds', 'Literature_Writing', 'Mathematics', 'Literacy']
    Shape of matrix after one hot encodig (50000, 30)
In [24]: # you can do the similar thing with state, teacher prefix and project grade category also
```

In [24]: # you can do the stmttar thing with state, teacher_prefix and project_grade_category atso

Following Code blocks provided by me.

```
In [25]: # Code took from original code provided.
    states = project_data['school_state'].unique()
    vectorizer = CountVectorizer(vocabulary=list(states), lowercase=False, binary=True)
    vectorizer.fit(project_data['school_state'].values)
    print(vectorizer.get_feature_names())

    school_state_one_hot = vectorizer.transform(project_data['school_state'].values)
    print("Shape of matrix after one hot encoding", school_state_one_hot.shape)
```

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

Shape of matrix after one hot encoding (50000, 51)

There are some NaN's in teacher_prefix column. replacing them with 'Mrs.' as that has high occurance in that column.

```
print("Number of NaN's before replacement in column: ", sum(project data['teacher prefix'].isna()))
In [26]:
         project data['teacher prefix'] = project data['teacher prefix'].replace(np.nan, 'Mrs.', regex=True)
         print("Number of NaN's after replacement in column: ", sum(project data['teacher prefix'].isna()))
         # Output may show both zeros as I re-run this several times. But there are 3 zeros in original column.
         Number of NaN's before replacement in column: 1
         Number of NaN's after replacement in column: 0
In [27]: # Code took from original code provided.
         prefixes = project data['teacher prefix'].unique()
         vectorizer = CountVectorizer(vocabulary=list(prefixes), lowercase=False, binary=True)
         vectorizer.fit(project data['teacher prefix'].values)
         print(vectorizer.get_feature_names())
         teacher prefix one hot = vectorizer.transform(project data['teacher prefix'].values)
          print("Shape of matrix after one hot encoding", teacher prefix one hot.shape)
         ['Mrs.', 'Mr.', 'Ms.', 'Teacher', 'Dr.']
         Shape of matrix after one hot encoding (50000, 5)
In [28]:
         grades = project data['project grade category'].unique()
         vectorizer = CountVectorizer(vocabulary=list(grades), lowercase=False, binary=True)
         vectorizer.fit(project data['project grade category'].values)
         print(vectorizer.get feature names())
         project_grade_category_one_hot = vectorizer.transform(project_data['project_grade_category'].values)
         print("Shape of matrix after one hot encoding", project grade category one hot.shape)
         ['Grades 3-5', 'Grades 6-8', 'Grades PreK-2', 'Grades 9-12']
         Shape of matrix after one hot encoding (50000, 4)
```

Following Code blocks present in original notebook.

1.5.2 Vectorizing Text data

1.5.2.1 Bag of words

```
In [29]: # We are considering only the words which appeared in at least 10 documents(rows or projects).
    vectorizer = CountVectorizer(min_df=10)
    text_bow = vectorizer.fit_transform(preprocessed_essays)
    print("Shape of matrix after one hot encodig ",text_bow.shape)

Shape of matrix after one hot encodig (50000, 12258)

In [30]: # you can vectorize the title also
    # before you vectorize the title make sure you preprocess it
```

Following Code blocks provided by me.

```
In [31]: # Code took from original code provided.
# We are considering only the words which appeared in at least 5 documents(rows or projects).
# Reduced number as title has less words
vectorizer = CountVectorizer(min_df=10)
titles_bow = vectorizer.fit_transform(preprocessed_titles)
print("Shape of matrix after one hot encodig ", titles_bow.shape)
```

Shape of matrix after one hot encodig (50000, 2025)

Shape of matrix after one hot encodig (50000, 12258)

Following Code blocks present in original notebook.

1.5.2.2 TFIDF vectorizer

```
In [32]: from sklearn.feature_extraction.text import TfidfVectorizer
    vectorizer = TfidfVectorizer(min_df=10)
    text_tfidf = vectorizer.fit_transform(preprocessed_essays)
    print("Shape of matrix after one hot encodig ",text_tfidf.shape)
```

1.5.2.3 Using Pretrained Models: Avg W2V

```
In [33]: # stronging variables into pickle files python: http://www.jessicayung.com/how-to-use-pickle-to-save-and-load-vd
         # make sure you have the glove vectors file
         with open('glove vectors', 'rb') as f:
             model = pickle.load(f)
             glove words = set(model.keys())
In [34]: # average Word2Vec
         # compute average word2vec for each review.
         avg w2v vectors = []; # the avg-w2v for each sentence/review is stored in this list
         for sentence in tqdm(preprocessed essays): # for each review/sentence
             vector = np.zeros(300) # as word vectors are of zero length
             cnt words =0; # num of words with a valid vector in the sentence/review
             for word in sentence.split(): # for each word in a review/sentence
                 if word in glove words:
                     vector += model[word]
                     cnt words += 1
             if cnt words != 0:
```

```
100%| 50000/50000 [00:29<00:00, 168 6.30it/s]
```

50000 300

1.5.2.3 Using Pretrained Models: TFIDF weighted W2V

vector /= cnt_words
avg w2v vectors.append(vector)

print(len(avg_w2v_vectors))
print(len(avg_w2v_vectors[0]))

```
In [35]: # S = ["abc def pqr", "def def def abc", "pqr pqr def"]
    tfidf_model = TfidfVectorizer()
    tfidf_model.fit(preprocessed_essays)
    # we are converting a dictionary with word as a key, and the idf as a value
    dictionary = dict(zip(tfidf_model.get_feature_names(), list(tfidf_model.idf_)))
    tfidf_words = set(tfidf_model.get_feature_names())
```

```
In [36]: # average Word2Vec
         # compute average word2vec for each review.
         tfidf w2v vectors = []; # the avg-w2v for each sentence/review is stored in this list
         for sentence in tqdm(preprocessed essays): # for each review/sentence
             vector = np.zeros(300) # as word vectors are of zero length
             tf_idf_weight =0; # num of words with a valid vector in the sentence/review
             for word in sentence.split(): # for each word in a review/sentence
                 if (word in glove words) and (word in tfidf words):
                     vec = model[word] # getting the vector for each word
                     # here we are multiplying idf value(dictionary[word]) and the tf value((sentence.count(word)/len(ser
                     tf idf = dictionary[word]*(sentence.count(word)/len(sentence.split())) # getting the tfidf value for
                     vector += (vec * tf idf) # calculating tfidf weighted w2v
                     tf_idf_weight += tf_idf
             if tf idf weight != 0:
                 vector /= tf idf weight
             tfidf w2v vectors.append(vector)
         print(len(tfidf w2v vectors))
         print(len(tfidf w2v vectors[0]))
         100%
                                                                                            50000/50000 [03:31<00:00, 23
         6.62it/sl
         50000
         300
         # Similarly you can vectorize for title also
In [37]:
```

Following Code blocks provided by me.

```
In [38]: # Code took from original code provided.
# tfidf of project titles
vectorizer = TfidfVectorizer(min_df=10)
titles_tfidf = vectorizer.fit_transform(preprocessed_titles)
print("Shape of matrix after one hot encodig ",titles_tfidf.shape)
Shape of matrix after one hot encodig (50000, 2025)
```

```
In [39]: # Code took from original code provided.
         # avg-w2v for project titles
         avg w2v titles = []
         for sentence in tqdm(preprocessed titles):
             vector = np.zeros(300)
             cnt words =0;
             for word in sentence.split():
                 if word in glove words:
                     vector += model[word]
                     cnt words += 1
             if cnt words != 0:
                 vector /= cnt words
             avg w2v titles.append(vector)
         print(len(avg_w2v_titles))
         print(len(avg_w2v_titles[0]))
         100%|
                                                                                          50000/50000 [00:01<00:00, 3849
         3.33it/s]
         50000
         300
In [40]: # Code took from original code provided.
         tfidf model = TfidfVectorizer()
         tfidf model.fit(preprocessed titles)
         dictionary = dict(zip(tfidf_model.get_feature_names(), list(tfidf_model.idf_)))
         tfidf words = set(tfidf model.get feature names())
```

```
In [41]: # Code took from original code provided.
         # tfidf-w2v for project titles
         tfidf w2v titles = []
         for sentence in tqdm(preprocessed titles):
             vector = np.zeros(300)
             tf idf weight =0
             for word in sentence.split():
                 if (word in glove words) and (word in tfidf words):
                     vec = model[word]
                     tf idf = dictionary[word]*(sentence.count(word)/len(sentence.split()))
                     vector += (vec * tf idf)
                     tf idf weight += tf idf
             if tf idf weight != 0:
                 vector /= tf idf weight
             tfidf w2v titles.append(vector)
         print(len(tfidf w2v titles))
         print(len(tfidf w2v titles[0]))
         100%
                                                                                          50000/50000 [00:02<00:00, 1965
         5.51it/s]
```

```
100%| 5.51it/s]
50000
300
```

Following Code blocks present in original notebook.

1.5.3 Vectorizing Numerical features

```
In [42]: 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 [43]: # check this one: https://www.youtube.com/watch?v=0HOqOcln3Z4&t=530s
         # standardization sklearn: https://scikit-learn.org/stable/modules/generated/sklearn.preprocessing.StandardScale
         from sklearn.preprocessing import StandardScaler
         # price standardized = standardScalar.fit(project data['price'].values)
         # this will rise the error
         # ValueError: Expected 2D array, got 1D array instead: array=[725.05 213.03 329. ... 399. 287.73 5.5].
         # Reshape your data either using array.reshape(-1, 1)
          price scalar = StandardScaler()
         price scalar.fit(project data['price'].values.reshape(-1,1)) # finding the mean and standard deviation of this
         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 standardized = price scalar.transform(project data['price'].values.reshape(-1, 1))
         Mean : 297.4139702, Standard deviation : 365.5914197335812
In [44]: price standardized
Out[44]: array([[ 1.12813378],
                [-0.40327525],
                [ 0.20141619],
                [-0.48530671],
                [-0.2883929],
                [-0.4786873 11)
```

Following Code blocks provided by me.

