DonorsChoose

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

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

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

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

About the DonorsChoose Data Set

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

| Feature | Description |
|------------|--------------------------------------------------------|
| project_id | A unique identifier for the proposed parample: p036502 |

| Feature | Description |
|----------------------------|-------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------|
| project_title | Title of the project. Examples: • Art Will Make You Happy • First Grade Fun |
| project_grade_category | Grade level of students for which the project is targeted. One of the follow enumerated values: • Grades PreK-2 • Grades 3-5 • Grades 6-8 • Grades 9-12 |
| project_subject_categories | One or more (comma-separated) su categories for the project from the fo enumerated list of values: • Applied Learning • Care & Hunger • Health & Sports • History & Civics • Literacy & Language • Math & Science • Music & The Arts • Special Needs • Warmth Examples: • Music & The Arts • Literacy & Language, Mark & Science |

| Feature | Description | |
|------------------------------------------|----------------------------------------------------------------------------------------------------------------------------------------|--|
| school_state | State where school is located (<u>Two-lu.S. postal code</u>). Example: WY | |
| <pre>project_subject_subcategories</pre> | One or more (comma-separated) su subcategories for the project. Exam • Literacy • Literature & Writing, Social Sciences | |
| project_resource_summary | An explanation of the resources need the project. Example: • My students need hands 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 w submitted. Example : 2016-04-28 12:43:56.245 | |
| teacher_id | A unique identifier for the teacher of proposed project. Example: bdf8baa8fedef6bfeec7ae4ff1c | |

| Feature | Description |
|----------------------------------------------|-------------------------------------------------------------------------------------|
| | Teacher's title. One of the following enumerated values: |
| | • nan |
| teacher_prefix | • Dr. |
| | • Mr. |
| | • Mrs. |
| | • Ms. |
| | • Teacher. |
| teacher_number_of_previously_posted_projects | Number of project applications previous submitted by the same teacher. Exa 2 |

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

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

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

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

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

| Label | Description |
|---------------------|---------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------|
| project_is_approved | A binary flag indicating whether DonorsChoose approved the project. A value of θ indicates the project was not approved, and a value of θ indicates the project was approved. |

Notes on the Essay Data

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

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

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

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

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

```
In [88]: %matplotlib inline
   import warnings
   warnings.filterwarnings("ignore")

import sqlite3
   import pandas as pd
   import numpy as np
```

```
import nltk
import string
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.feature extraction.text import TfidfTransformer
from sklearn.feature extraction.text import TfidfVectorizer
from sklearn.feature extraction.text import CountVectorizer
from sklearn.metrics import confusion matrix
from sklearn import metrics
from sklearn.metrics import roc curve, auc
from nltk.stem.porter import PorterStemmer
import re
# Tutorial about Python regular expressions: https://pymotw.com/2/re/
import string
from nltk.corpus import stopwords
from nltk.stem import PorterStemmer
from nltk.stem.wordnet import WordNetLemmatizer
from gensim.models import Word2Vec
from gensim.models import KeyedVectors
import pickle
from tqdm import tqdm
import os
from plotly import plotly
import plotly.offline as offline
import plotly.graph objs as go
offline.init notebook mode()
from collections import Counter
IOPub data rate exceeded.
The notebook server will temporarily stop sending output
to the client in order to avoid crashing it.
To change this limit, set the config variable
`--NotebookApp.iopub data rate limit`.
```