```
In [45]: | warnings.filterwarnings("ignore")
         # Code took from original code provided
          scalar = StandardScaler()
          scalar.fit(project data['teacher number of previously posted projects'].values.reshape(-1, 1))
          print(f"Mean : {scalar.mean [0]}, Standard deviation : {np.sqrt(scalar.var [0])}")
         # Now standardize the data with above maen and variance.
          previously posted projects standardized = \
                          scalar.transform(project data['teacher number of previously posted projects'].values.reshape(-1,
          print(previously posted projects standardized)
         Mean : 11.24058, Standard deviation : 27.947352312224496
         [[-0.04438989]
          [-0.40220554]
           [-0.36642398]
           [-0.18751615]
           [-0.36642398]
           [-0.36642398]]
```

Following Code blocks present in original notebook.

1.5.4 Merging all the above features

we need to merge all the numerical vectors i.e catogorical, text, numerical vectors

http://localhost:8888/notebooks/DonorsChoose%20Data%20and%20Assignments/ilmnarayana%40gmail.com 8.ipynb

Computing Sentiment Scores

```
import nltk
In [49]:
         from nltk.sentiment.vader import SentimentIntensityAnalyzer
         # import nltk
         nltk.download('vader_lexicon')
         sid = SentimentIntensityAnalyzer()
         for sentiment = 'a person is a person no matter how small dr seuss i teach the smallest students with the bigges
         for learning my students learn in many different ways using all of our senses and multiple intelligences i use a
         of techniques to help all my students succeed students in my class come from a variety of different backgrounds
         for wonderful sharing of experiences and cultures including native americans our school is a caring community of
         learners which can be seen through collaborative student project based learning in and out of the classroom kind
         in my class love to work with hands on materials and have many different opportunities to practice a skill before
         mastered having the social skills to work cooperatively with friends is a crucial aspect of the kindergarten cur
         montana is the perfect place to learn about agriculture and nutrition my students love to role play in our prete
         in the early childhood classroom i have had several kids ask me can we try cooking with real food i will take th
         and create common core cooking lessons where we learn important math and writing concepts while cooking deliciou
         food for snack time my students will have a grounded appreciation for the work that went into making the food an
         of where the ingredients came from as well as how it is healthy for their bodies this project would expand our l
         nutrition and agricultural cooking recipes by having us peel our own apples to make homemade applesauce make our
         and mix up healthy plants from our classroom garden in the spring we will also create our own cookbooks to be pr
         shared with families students will gain math and literature skills as well as a life long enjoyment for healthy
         nannan'
         ss = sid.polarity scores(for sentiment)
         for k in ss:
             print('{0}: {1}, '.format(k, ss[k]), end='')
         # we can use these 4 things as features/attributes (neg. neu, pos. compound)
         # neq: 0.0, neu: 0.753, pos: 0.247, compound: 0.93
         [nltk data] Downloading package vader lexicon to
         [nltk data]
                         C:\Users\724463\AppData\Roaming\nltk data...
         [nltk data]
                       Package vader lexicon is already up-to-date!
```

Assignment 8: DT

neg: 0.01, neu: 0.745, pos: 0.245, compound: 0.9975,

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 paramter tuning (best `depth` in range [1, 5, 10, 50, 100, 500, 100], and the best `min_samples_split` in range [5, 10, 100, 500])

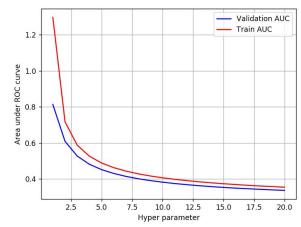
- Find the best hyper parameter which will give the maximum <u>AUC (https://www.appliedaicourse.com/course/applied-ai-course-online/lessons/receiver-operating-characteristic-curve-roc-curve-and-auc-1/) value</u>
- Find the best hyper paramter using k-fold cross validation or simple cross validation data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

3. Graphviz

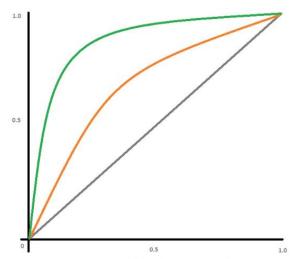
- Visualize your decision tree with Graphviz. It helps you to understand how a decision is being made, given a new vector.
- Since feature names are not obtained from word2vec related models, visualize only BOW & TFIDF decision trees using Graphviz
- · Make sure to print the words in each node of the 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



• 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 <u>confusion matrix (https://www.appliedaicourse.com/course/applied-aicourse-online/lessons/confusion-matrix-tpr-fpr-fnr-tnr-1/)</u> with predicted and original labels of test data points

| | Predicted: NO | Predicted: YES |
|-------------|------------------|-------------------|
| Actual: NO | TN = ?? | FP = ?? |
| Actual: YES | FN = ?? | TP = ?? |

- 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/)
 - 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 <u>`feature_importances_`</u> (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 you choice i.e. (Dession 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/)

| Vectorizer | Model | + Hyper parameter | AUC |
|------------|-------|------------------------|------|
| BOW | Brute | 7 | 0.78 |
| TFIDF | Brute | 12 | 0.79 |
| W2V | Brute | 10 | 0.78 |
| TFIDFW2V | Brute | 6 | 0.78 |

2. Decision Tree

Some code blocks are taken from previous assignments. And some used the code present in original file ('8_DonorsChoose_DT.ipynb') which is mentioned in comments.

Following Code blocks provided by me.

Adding a column summary_numeric_bool instead of project_resource_summary column which tells if resource summary has a number in it

```
In [50]: # ref: https://stackoverflow.com/questions/4138202/using-isdigit-for-floats
         def nums in str(text):
             Returns list of numbers present in the given string. Numbers := floats ints etc.
             result = []
             for s in text.split():
                  try:
                     x = float(s)
                     result.append(x)
                  except:
                      continue
             return result
In [51]: print(nums in str('HE44LLo 56 are -89 I 820.353 in -78.39 what .293 about 00'))
         [56.0, -89.0, 820.353, -78.39, 0.293, 0.0]
In [52]:
         numbers in summary = np.array([len(nums in str(s)) for s in project data['project resource summary']])
         project data['summary numeric bool'] = list(map(int, numbers in summary>0))
```

Taking Relevant columns as X (input data to model) and y (output class label)

```
In [53]: | project_data.columns
Out[53]: Index(['Unnamed: 0', 'id', 'teacher id', 'teacher prefix', 'school state',
                  'project submitted datetime', 'project grade category', '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',
                 'clean categories', 'clean subcategories', 'essay', 'price', 'quantity',
                 'summary numeric bool'],
                dtype='object')
In [54]:
          project data.head(2)
Out[54]:
             Unnamed:
                            id
                                                    teacher_id teacher_prefix school_state project_submitted_datetime project_grade_categor
                 79480 p045266
                               a3cd759de4924a1a7dbfb16cd59f444d
                                                                      Mrs.
                                                                                   UT
                                                                                              2016-09-01 08:15:55
                                                                                                                         Grades 3-
           1
                135951 p060970 7125cd2ad94bbdbc44c3c98eab53808b
                                                                      Mrs.
                                                                                   IN
                                                                                              2016-07-06 22:17:29
                                                                                                                         Grades 3-
          2 rows × 21 columns
         # Categorical and numerical columns are listed below.
In [55]:
          X_columns = ['teacher_prefix', 'school_state', 'project_grade_category', 'summary_numeric_bool',\
                        'teacher number of previously posted projects', 'clean categories', 'clean subcategories',\
                        'price', 'quantity']
          X = project_data[X columns]
          y = project data['project is approved']
```

Adding preprocessed_essays and preprocessed_titles as columns to X before splitting

```
In [56]: X['essay'] = preprocessed_essays
    X['project_title'] = preprocessed_titles
    X_columns.append('essay')
    X_columns.append('project_title')
    print('final columns used in input data are: ', X_columns)
```

final columns used in input data are: ['teacher_prefix', 'school_state', 'project_grade_category', 'summary_n umeric_bool', 'teacher_number_of_previously_posted_projects', 'clean_categories', 'clean_subcategories', 'pric e', 'quantity', 'essay', 'project_title']

Adding essays and calculating sentiments to Input data X before splitting as we have to use same train and test rows later for Task-2 analysis. These columns are not considered in our Task-1 analysis

2.1 Splitting data into Train and cross validation(or test): Stratified Sampling

```
In [61]: # please write all the code with proper documentation, and proper titles for each subsection
# go through documentations and blogs before you start coding
# first figure out what to do, and then think about how to do.
# reading and understanding error messages will be very much helpfull in debugging your code
# when you plot any graph make sure you use
# a. Title, that describes your plot, this will be very helpful to the reader
# b. Legends if needed
# c. X-axis label
# d. Y-axis label
```

Not creating CV data as I am using K-fold validation

```
In [62]: # Code took from SAMPLE_SOLUTION notebook
# splitting into 80-20 ratio for train-test data
from sklearn.model_selection import train_test_split
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.20, stratify=y)
```

2.2 Make Data Model Ready: encoding numerical, categorical features

```
In [64]: # please write all the code with proper documentation, and proper titles for each subsection
# go through documentations and blogs before you start coding
# first figure out what to do, and then think about how to do.
# reading and understanding error messages will be very much helpfull in debugging your code
# make sure you featurize train and test data separatly

# when you plot any graph make sure you use
# a. Title, that describes your plot, this will be very helpful to the reader
# b. Legends if needed
# c. X-axis label
# d. Y-axis label
```

numerical columns

- teacher_number_of_previously_posted_projects
- price
- quantity

Leaving summary_numeric_bool as it is because it only has 0's and 1's in it.

categorical columns

- teacher_prefix
- school_state
- project_grade_category

- clean categories
- clean_subcategories

Normalizing teacher_number_of_previously_posted_projects column

```
In [65]: warnings.filterwarnings("ignore")
# Code took from original Code provided.
scaler = StandardScaler()
scaler.fit(X_train['teacher_number_of_previously_posted_projects'].values.reshape(-1,1))
print(f"Mean : {scaler.mean_[0]}, Standard deviation : {np.sqrt(scaler.var_[0])}")

Mean : 11.238725, Standard deviation : 27.812915801375

In [66]: warnings.filterwarnings("ignore")
X_train_tnppp_norm = scaler.transform(X_train['teacher_number_of_previously_posted_projects'].values.reshape(-1, X_test_tnppp_norm = scaler.transform(X_test['teacher_number_of_previously_posted_projects'].values.reshape(-1, 1)
```

Normalizing price column

Normalizing quantity column

```
warnings.filterwarnings("ignore")
In [69]:
         # Code took from original Code provided.
         scaler = StandardScaler()
         scaler.fit(X train['quantity'].values.reshape(-1,1))
         print(f"Mean : {scaler.mean [0]}, Standard deviation : {np.sqrt(scaler.var [0])}")
         Mean: 16.849575, Standard deviation: 25.33939713804129
         warnings.filterwarnings("ignore")
In [70]:
         X train quant norm = scaler.transform(X train['quantity'].values.reshape(-1,1))
         X test quant norm = scaler.transform(X test['quantity'].values.reshape(-1,1))
         Using a array to store column names data to use at last when interpreting the model
In [71]:
         # when combining the input matrix the order of columns is same as cat num columns
         cat num columns = ['previously posted projects', 'price', 'quantity', 'summary numeric bool']
         Encoding teacher prefix column
In [72]: # Code took from SAMPLE SOLUTION notebook.
         vectorizer = CountVectorizer()
         vectorizer.fit(X train['teacher prefix'].values)
          print(vectorizer.get feature names())
         ['dr', 'mr', 'mrs', 'ms', 'teacher']
In [73]: | # Code took from SAMPLE SOLUTION notebook.
         X_train_prefix_ohe = vectorizer.transform(X_train['teacher_prefix'].values)
         X test prefix ohe = vectorizer.transform(X test['teacher prefix'].values)
         print(X train prefix ohe.shape, y train.shape)
         print(X test prefix ohe.shape, y test.shape)
          (40000, 5) (40000,)
          (10000, 5) (10000,)
```

In [74]: | cat num columns.extend(['prefix '+i for i in vectorizer.get feature names()])

Encoding school state column

```
In [75]: # Code took from SAMPLE SOLUTION notebook.
         vectorizer = CountVectorizer()
         vectorizer.fit(X train['school state'].values)
          print(vectorizer.get feature names())
         ['ak', 'al', 'ar', 'az', 'ca', 'co', 'ct', 'dc', 'de', 'fl', 'ga', 'hi', 'ia', 'id', 'il', 'in', 'ks', 'ky',
         'la', 'ma', 'md', 'me', 'mi', 'mn', 'mo', 'ms', 'mt', 'nc', 'nd', 'ne', 'nh', 'nj', 'nm', 'nv', 'ny', 'oh', 'o
         k', 'or', 'pa', 'ri', 'sc', 'sd', 'tn', 'tx', 'ut', 'va', 'vt', 'wa', 'wi', 'wv', 'wy']
In [76]: # Code took from SAMPLE SOLUTION notebook.
         X train school ohe = vectorizer.transform(X train['school state'].values)
         X test school ohe = vectorizer.transform(X test['school state'].values)
         print(X train school ohe.shape, y train.shape)
         print(X test school ohe.shape, y test.shape)
          (40000, 51) (40000,)
          (10000, 51) (10000,)
In [77]: | cat num columns.extend(['state '+i for i in vectorizer.get feature names()])
         print(len(cat num columns))
         60
```

Encoding project grade category column

```
In [78]: # Code took from original Code provided.
grades = X_train['project_grade_category'].unique()
vectorizer = CountVectorizer(vocabulary=list(grades), lowercase=False, binary=True)
vectorizer.fit(X_train['project_grade_category'].values)
print(vectorizer.get_feature_names())

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

```
In [79]: # Code took from SAMPLE_SOLUTION notebook.
    X_train_grade_ohe = vectorizer.transform(X_train['project_grade_category'].values)
    X_test_grade_ohe = vectorizer.transform(X_test['project_grade_category'].values)

print(X_train_grade_ohe.shape, y_train.shape)
print(X_test_grade_ohe.shape, y_test.shape)

(40000, 4) (40000,)
(10000, 4) (10000,)

In [80]: cat_num_columns.extend(vectorizer.get_feature_names())
print(len(cat_num_columns))
```

Encoding clean categories column

```
In [81]: # Code took from original Code provided.
    vectorizer = CountVectorizer(vocabulary=list(sorted_cat_dict.keys()), lowercase=False, binary=True)
    vectorizer.fit(X_train['clean_categories'].values)
    print(vectorizer.get_feature_names())

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

In [82]: # Code took from SAMPLE_SOLUTION notebook.
    X_train_categ_ohe = vectorizer.transform(X_train['clean_categories'].values)
    X_test_categ_ohe = vectorizer.transform(X_test['clean_categories'].values)
    print(X_train_categ_ohe.shape, y_train.shape)
    print(X_test_categ_ohe.shape, y_test.shape)

    (40000, 9) (40000,)
    (10000, 9) (10000,)

In [83]: cat_num_columns.extend(['categ_'+i for i in vectorizer.get_feature_names()])
    print(len(cat_num_columns))
```

73

Encoding clean_subcategories column

```
# Code took from original Code provided.
In [84]:
         vectorizer = CountVectorizer(vocabulary=list(sorted sub cat dict.keys()), lowercase=False, binary=True)
         vectorizer.fit(X train['clean subcategories'].values)
          print(vectorizer.get feature names())
         ['Economics', 'CommunityService', 'FinancialLiteracy', 'ParentInvolvement', 'Extracurricular', 'Civics Governm
         ent', 'ForeignLanguages', 'Warmth', 'Care_Hunger', 'NutritionEducation', 'SocialSciences', 'PerformingArts',
         'CharacterEducation', 'TeamSports', 'Other', 'College CareerPrep', 'History Geography', 'Music', 'Health LifeS
         cience', 'EarlyDevelopment', 'ESL', 'Gym Fitness', 'EnvironmentalScience', 'VisualArts', 'Health Wellness', 'A
         ppliedSciences', 'SpecialNeeds', 'Literature Writing', 'Mathematics', 'Literacy']
In [85]: # Code took from SAMPLE SOLUTION notebook.
         X train subcat ohe = vectorizer.transform(X train['clean subcategories'].values)
         X test subcat ohe = vectorizer.transform(X test['clean subcategories'].values)
         print(X train subcat ohe.shape, y train.shape)
         print(X test subcat ohe.shape, y test.shape)
          (40000, 30) (40000,)
         (10000, 30) (10000,)
In [86]:
         cat num columns.extend(['subcateg '+i for i in vectorizer.get feature names()])
         print(len(cat num columns))
         103
```

Combining categorical and numerical data for further use.