1.1 Reading Data

```
In [89]: project data = pd.read csv('train data.csv')
         resource data = pd.read csv('resources.csv')
In [90]:
         print("Number of data points in train data", project data.shape)
         print('-'*50)
         print("The attributes of data :", project data.columns.values)
         Number of data points in train data (109248, 17)
         The attributes of data : ['Unnamed: 0' 'id' 'teacher_id' 'teacher_prefi
         x' 'school state'
           'project submitted datetime' 'project grade category'
           'project_subject_categories' 'project_subject_subcategories'
           'project_title' 'project_essay_1' 'project_essay_2' 'project_essay_3'
           'project essay 4' 'project resource summary'
           'teacher number of previously posted projects' 'project is approved']
In [91]: 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[91]:
                                                    description quantity
                 id
                                                                       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 [92]: catogories = list(project data['project subject categories'].values)
```

```
# remove special characters from list of strings python: https://stacko
verflow.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 & H
unaer"
    for j in i.split(','): # it will split it in three parts ["Math & S
cience", "Warmth", "Care & Hunger"]
       if 'The' in j.split(): # this will split each of the catogory b
ased on space "Math & Science"=> "Math", "&", "Science"
            j=j.replace('The','') # if we have the words "The" we are g
oing to replace it with ''(i.e removing 'The')
       j = j.replace(' ','') # we are placeing all the ' '(space) with
 ''(empty) ex:"Math & Science"=>"Math&Science"
       temp+=j.strip()+" " #" abc ".strip() will return "abc", remove
the trailing spaces
       temp = temp.replace('&',' ') # we are replacing the & value int
    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 [93]: sub catogories = list(project data['project subject subcategories'].val
         ues)
         # remove special characters from list of strings python: https://stacko
         verflow.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 & H
         unger"
             for j in i.split(','): # it will split it in three parts ["Math & S
         cience", "Warmth", "Care & Hunger"]
                 if 'The' in j.split(): # this will split each of the catogory b
         ased on space "Math & Science"=> "Math", "&", "Science"
                     j=j.replace('The','') # if we have the words "The" we are g
         oing to replace it with ''(i.e removing 'The')
                 j = j.replace(' ','') # we are placeing all the ' '(space) with
          ''(empty) ex:"Math & Science"=>"Math&Science"
                 temp +=j.strip()+" "#" abc ".strip() will return "abc", remove
          the trailing spaces
                 temp = temp.replace('&',' ')
             sub cat list.append(temp.strip())
         project data['clean subcategories'] = sub cat list
         project data.drop(['project subject subcategories'], axis=1, inplace=Tr
         ue)
         # 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

In [95]: project_data.head(2)

Out[95]:

| | Unnamed: | id | teacher_id | teacher_prefix | school_state |
|---|----------|---------|----------------------------------|----------------|--------------|
| 0 | 160221 | p253737 | c90749f5d961ff158d4b4d1e7dc665fc | Mrs. | IN |

| | Unnamed: 0 | id | teacher_id | teacher_prefix | school_state |
|---|---------------|---------|----------------------------------|----------------|--------------|
| 1 | 140945 | p258326 | 897464ce9ddc600bced1151f324dd63a | Mr. | FL |