2.3 Make Data Model Ready: encoding eassay, and project_title

```
In [90]: # please write all the code with proper documentation, and proper titles for each subsection
# go through documentations and blogs before you start coding
# first figure out what to do, and then think about how to do.
# reading and understanding error messages will be very much helpfull in debugging your code
# make sure you featurize train and test data separatly

# when you plot any graph make sure you use
# a. Title, that describes your plot, this will be very helpful to the reader
# b. Legends if needed
# c. X-axis label
# d. Y-axis label
```

Converting essay column to vector using Bag of Words (BoW).

```
In [91]: # Code took from original Code provided.
    vectorizer = CountVectorizer(ngram_range=(1,2), min_df=10, max_features=5000)
    vectorizer.fit(X_train['essay'].values)
    print(len(vectorizer.get_feature_names()))
```

5000

```
In [92]: # Code took from SAMPLE SOLUTION notebook.
         X train essay bow = vectorizer.transform(X train['essay'].values)
         X test essay bow = vectorizer.transform(X test['essay'].values)
          print(X train essay bow.shape, y train.shape)
         print(X test essay bow.shape, y test.shape)
          (40000, 5000) (40000,)
          (10000, 5000) (10000,)
         essay bow columns = ['essay '+i for i in vectorizer.get feature names()]
In [93]:
         print(len(essay bow columns))
         5000
In [94]:
         import random
         random.sample(essay bow columns, 10)
Out[94]: ['essay prep',
           'essay students enthusiastic',
           'essay many books',
           'essay immediately',
           'essay neighborhood school',
           'essay students receiving',
           'essay_reasoning',
           'essay receive',
           'essay view',
           'essay education possible']
```

Converting essay column to vector using TFIDF Vectorizer.

```
In [95]: # Code took from original Code provided.
    vectorizer = TfidfVectorizer(ngram_range=(1,2), min_df=10, max_features=5000)
    vectorizer.fit(X_train['essay'].values)
    print(len(vectorizer.get_feature_names()))
```

5000

```
In [96]: # Code took from SAMPLE_SOLUTION notebook.
    X_train_essay_tfidf = vectorizer.transform(X_train['essay'].values)
    X_test_essay_tfidf = vectorizer.transform(X_test['essay'].values)

    print(X_train_essay_tfidf.shape, y_train.shape)
    print(X_test_essay_tfidf.shape, y_test.shape)

    (40000, 5000) (40000,)
    (10000, 5000) (10000,)

In [97]: essay_tfidf_columns = ['essay_'+i for i in vectorizer.get_feature_names()]
    print(len(essay_tfidf_columns))

5000
```

Converting essay column to vector using Average Word2Vec.

Creating function to return average word2vec vectors given sentences

```
In [98]:
          # Code took from original Code provided.
         def avg w2v(arr):
             Returns array of vectors given array of sentences. Array of vectors are created by Average Word2Vec method a
             words is taken from 'glove vectors' file.
             avg w2v vectors = []
             for sentence in tqdm(arr):
                 vector = np.zeros(300)
                  cnt words = 0
                  for word in sentence.split():
                     if word in glove words:
                          vector += model[word]
                          cnt words += 1
                 if cnt words != 0:
                     vector /= cnt words
                 avg w2v vectors.append(vector)
             return avg w2v vectors
```

Converting essay column to vector using TFIDF weighted Word2Vec.

Creating function to return tfidf weighted word2vec vectors given sentences and idf dictionary for words

```
In [100]: # Code took from original Code provided.
          def tfidf w2v(arr, idf dict):
              Returns array of vectors given array of sentences and dictionary containing IDF values for words.
              Array of vectors are created by TFIDF weighted Word2Vec method and vectors for words is taken from 'glove ve
               0.00
              tfidf w2v vectors = []
              for sentence in tqdm(arr):
                  vector = np.zeros(300)
                  tf idf weight = 0;
                  for word in sentence.split():
                      if (word in glove words) and (word in idf dict):
                           vec = model[word]
                          tf idf = idf dict[word]/len(sentence.split())
                           vector += (vec * tf idf)
                          tf idf weight += tf idf
                  if tf idf weight != 0:
                      vector /= tf idf weight
                  tfidf w2v vectors.append(vector)
              return tfidf w2v vectors
```

Getting idf values for the words in X_train.essay data

```
In [101]: # Code took from original Code provided.
          tfidf model = TfidfVectorizer()
          tfidf model.fit(X train['essay'])
          # we are converting a dictionary with word as a key, and the idf as a value
          dictionary = dict(zip(tfidf model.get feature names(), list(tfidf model.idf )))
In [102]: | X train essay tfidfw2v = np.array(tfidf w2v(X train['essay'].values, dictionary))
          X test essay tfidfw2v = np.array(tfidf w2v(X test['essay'].values, dictionary))
          print(X train essay tfidfw2v.shape, y train.shape)
          print(X test essay tfidfw2v.shape, y test.shape)
          100%
                                                                                            40000/40000 [02:36<00:00, 25
          5.15it/s]
          100%|
                                                                                            10000/10000 [00:35<00:00, 27
          7.89it/sl
          (40000, 300) (40000,)
          (10000, 300) (10000,)
          Converting project title column to vector using Bag of Words (BoW).
In [103]:
          # Code took from original Code provided.
          vectorizer = CountVectorizer(ngram range=(1,2), min df=10, max features=5000)
          vectorizer.fit(X train['project title'].values)
          print(len(vectorizer.get feature names()))
          2723
In [104]:
          # Code took from SAMPLE SOLUTION notebook.
          X train title bow = vectorizer.transform(X train['project title'].values)
          X test title bow = vectorizer.transform(X test['project title'].values)
          print(X train title bow.shape, y train.shape)
          print(X test title bow.shape, y test.shape)
           (40000, 2723) (40000,)
          (10000, 2723) (10000,)
```

```
In [105]: title_bow_columns = ['title_'+i for i in vectorizer.get_feature_names()]
    print(len(title_bow_columns))
2723
```

Converting project title column to vector using TFIDF Vectorizer.

```
In [106]: # Code took from original Code provided.
    vectorizer = TfidfVectorizer(ngram_range=(1,2), min_df=10, max_features=5000)
    vectorizer.fit(X_train['project_title'].values)
    print(len(vectorizer.get_feature_names()))

2723

In [107]: # Code took from SAMPLE_SOLUTION notebook.
    X_train_title_tfidf = vectorizer.transform(X_train['project_title'].values)
    X_test_title_tfidf = vectorizer.transform(X_test['project_title'].values)
    print(X_train_title_tfidf.shape, y_train.shape)
    print(X_test_title_tfidf.shape, y_test.shape)

    (40000, 2723) (40000,)
    (10000, 2723) (10000,)

In [108]: title_tfidf_columns = ['title_'+i for i in vectorizer.get_feature_names()]
    print(len(title_tfidf_columns))
```

Converting project_title column to vector using Average Word2Vec.

Can use avg_w2v function

2723

Converting project_title column to vector using TFIDF weighted Word2Vec.

Can use tfidf_w2v function but should calculate idf dictionary before using it

```
In [110]: # Code took from original Code provided.
          tfidf model = TfidfVectorizer()
          tfidf model.fit(X train['project title'])
          # we are converting a dictionary with word as a key, and the idf as a value
          dictionary = dict(zip(tfidf model.get feature names(), list(tfidf model.idf )))
In [111]: X train title tfidfw2v = np.array(tfidf w2v(X train['project title'].values, dictionary))
          X test title tfidfw2v = np.array(tfidf w2v(X test['project title'].values, dictionary))
          print(X train title tfidfw2v.shape, y train.shape)
          print(X test title tfidfw2v.shape, y test.shape)
          100%
                                                                                           40000/40000 [00:01<00:00, 2114
          0.04it/sl
          100%
                                                                                           10000/10000 [00:00<00:00, 2302
          5.03it/sl
           (40000, 300) (40000,)
           (10000, 300) (10000,)
```

2.4 Appling Decision Tree on different kind of featurization as mentioned in the

instructions

Apply Decision Tree on different kind of featurization as mentioned in the instructions For Every model that you work on make sure you do the step 2 and step 3 of instrucations

```
In [112]: # please write all the code with proper documentation, and proper titles for each subsection # go through documentations and blogs before you start coding # first figure out what to do, and then think about how to do. # reading and understanding error messages will be very much helpfull in debugging your code # when you plot any graph make sure you use # a. Title, that describes your plot, this will be very helpful to the reader # b. Legends if needed # c. X-axis label # d. Y-axis label
```

```
bow train = hstack((cat num train, X train essay bow, X train title bow)).tocsr()
In [113]:
          bow test = hstack((cat num test, X test essay bow, X test title bow)).tocsr()
          tfidf train = hstack((cat num train, X train essay tfidf, X train title tfidf)).tocsr()
          tfidf test = hstack((cat num test, X test essay tfidf, X test title tfidf)).tocsr()
          avgw2v train = np.hstack((cat num train.toarray(), X train essay avgw2v, X train title avgw2v))
          avgw2v test = np.hstack((cat num test.toarray(), X test essay avgw2v, X test title avgw2v))
          tfidfw2v train = np.hstack((cat num train.toarray(), X train essay tfidfw2v, X train title tfidfw2v))
          tfidfw2v test = np.hstack((cat num test.toarray(), X test essay tfidfw2v, X test title tfidfw2v))
          print('='*30)
          print(bow train.shape)
          print(bow test.shape)
          print('='*30)
          print(tfidf train.shape)
          print(tfidf test.shape)
          print('='*30)
          print(avgw2v train.shape)
          print(avgw2v test.shape)
          print('='*30)
          print(tfidfw2v train.shape)
          print(tfidfw2v_test.shape)
           print('='*30)
```

```
In [114]: bow_columns = cat_num_columns + essay_bow_columns + title_bow_columns
    tfidf_columns = cat_num_columns + essay_tfidf_columns + title_tfidf_columns

print(len(bow_columns))
    print(len(tfidf_columns))
```

7826 7826

Writing several functions to reuse them later

Function to plot AUC values with respect to hyper-parameter C given train data using K-fold validation

```
In [115]: | from sklearn.tree import DecisionTreeClassifier
          from sklearn.metrics import roc auc score
          from sklearn.model selection import GridSearchCV
          import math
          # Code inside function took from SAMPLE SOLUTION notebook
          def auc vs K plot(X train, y train, depth s, mss s):
              Plots the AUC results for different max depths and min samples split values on train and CV data
              Parameters:
              X train, v train - data which is used for K-fold validation and used to train DecisionTreeClassifier model
               depth s - list of max depth values on which we have to train the data and plot the results
              mss s - list of min samples split values on which we have to train the data and plot the results
              dt model = DecisionTreeClassifier()
              parameters = {'max depth': depth s, 'min samples split': mss s}
              clf = GridSearchCV(dt model, parameters, cv=3, scoring='roc auc', return train score=True)
              clf.fit(X train, v train)
              train auc= clf.cv results ['mean train score']
              cv auc = clf.cv results ['mean test score']
              train auc = train auc.reshape((len(depth s), len(mss s)))
              cv auc = cv auc.reshape((len(depth s), len(mss s)))
               sns.heatmap(train auc, vmin=0, vmax=1, annot=True, xticklabels=mss s, yticklabels=depth s, fmt=".3f", cmap=
               plt.xlabel("min sample split")
               plt.vlabel("max depth")
               plt.title("Train score")
               plt.show()
               sns.heatmap(cv auc, vmin=0, vmax=1, annot=True, xticklabels=mss s, yticklabels=depth s, fmt=".3f", cmap=sns.
               plt.xlabel("min sample split")
               plt.vlabel("max depth")
              plt.title("CV score")
               plt.show()
```

Function to plots ROC curves and confusion matrices for train and test data. Function returns AUC Values for train, test data

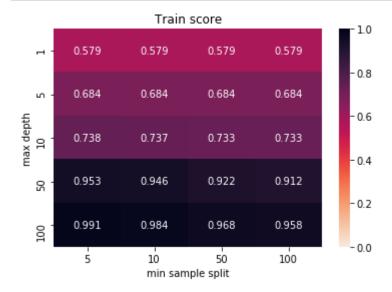
```
In [116]: | from sklearn.metrics import roc curve, auc, precision recall curve
          from IPython.display import Markdown, display
           # Code inside function took from SAMPLE SOLUTION notebook
          def ROC conf mat(X train, y train, X test, y test, best depth, best mss, plots = True):
              Plots ROC Curve given best hyper parameter values, Train data and Test data using DecisionTreeClassifier.
              And also plots confusion matrix for train data and test data taking a optimal threshold from ROC curve.
              Returns Area Under ROC Curve for Train, Test data which can be taken as performance of the model. And also references
              # Plotting ROC Curve code
              dt model = DecisionTreeClassifier(max depth = best depth, min samples split = best mss)
              dt model.fit(X train, y train)
              y train pred = dt model.predict proba(X train)[:, 1]
              y test pred = dt model.predict proba(X test)[:, 1]
              train fpr, train tpr, tr thresholds = roc curve(y train, y train pred)
              test fpr, test tpr, te thresholds = roc curve(y test, y test pred)
              result = {}
              result['train auc'], result['test auc'] = (auc(train fpr, train tpr), auc(test fpr, test tpr))
              result['model'] = dt model
              thr train = tr thresholds[np.argmax(train tpr*(1-train fpr))]
              thr test = te thresholds[np.argmax(test tpr*(1-test fpr))]
              train predictions = []
              for i in y train pred:
                   if i >= thr train:
                      train predictions.append(1)
                   else:
                       train predictions.append(0)
              test predictions = []
              for i in y test pred:
                   if i >= thr test:
                      test predictions.append(1)
                   else:
                       test predictions.append(0)
```

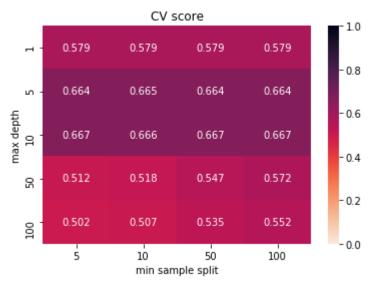
```
# Collecting False Positive indices from the test data.
result['false positive'] = [i for i in range(len(y test)) if test predictions[i]==1 and y test.iloc[i]==0]
if(plots):
    display(Markdown(f"**Analysis for max depth = {best depth} and min samples split = {best mss}**"))
    plt.plot(train fpr, train tpr, label="train AUC ="+str(np.round(result['train auc'], 3)))
    plt.plot(test fpr, test tpr, label="test AUC ="+str(np.round(result['test auc'], 3)))
    plt.legend()
    plt.xlabel("False Positive rate")
    plt.ylabel("True Positive rate")
    plt.title("ROC Curves for Train and Test data")
    plt.grid()
    plt.show()
    # Printing confusion matrices code
    print(f"\nConfusion matrix for Train data with {thr_train} as threshold:")
    ax = sns.heatmap(confusion matrix(y train, train predictions), annot=True, fmt='g')
    ax.set yticklabels(['Rejected', 'Accepted'])
    ax.set_xticklabels(['Rejected', 'Accepted'])
    plt.xlabel("Predicted")
    plt.ylabel("Actual")
    plt.title('Confusion matrix for Train')
    plt.show()
    print(f"\nConfusion matrix for Test data with {thr test} as threshold:")
    ax = sns.heatmap(confusion matrix(y test, test predictions), annot=True, fmt='g')
    ax.set yticklabels(['Rejected', 'Accepted'])
    ax.set xticklabels(['Rejected', 'Accepted'])
    plt.xlabel("Predicted")
    plt.ylabel("Actual")
    plt.title('Confusion matrix for Test')
    plt.show()
return result
```

```
In [117]: from wordcloud import WordCloud
          def extra plots(indices, data=X test, wordcloud=True):
              new data = data.iloc[indices, :]
              if wordcloud:
                  essays = new_data['essay']
                  all_essays = ' '.join(essays)
                  wordcloud = WordCloud(width = 800, height = 800, background color ='white', min font size = 10).generat€
                  plt.figure(figsize = (8, 8), facecolor = None)
                  plt.imshow(wordcloud)
                  plt.axis("off")
                  plt.tight layout(pad = 0)
                  plt.show()
              plt.boxplot(new_data['price'])
              plt.title('BoxPlot of price of false negative points')
              plt.show()
              sns.distplot(new data['teacher number of previously posted projects'])
              plt.title('DistPlot of teacher number of previously posted projects')
              plt.show()
```

2.4.1 Applying Decision Trees on BOW, SET 1

In [118]: # Please write all the code with proper documentation
 depth_s = [1, 5, 10, 50, 100]
 mss_s = [5, 10, 50, 100]
 auc_vs_K_plot(bow_train, y_train, depth_s, mss_s)