```
In [96]: #### 1.4.2.3 Using Pretrained Models: TFIDF weighted W2V
```

```
In [97]: # printing some random reviews
    print(project_data['essay'].values[0])
    print(project_data['essay'].values[150])
    print("="*50)
    print(project_data['essay'].values[1000])
    print("="*50)
    print(project_data['essay'].values[20000])
```

My students are English learners that are working on English as their s econd or third languages. We are a melting pot of refugees, immigrants, and native-born Americans bringing the gift of language to our school. \r\n\r\n We have over 24 languages represented in our English Learner p rogram with students at every level of mastery. We also have over 40 c ountries represented with the families within our school. Each student brings a wealth of knowledge and experiences to us that open our eyes t o new cultures, beliefs, and respect.\"The limits of your language are the limits of your world.\"-Ludwig Wittgenstein Our English learner's have a strong support system at home that begs for more resources. Man y times our parents are learning to read and speak English along side of their children. Sometimes this creates barriers for parents to be ab

le to help their child learn phonetics, letter recognition, and other r eading skills.\r\n\r\nBy providing these dvd's and players, students ar e able to continue their mastery of the English language even if no one at home is able to assist. All families with students within the Level 1 proficiency status, will be a offered to be a part of this program. These educational videos will be specially chosen by the English Learne r Teacher and will be sent home regularly to watch. The videos are to help the child develop early reading skills.\r\n\r\nParents that do not have access to a dvd player will have the opportunity to check out a dv d player to use for the year. The plan is to use these videos and educ ational dvd's for the years to come for other EL students.\r\nnannan

The 51 fifth grade students that will cycle through my classroom this y ear all love learning, at least most of the time. At our school, 97.3% of the students receive free or reduced price lunch. Of the 560 student s, 97.3% are minority students. \r\nThe school has a vibrant community that loves to get together and celebrate. Around Halloween there is a w hole school parade to show off the beautiful costumes that students wea r. On Cinco de Mayo we put on a big festival with crafts made by the st udents, dances, and games. At the end of the year the school hosts a ca rnival to celebrate the hard work put in during the school year, with a dunk tank being the most popular activity. My students will use these fi ve brightly colored Hokki stools in place of regular, stationary, 4-leg ged chairs. As I will only have a total of ten in the classroom and not enough for each student to have an individual one, they will be used in a variety of ways. During independent reading time they will be used as special chairs students will each use on occasion. I will utilize them in place of chairs at my small group tables during math and reading tim es. The rest of the day they will be used by the students who need the highest amount of movement in their life in order to stay focused on sc hool.\r\n\r\nWhenever asked what the classroom is missing, my students always say more Hokki Stools. They can't get their fill of the 5 stools we already have. When the students are sitting in group with me on the Hokki Stools, they are always moving, but at the same time doing their work. Anytime the students get to pick where they can sit, the Hokki St ools are the first to be taken. There are always students who head over to the kidney table to get one of the stools who are disappointed as th ere are not enough of them. \r\n\r\nWe ask a lot of students to sit for 7 hours a day. The Hokki stools will be a compromise that allow my stud ents to do desk work and move at the same time. These stools will help students to meet their 60 minutes a day of movement by allowing them to activate their core muscles for balance while they sit. For many of my students, these chairs will take away the barrier that exists in school s for a child who can't sit still.nannan

How do you remember your days of school? Was it in a sterile environmen t with plain walls, rows of desks, and a teacher in front of the room? A typical day in our room is nothing like that. I work hard to create a warm inviting themed room for my students look forward to coming to eac h day.\r\nMy class is made up of 28 wonderfully unique boys and gir ls of mixed races in Arkansas.\r\nThey attend a Title I school, which m eans there is a high enough percentage of free and reduced-price lunch to qualify. Our school is an \"open classroom\" concept, which is very unique as there are no walls separating the classrooms. These 9 and 10 year-old students are very eager learners; they are like sponges, absor bing all the information and experiences and keep on wanting more. With these resources such as the comfy red throw pillows and the whimsical n autical hanging decor and the blue fish nets, I will be able to help cr eate the mood in our classroom setting to be one of a themed nautical e nvironment. Creating a classroom environment is very important in the s uccess in each and every child's education. The nautical photo props wi ll be used with each child as they step foot into our classroom for the first time on Meet the Teacher evening. I'll take pictures of each chil d with them, have them developed, and then hung in our classroom ready for their first day of 4th grade. This kind gesture will set the tone before even the first day of school! The nautical thank you cards will be used throughout the year by the students as they create thank you ca rds to their team groups.\r\n\r\nYour generous donations will help me t o help make our classroom a fun, inviting, learning environment from da y one.\r\n\r\nIt costs lost of money out of my own pocket on resources to get our classroom ready. Please consider helping with this project t o make our new school vear a very successful one. Thank vou!nannan

My kindergarten students have varied disabilities ranging from speech a nd language delays, cognitive delays, gross/fine motor delays, to autis m. They are eager beavers and always strive to work their hardest working past their limitations. \r\n\r\nThe materials we have are the ones I seek out for my students. I teach in a Title I school where most of the

students receive free or reduced price lunch. Despite their disabiliti es and limitations, my students love coming to school and come eager to learn and explore. Have you ever felt like you had ants in your pants and you needed to groove and move as you were in a meeting? This is how my kids feel all the time. The want to be able to move as they learn or so they say. Wobble chairs are the answer and I love then because they develop their core, which enhances gross motor and in Turn fine motor skills. \r\nThey also want to learn through games, my kids don't want to sit and do worksheets. They want to learn to count by jumping and playing. Physical engagement is the key to our success. The number toss and color and shape mats can make that happen. My students will forget they are doing work and just have the fun a 6 year old deserves.nannan

```
In [98]: # 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 [99]: sent = decontracted(project_data['essay'].values[20000])
    print(sent)
    print("="*50)
```

My kindergarten students have varied disabilities ranging from speech a nd language delays, cognitive delays, gross/fine motor delays, to autis

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

```
In [100]: # \r \n \t remove from string python: http://texthandler.com/info/remov
    e-line-breaks-python/
    sent = sent.replace('\\r', ' ')
    sent = sent.replace('\\"', ' ')
    sent = sent.replace('\\"', ' ')
    print(sent)
```

My kindergarten students have varied disabilities ranging from speech a nd language delays, cognitive delays, gross/fine motor delays, to autis m. They are eager beavers and always strive to work their hardest worki ng past their limitations. The materials we have are the ones I see k out for my students. I teach in a Title I school where most of the st udents receive free or reduced price lunch. Despite their disabilities and limitations, my students love coming to school and come eager to le arn and explore. Have you ever felt like you had ants in your pants and you needed to groove and move as you were in a meeting? This is how my kids feel all the time. The want to be able to move as they learn or so they say. Wobble chairs are the answer and I love then because they deve lop their core, which enhances gross motor and in Turn fine motor skill They also want to learn through games, my kids do not want to sit and do worksheets. They want to learn to count by jumping and playing. Physical engagement is the key to our success. The number toss and colo r and shape mats can make that happen. My students will forget they are doing work and just have the fun a 6 year old deserves.nannan

```
In [101]: #remove spacial character: https://stackoverflow.com/a/5843547/4084039
    sent = re.sub('[^A-Za-z0-9]+', ' ', sent)
    print(sent)
```

Mv kindergarten students have varied disabilities ranging from speech a nd language delays cognitive delays gross fine motor delays to autism T hey are eager beavers and always strive to work their hardest working p ast their limitations The materials we have are the ones I seek out for my students I teach in a Title I school where most of the students rece ive free or reduced price lunch Despite their disabilities and limitati ons my students love coming to school and come eager to learn and explo re Have you ever felt like you had ants in your pants and you needed to groove and move as you were in a meeting This is how my kids feel all t he time The want to be able to move as they learn or so they say Wobble chairs are the answer and I love then because they develop their core w hich enhances gross motor and in Turn fine motor skills They also want to learn through games my kids do not want to sit and do worksheets The y want to learn to count by jumping and playing Physical engagement is the key to our success The number toss and color and shape mats can mak e that happen My students will forget they are doing work and just have the fun a 6 year old deserves nannan