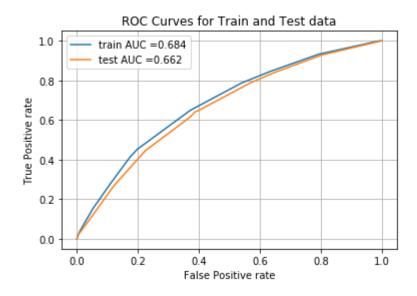


We can see the train score is increasing with max depth but the cv score is reducing after a certain depth value. So this shows the easy overfitting property of the decision trees. Taking depth = 5 and min sample aplit = 5. Not taking depth = 10 as there is

slight more difference between train and cv scores.

```
In [119]: bow_result = {}
bow_result['5,5'] = ROC_conf_mat(bow_train, y_train, bow_test, y_test, 5, 5)
```

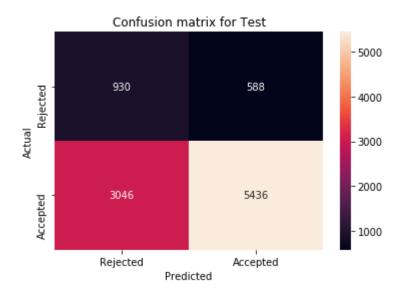
Analysis for max_depth = 5 and min_samples_split = 5



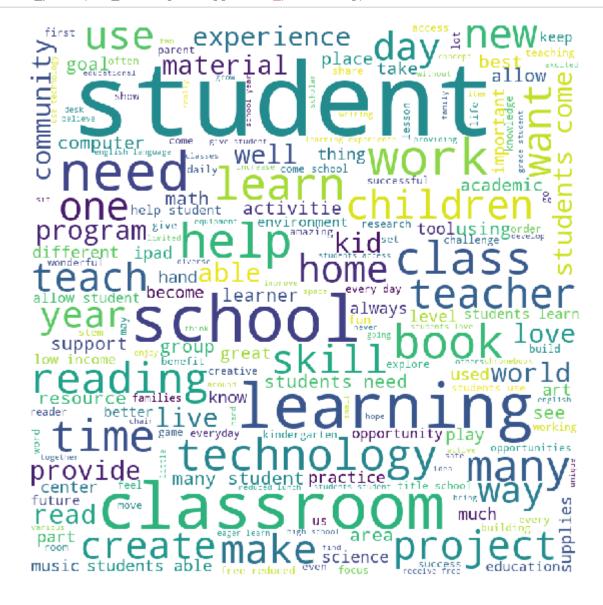
Confusion matrix for Train data with 0.8598726114649682 as threshold:

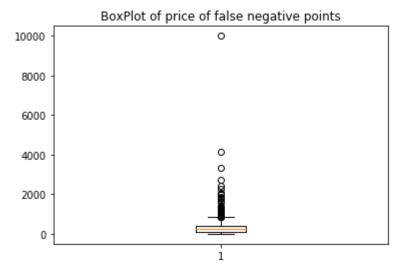


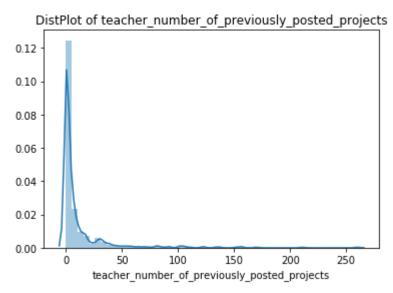
Confusion matrix for Test data with 0.8630393996247655 as threshold:



In [120]: extra_plots(bow_result['5,5']['false_positive'])







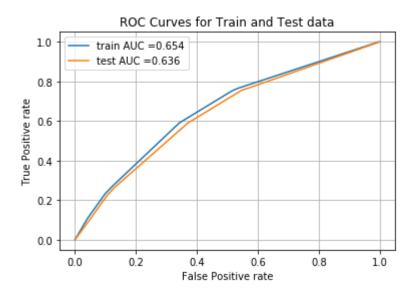
2.4.1.1 Graphviz visualization of Decision Tree on BOW, SET 1

In [121]: # Please write all the code with proper documentation

The Image is not clear. So training on less depth and retrieving the image again

In [123]: bow_result['3,5'] = ROC_conf_mat(bow_train, y_train, bow_test, y_test, 3, 5)

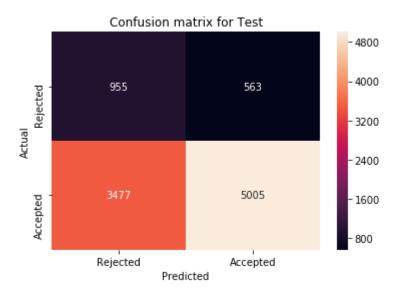
Analysis for max_depth = 3 and min_samples_split = 5



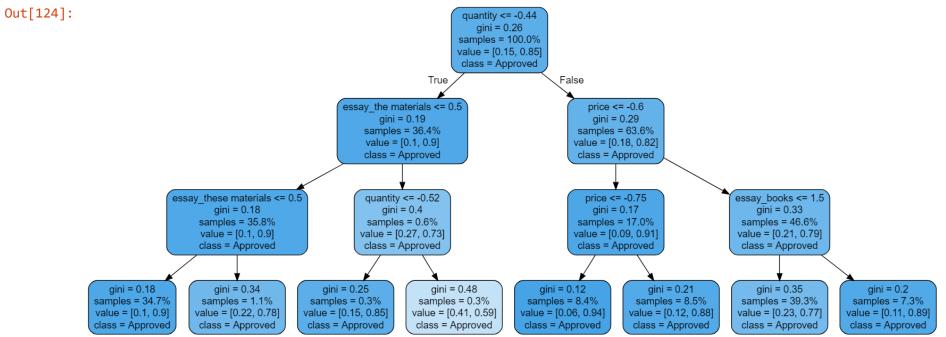
Confusion matrix for Train data with 0.8905085307271824 as threshold:



Confusion matrix for Test data with 0.8905085307271824 as threshold:





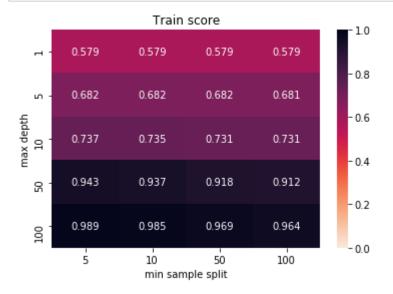


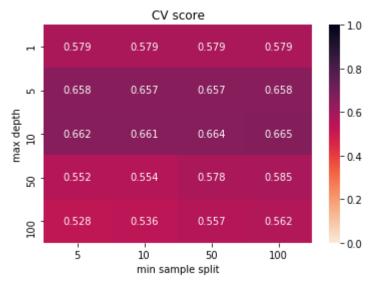
With low depth we can see that our model underfitts a lot as all the leaf nodes are predicted as Approved. This is because of majority of the class. But with depth = 5 we can see rejected leaf nodes in previous image (indicated by orange nodes)

GraphViz images are not created by command line. I manually converted dot files to png in online. So when running the above codes (and any other codes which come after this which outputs graphviz) will show error.

2.4.2 Applying Decision Trees on TFIDF, SET 2

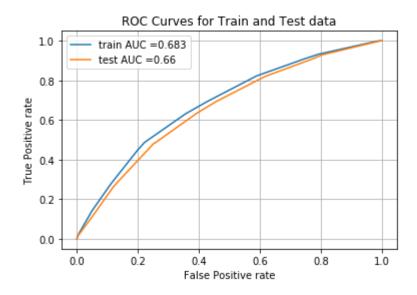
In [125]: # Please write all the code with proper documentation
 depth_s = [1, 5, 10, 50, 100]
 mss_s = [5, 10, 50, 100]
 auc_vs_K_plot(tfidf_train, y_train, depth_s, mss_s)





```
In [126]: tfidf_result = {}
tfidf_result['5,50'] = ROC_conf_mat(tfidf_train, y_train, tfidf_test, y_test, 5, 50)
```

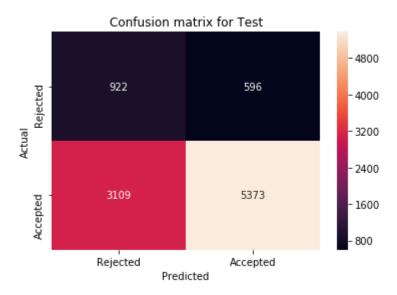
Analysis for max_depth = 5 and min_samples_split = 50