```
'because', 'as', 'until', 'while', 'of', \
                       'at', 'by', 'for', 'with', 'about', 'against', 'between',
           'into', 'through', 'during', 'before', 'after',\
                      'above', 'below', 'to', 'from', 'up', 'down', 'in', 'out',
          'on', 'off', 'over', 'under', 'again', 'further',\
                      'then', 'once', 'here', 'there', 'when', 'where', 'why', 'h
          ow', 'all', 'any', 'both', 'each', 'few', 'more',\
                      'most', 'other', 'some', 'such', 'only', 'own', 'same', 's
          o', 'than', 'too', 'very', \
                      's', 't', 'can', 'will', 'just', 'don', "don't", 'should',
          "should've", 'now', 'd', 'll', 'm', 'o', 're', \
                      've', 'y', 'ain', 'aren', "aren't", 'couldn', "couldn't",
           'didn', "didn't", 'doesn', "doesn't", 'hadn',\
                      "hadn't", 'hasn', "hasn't", 'haven', "haven't", 'isn', "is
          n't", 'ma', 'mightn', "mightn't", 'mustn',\
                      "mustn't", 'needn', "needn't", 'shan', "shan't", 'shouldn',
           "shouldn't", 'wasn', "wasn't", 'weren', "weren't", \
                       'won', "won't", 'wouldn', "wouldn't"]
In [103]: # Combining all the above stundents
          from tgdm import tgdm
          preprocessed essays = []
          # tqdm is for printing the status bar
          for sentance in tqdm(project data['essay'].values):
              sent = decontracted(sentance)
              sent = sent.replace('\\r', ' ')
              sent = sent.replace('\\"', ' ')
              sent = sent.replace('\\n', ' ')
              sent = re.sub('[^A-Za-z0-9]+', ' ', sent)
              # https://gist.github.com/sebleier/554280
              sent = ' '.join(e for e in sent.split() if e not in stopwords)
              preprocessed essays.append(sent.lower().strip())
          100%|
                   109248/109248 [02:14<00:00, 812.41it/s]
In [104]: # after preprocesing
          preprocessed essays[20000]
```

Out[104]: 'my kindergarten students varied disabilities ranging speech language d elays cognitive delays gross fine motor delays autism they eager beaver s always strive work hardest working past limitations the materials one s i seek students i teach title i school students receive free reduced price lunch despite disabilities limitations students love coming school come eager learn explore have ever felt like ants pants needed groove move meeting this kids feel time the want able move learn say wobble chairs answer i love develop core enhances gross motor turn fine motor skills they also want learn games kids not want sit worksheets they want learn count jumping playing physical engagement key success the number toss color shape mats make happen my students forget work fun 6 year old deserves nannan'

```
In [235]: project_data["preprocessed_essays"] = preprocessed_essays
    essay_word_count = []

for sentence in project_data["preprocessed_essays"] :
    word = len(sentence.split())
    essay_word_count.append(word)

project_data["essay_word_count"] = essay_word_count
project_data.head(5)
```

Out[235]:

| | Unnamed: 0 | id | teacher_id | teacher_prefix | school_state |
|---|---------------|---------|----------------------------------|----------------|--------------|
| 0 | 160221 | p253737 | c90749f5d961ff158d4b4d1e7dc665fc | Mrs. | IN |

| Mr. | FL |
|------|------|
| | |
| Ms. | AZ |
| Mrs. | кү |
| Mrs. | TX |
| | Mrs. |

5 rows × 23 columns

```
In [245]: sid = SentimentIntensityAnalyzer()
    neg = []
    pos = []
    neu = []
```

```
compound = []
for a in tqdm(project_data["preprocessed_essays"]) :
    b = sid.polarity_scores(a)['neg']
    c = sid.polarity_scores(a)['pos']
    d = sid.polarity_scores(a)['neu']
    e = sid.polarity_scores(a)['compound']
    neg.append(b)
    pos.append(c)
   neu.append(d)
    compound.append(e)
project data["pos"] = pos
project data["neg"] = neg
project data["neu"] = neu
project data["compound"] = compound
project_data.head(5)
100%|
        | 109248/109248 [25:12<00:00, 72.23it/s]
```

Out[245]:

| | Unnamed: | id | teacher_id | teacher_prefix | school_state |
|---|----------|---------|----------------------------------|----------------|--------------|
| 0 | 160221 | p253737 | c90749f5d961ff158d4b4d1e7dc665fc | Mrs. | IN |
| 1 | 140945 | p258326 | 897464ce9ddc600bced1151f324dd63a | Mr. | FL |

| | Unnamed: 0 | id | teacher_id | teacher_prefix | school_state |
|---|---------------|---------|----------------------------------|----------------|--------------|
| 2 | 21895 | p182444 | 3465aaf82da834c0582ebd0ef8040ca0 | Ms. | AZ |
| 3 | 45 | p246581 | f3cb9bffbba169bef1a77b243e620b60 | Mrs. | KY |
| 4 | 172407 | p104768 | be1f7507a41f8479dc06f047086a39ec | Mrs. | TX |

5 rows × 27 columns

1.4 Preprocessing of `project_title`

```
Interactive Math Tools
     ______
     Flexible Seating for Mrs. Jarvis' Terrific Third Graders!!
     Chromebooks for Special Education Reading Program
     ______
     It's the 21st Century
     ______
     Targeting More Success in Class
     Just For the Love of Reading--\r\nPure Pleasure
     ______
     Reading Changes Lives
     ______
     Elevating Academics and Parent Rapports Through Technology
     _____
     Building Life Science Experiences
     Everyone deserves to be heard!
     ______
     TABLETS CAN SHOW US THE WORLD
     ______
     Making Recess Active
     ______
     Making Great LEAP's With Leapfrog!
     Technology Teaches Tomorrow's Talents Today
     Test Time
     ______
     Wiggling Our Way to Success
     _____
     Magic Carpet Ride in Our Library
     ______
In [106]: preprocessed_titles = []
     for dataset in tqdm(project_data['project_title'].values):
```

```
data = decontracted(dataset) # Replacing some specific and general
          short form into proper word/stopword.
             data = re.sub(r"it's", "it is", data) # Replacing it's with it is a
         s it is not part of function decontracted
             data = data.replace('\\r', ' ') # Replacing \r with space
             data = data.replace('\\"', ' ') # Replacing \ with space
             data = data.replace('\\n', ' ') # Replacing \n with space
             data = re.sub('[^A-Za-z0-9]+', ' ', data) # Replacing special chara
         cters with space
             data = re.sub("\S*\d\S*", "", data).strip() # Trimming numbers cont
         aining digits
             data = ' '.join(e for e in data.split() if e not in stopwords) # Re
         movina stopwords
             preprocessed titles.append(data.lower().strip()) # Creating array i
         n all the lower cases.
         100%
              | 109248/109248 [00:07<00:00, 14572.40it/s]
In [107]: for i in range (0,21):
             print(preprocessed titles[i])
             print("="*50)
         educational support english learners home
         wanted projector hungry learners
         soccer equipment awesome middle school students
         _____
         techie kindergarteners
         interactive math tools
         _____
         flexible seating mrs jarvis terrific third graders
         chromebooks special education reading program
         _____
         it centurv
```

```
targeting more success class
      ______
      just for love reading pure pleasure
      ______
      reading changes lives
      _____
      elevating academics parent rapports through technology
      building life science experiences
      ______
      everyone deserves heard
      tablets can show us the world
      _____
      making recess active
        ______
      making great leap with leapfrog
      ______
      technology teaches tomorrow talents today
      test time
      _____
      wiggling our way success
      _____
      magic carpet ride our library
      ______
In [234]: project data["preprocessed titles"] = preprocessed titles
      title word count = []
      for sentence in project data["preprocessed titles"] :
        word = len(sentence.split())
        title word count.append(word)
      project_data["title_word_count"] = title_word_count
      project data.head(5)
```

Out[234]:

| | Unnamed: | id | teacher_id | teacher_prefix | school_state |
|---|----------|---------|----------------------------------|----------------|--------------|
| 0 | 160221 | p253737 | c90749f5d961ff158d4b4d1e7dc665fc | Mrs. | IN |
| 1 | 140945 | p258326 | 897464ce9ddc600bced1151f324dd63a | Mr. | FL |
| 2 | 21895 | p182444 | 3465aaf82da834c0582ebd0ef8040ca0 | Ms. | AZ |
| 3 | 45 | p246581 | f3cb9bffbba169bef1a77b243e620b60 | Mrs. | KY |

| | Unnamed: 0 | id | teacher_id | teacher_prefix | school_state |
|---|---------------|---------|----------------------------------|----------------|--------------|
| 4 | 172407 | p104768 | be1f7507a41f8479dc06f047086a39ec | Mrs. | TX |

5 rows × 21 columns

1.5 Preparing data for models