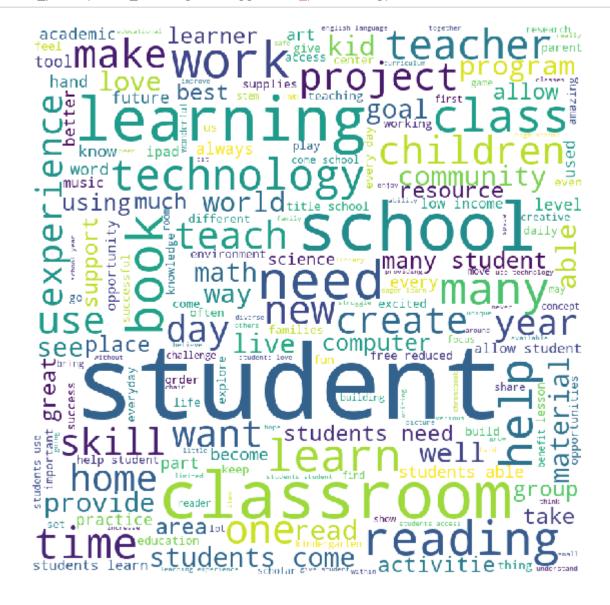
Confusion matrix for Train data with 0.8570432357043236 as threshold:

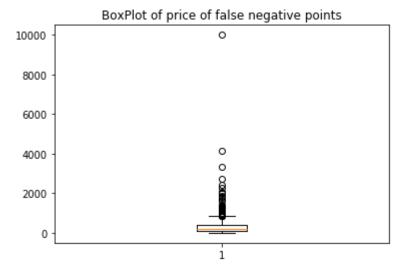


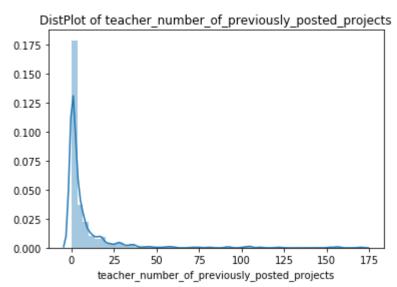
Confusion matrix for Test data with 0.8570432357043236 as threshold:



In [127]: extra_plots(tfidf_result['5,50']['false_positive'])







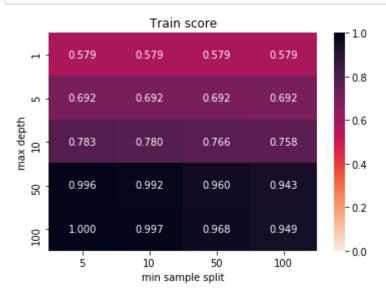
2.4.2.1 Graphviz visualization of Decision Tree on TFIDF, SET 2

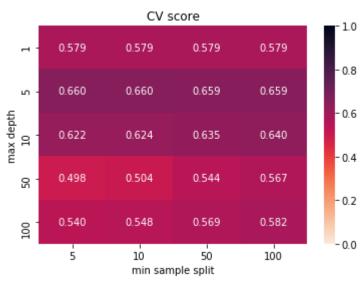
Training a less depth model to create graphviz image

```
In [128]:
              # Please write all the code with proper documentation
               tfidf result['3,5'] = ROC conf mat(tfidf train, y train, tfidf test, y test, 3, 5, plots=False)
               export graphviz(tfidf result['3,5']['model'], out file='tree tfidf.dot', feature names=tfidf columns,\
                                      class names=['Rejected', 'Approved'], rounded = True, proportion = True, precision = 2, filled =
               Image(filename='tree tfidf 3 5.png')
Out[128]:
                                                                                 essay_materials <= 0.04
                                                                                       gini = 0.26
                                                                                   samples = 100.0%
                                                                                   value = [0.15, 0.85]
                                                                                   class = Approved
                                                                                 True /
                                                                                                     False
                                                                                                          price <= -0.5
                                                                         quantity <= -0.39
                                                                            gini = 0.23
                                                                                                          gini = 0.35
                                                                         samples = 79.6%
                                                                                                        samples = 20.4%
                                                                         value = [0.13, 0.87]
                                                                                                       value = [0.23, 0.77]
                                                                         class = Approved
                                                                                                        class = Approved
                                     essay feed <= 0.13
                                                                           price <= -0.23
                                                                                                      essay reason <= 0.1
                                                                                                                                            quantity <= -0.5
                                        gini = 0.17
                                                                            gini = 0.27
                                                                                                           gini = 0.23
                                                                                                                                              gini = 0.39
                                      samples = 34.7%
                                                                         samples = 44.8%
                                                                                                        samples = 5.2\%
                                                                                                                                           samples = 15.3%
                                     value = [0.09, 0.91]
                                                                         value = [0.16, 0.84]
                                                                                                       value = [0.13, 0.87]
                                                                                                                                           value = [0.26, 0.74]
                                      class = Approved
                                                                         class = Approved
                                                                                                        class = Approved
                                                                                                                                           class = Approved
                    gini = 0.17
                                        gini = 0.46
                                                             gini = 0.21
                                                                                 gini = 0.34
                                                                                                     gini = 0.23
                                                                                                                                              aini = 0.22
                                                                                                                                                                  gini = 0.42
                                                                                                                          gini = 0.22
                 samples = 34.7%
                                      samples = 0.0%
                                                          samples = 26.7%
                                                                              samples = 18.1%
                                                                                                   samples = 5.2\%
                                                                                                                        samples = 0.0\%
                                                                                                                                            samples = 3.1%
                                                                                                                                                               samples = 12.2%
                 value = [0.09, 0.91]
                                                          value = [0.12, 0.88]
                                                                              value = [0.22, 0.78]
                                                                                                  value = [0.13, 0.87]
                                                                                                                       value = [0.88, 0.12]
                                                                                                                                                               value = [0.3, 0.7]
                                     value = [0.65, 0.35]
                                                                                                                                           value = [0.13, 0.87]
                 class = Approved
                                      class = Rejected
                                                          class = Approved
                                                                              class = Approved
                                                                                                   class = Approved
                                                                                                                        class = Rejected
                                                                                                                                           class = Approved
                                                                                                                                                               class = Approved
```

2.4.3 Applying Decision Trees on AVG W2V, SET 3

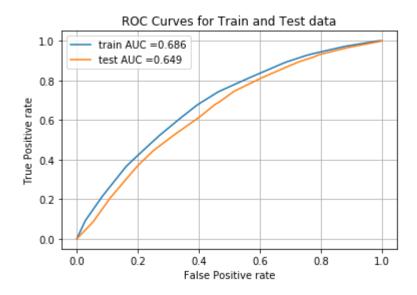
In [129]: # Please write all the code with proper documentation
 depth_s = [1, 5, 10, 50, 100]
 mss_s = [5, 10, 50, 100]
 auc_vs_K_plot(avgw2v_train, y_train, depth_s, mss_s)





```
In [130]: avgw2v_result = {}
avgw2v_result['5,10'] = ROC_conf_mat(avgw2v_train, y_train, avgw2v_test, y_test, 5, 10)
```

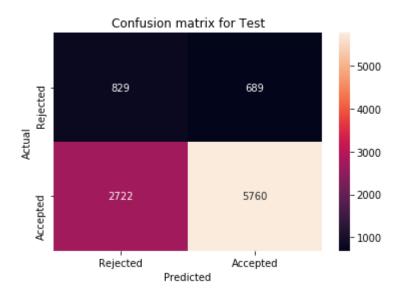
Analysis for max_depth = 5 and min_samples_split = 10



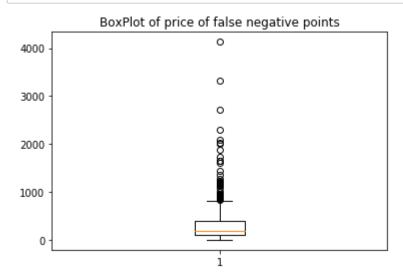
Confusion matrix for Train data with 0.8704453441295547 as threshold:

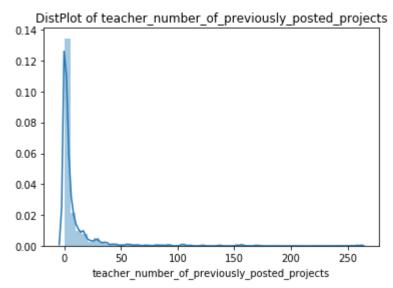


Confusion matrix for Test data with 0.8704453441295547 as threshold:



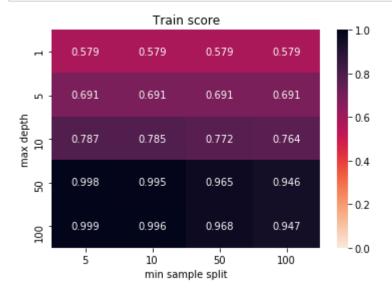
In [131]: extra_plots(avgw2v_result['5,10']['false_positive'], wordcloud=False)