```
In [108]: project_data.columns
Out[108]: Index(['Unnamed: 0', 'id', 'teacher_id', 'teacher_prefix', 'school_stat
         е',
                'project submitted datetime', 'project grade category', 'project
         title'
                'project essay 4', 'project resource summary',
                'teacher number of previously posted projects', 'project is appr
         oved',
                'clean categories', 'clean subcategories', 'essay'],
               dtype='object')
         we are going to consider
               - school state : categorical data
               - clean categories : categorical data
               - clean subcategories : categorical data
               - project grade category : categorical data
               - teacher prefix : categorical data
```

```
- project_title : text data
- text : text data
- project_resource_summary: text data (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/

```
In [109]: # 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()), l
          owercase=False, binary=True)
          categories one hot = vectorizer.fit transform(project data['clean categories')
          ories'l.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', 'AppliedLearn
          ing', 'SpecialNeeds', 'Health Sports', 'Math Science', 'Literacy Langua
          ge']
          Shape of matrix after one hot encodig (109248, 9)
In [110]: # 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 s
          ubcategories'l.values)
```

```
print(vectorizer.get feature names())
          print("Shape of matrix after one hot encodig ", sub categories one hot.s
          hape)
          ['Economics', 'CommunityService', 'FinancialLiteracy', 'ParentInvolveme
          nt', 'Extracurricular', 'Civics Government', 'ForeignLanguages', 'Nutri
          tionEducation', 'Warmth', 'Care Hunger', 'SocialSciences', 'PerformingA
          rts', 'CharacterEducation', 'TeamSports', 'Other', 'College CareerPre
          p', 'Music', 'History Geography', 'Health LifeScience', 'EarlyDevelopme
          nt', 'ESL', 'Gym Fitness', 'EnvironmentalScience', 'VisualArts', 'Healt
          h Wellness', 'AppliedSciences', 'SpecialNeeds', 'Literature Writing',
          'Mathematics', 'Literacy']
          Shape of matrix after one hot encodig (109248, 30)
In [111]: school state vectorizer = CountVectorizer(lowercase=False, binary=True)
          school state vectorizer.fit(project data['school state'].values)
          print(school state vectorizer.get feature names())
          school state one hot = school state vectorizer.transform(project data[
          'school state'].values)
          print("Shape of matrix after one hot encodig ",school_state_one_hot.sha
          pe)
          print("the type of count vectorizer ",type(school state one hot))
          ['AK', 'AL', 'AR', 'AZ', 'CA', 'CO', 'CT', 'DC', 'DE', 'FL', 'GA', 'H
          I', 'IA', 'ID', 'IL', 'IN', 'KS', 'KY', 'LA', 'MA', 'MD', 'ME', 'MI',
          'MN', 'MO', 'MS', 'MT', 'NC', 'ND', 'NE', 'NH', 'NJ', 'NM', 'NV', 'NY',
          'OH', 'OK', 'OR', 'PA', 'RI', 'SC', 'SD', 'TN', 'TX', 'UT', 'VA', 'VT',
          'WA', 'WI', 'WV', 'WY']
          Shape of matrix after one hot encodig (109248, 51)
          the type of count vectorizer <class 'scipy.sparse.csr.csr matrix'>
In [112]: # https://www.geeksforgeeks.org/python-pandas-dataframe-fillna-to-repla
          ce-null-values-in-dataframe/
          project data["teacher prefix"].fillna("No Prefix", inplace = True)
          teacher prefix vectorizer = CountVectorizer(lowercase=False, binary=Tru
          e)
          teacher prefix vectorizer.fit(project data['teacher prefix'].values)
```

```
print(teacher prefix vectorizer.get feature names())
          teacher prefix one hot = teacher prefix vectorizer.transform(project da
          ta['teacher prefix'].values)
          print("Shape of matrix after one hot encodig ",teacher prefix one hot.s
          hape)
          ['Dr', 'Mr', 'Mrs', 'Ms', 'No Prefix', 'Teacher']
          Shape of matrix after one hot encodig (109248, 6)
In [113]: my grade counter = Counter()
          for project grade in project data['project grade category'].values:
              if (' ' in project grade):
                  project grade = project grade.replace(" ", "~")
              my grade counter.update(project grade.split())
          project grade cat dict = dict(my grade counter)
          sorted project grade cat dict = dict(sorted(project grade cat dict.item
          s(), key=lambda kv: kv[1]))
          grade cat vectorizer = CountVectorizer(vocabulary=list(sorted project q
          rade cat dict.keys()), lowercase=False, binary=True)
          grade cat vectorizer.fit(project data['project grade category'].values)
          print(grade cat vectorizer.get feature names())
          grade cat one hot = grade cat vectorizer.transform(project data['projec
          t grade category'].values)
          print("Shape of matrix after one hot encodig ", grade cat one hot.shape)
          ['Grades~9-12', 'Grades~6-8', 'Grades~3-5', 'Grades~PreK-2']
          Shape of matrix after one hot encodig (109248, 4)
```