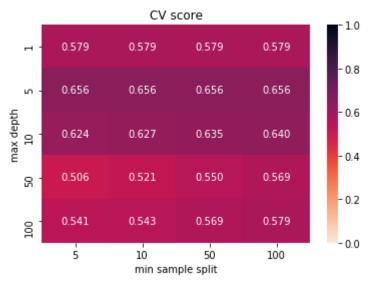




2.4.4 Applying Decision Trees on TFIDF W2V, SET 4

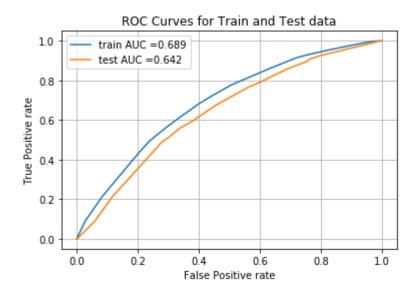
In [132]: # Please write all the code with proper documentation
 depth_s = [1, 5, 10, 50, 100]
 mss_s = [5, 10, 50, 100]
 auc_vs_K_plot(tfidfw2v_train, y_train, depth_s, mss_s)





```
In [133]: tfidfw2v_result = {}
tfidfw2v_result['5,5'] = ROC_conf_mat(tfidfw2v_train, y_train, tfidfw2v_test, y_test, 5, 5)
```

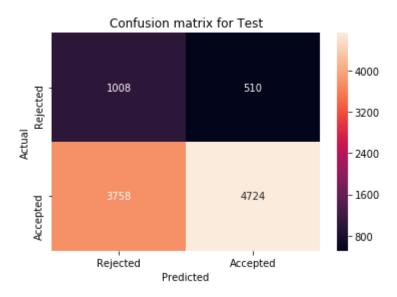
Analysis for max_depth = 5 and min_samples_split = 5



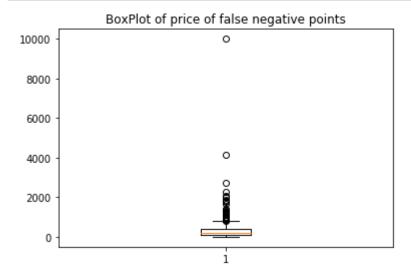
Confusion matrix for Train data with 0.8596813725490197 as threshold:

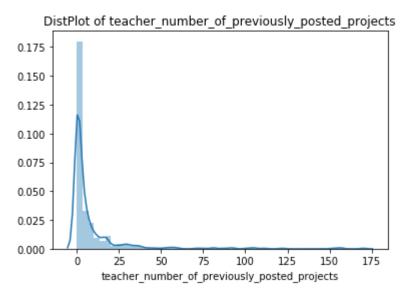


Confusion matrix for Test data with 0.8725824800910125 as threshold:



In [134]: extra_plots(tfidfw2v_result['5,5']['false_positive'], wordcloud=False)



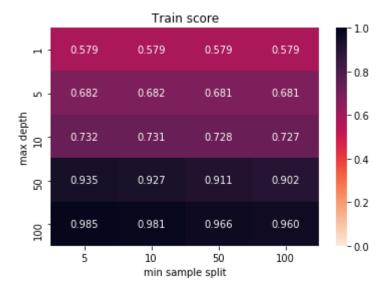


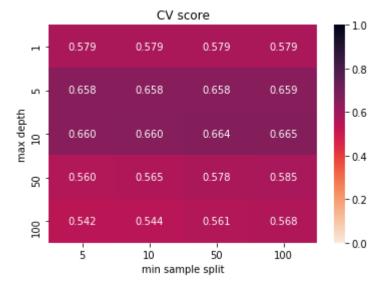
2.5 [Task-2]Getting top 5k features using `feature_importances_`

```
In [135]: # please write all the code with proper documentation, and proper titles for each subsection
# go through documentations and blogs before you start coding
# first figure out what to do, and then think about how to do.
# reading and understanding error messages will be very much helpfull in debugging your code
# when you plot any graph make sure you use
# a. Title, that describes your plot, this will be very helpful to the reader
# b. Legends if needed
# c. X-axis label
# d. Y-axis label
```

Taking Set-2 trained model to get top features.

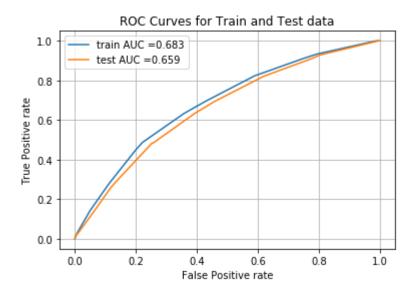
```
In [137]: depth_s = [1, 5, 10, 50, 100]
    mss_s = [5, 10, 50, 100]
    auc_vs_K_plot(task2_train, y_train, depth_s, mss_s)
```





```
In [138]: task2_result = {}
task2_result['5,5'] = ROC_conf_mat(task2_train, y_train, task2_test, y_test, 5, 5)
```

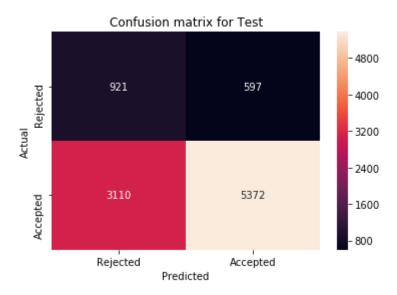
Analysis for max_depth = 5 and min_samples_split = 5



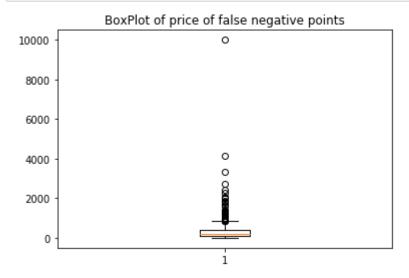
Confusion matrix for Train data with 0.8570432357043236 as threshold:

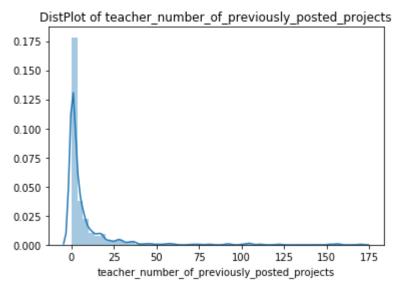


Confusion matrix for Test data with 0.8570432357043236 as threshold:



In [139]: extra_plots(task2_result['5,5']['false_positive'], wordcloud=False)





I used only max_depth upto 100 only as I can see even with 100 depth the models are highly overfitting and increasing depth only makes it worse. So I considered only upto 100 depth. And min_sample_split also I considered only upto 100 as it doesnt giving much performance to our models

3. Conclusion

```
In [140]: # Please compare all your models using Prettytable library
```

```
In [141]: from prettytable import PrettyTable
    table = PrettyTable()
    table.field_names = ['Vectorizer', 'max_depth', 'min_samples_split', 'Train AUC', 'Test AUC']
    table.add_row(['BOW', '5', '5', np.round(bow_result['5,5']['train_auc'], 3), np.round(bow_result['5,5']['test_au table.add_row(['TfIdf', '5', '50', np.round(tfidf_result['5,50']['train_auc'], 3), np.round(tfidf_result['5,50']
    table.add_row(['Average w2v', '5', '10', np.round(avgw2v_result['5,10']['train_auc'], 3), np.round(avgw2v_result table.add_row(['TfIdf w2v', '5', '5', np.round(tfidfw2v_result['5,5']['train_auc'], 3), np.round(tfidfw2v_result['5,5']['train_auc'], 3), np.round(task2_result['5,5']['train_auc'], 3), np.round(task2_result['
```

| Vectorizer | + max_depth + | min_samples_split | Train AUC | + Test AUC |
|--|-----------------------|-------------------|-----------|-----------------|
| BOW TfIdf Average w2v TfIdf w2v TfIdf Task 2 | 5 | 5 | 0.684 | 0.662 |
| | 5 | 50 | 0.683 | 0.66 |
| | 5 | 10 | 0.686 | 0.649 |
| | 5 | 5 | 0.689 | 0.642 |
| | 5 | 5 | 0.683 | 0.659 |

SUMMARY:

- We can see all the models are doing similar. But BOW, Tfldf has slight better results. And if we see the train auc (in above codes for all models), for higher depth values we can see a lot overfitting.
- Decision trees didnt perform well when compared to other models i.e. Logistic Regression, SVM etc. I think because of high dimensionality in data.
- And after plotting word cloud of false positive points, we see words that are common in these essays (i.e. we expect them
 to be in high numbers as these essays are related to school projects). Words like student, school, learning, classroom are
 there in wordcloud from which we cant say much as they are common in these essays
- The pdf and boxplots of price and teacher_number_of_prev_posted_proj plots of false positive points seems to be having small values.
- Taking top 5K important features (Task 2) seems to be a good idea as the performance didnt reduce much. But If we see feature_importance values of features, most of them have a value of zero. So I guess we can reduce the dimention even furthur without reducing models performance

Note: The images of graphviz are created by converting the dot files externally (using online tools) and not by the code. So They are found in pdf but if we re-run the notebook the images may not be produced and results in error at that code blocks

| In []: | ı: [| |
|---------|------|--|
| | | |