1.5.2 Vectorizing Text data

1.5.2.1 Bag of words

```
In [114]: # We are considering only the words which appeared in at least 10 docum
          ents(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 (109248, 16623)
In [115]: titles vectorizer = CountVectorizer(min df=10)
          titles bow = titles vectorizer.fit transform(preprocessed titles)
          print("some sample features(unique words in the corpus)",titles vectori
          zer.get feature names()[0:10])
          print("Shape of matrix after one hot encodig ",titles bow.shape)
          print("the type of count vectorizer ", type(titles bow))
          print("the number of unique words ", titles bow.get shape()[1])
          some sample features(unique words in the corpus) ['abc', 'abcs', 'abili
          ties', 'ability', 'able', 'aboard', 'about', 'above', 'abstract', 'acad
          emic'l
          Shape of matrix after one hot encodig (109248, 3290)
          the type of count vectorizer <class 'scipy.sparse.csr.csr matrix'>
          the number of unique words 3290
          1.5.2.2 TFIDF vectorizer
In [116]: 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)
          Shape of matrix after one hot encodig (109248, 16623)
```

1.5.2.3 Using Pretrained Models: Avg W2V

```
In [117]:
          # Reading glove vectors in python: https://stackoverflow.com/a/3823034
          9/4084039
          def loadGloveModel(gloveFile):
              print ("Loading Glove Model")
              f = open(gloveFile, 'r', encoding="utf8")
              model = \{\}
              for line in tqdm(f):
                  splitLine = line.split()
                  word = splitLine[0]
                  embedding = np.array([float(val) for val in splitLine[1:]])
                  model[word] = embedding
              print ("Done.",len(model)," words loaded!")
              return model
          model = loadGloveModel('glove.42B.300d.txt')
          Output:
          Loading Glove Model
          1917495it [06:32, 4879.69it/s]
          Done. 1917495 words loaded!
          words = []
          for i in preproced texts:
              words.extend(i.split(' '))
          for i in preproced titles:
              words.extend(i.split(' '))
          print("all the words in the coupus", len(words))
          words = set(words)
          print("the unique words in the coupus", len(words))
          inter words = set(model.keys()).intersection(words)
          print("The number of words that are present in both glove vectors and o
```

Out[117]: '\n# Reading glove vectors in python: https://stackoverflow.com/a/38230 349/4084039\ndef loadGloveModel(gloveFile):\n print ("Loading Glove f = open(gloveFile,\'r\', encoding="utf8")\n Model")\n $model = \{\}$ for line in tqdm(f):\n splitLine = line.split()\n ord = splitLine[0]\n embedding = np.array([float(val) for val in splitLine[1:]]\n model[word] = embedding\n print ("Done.",le n(model)," words loaded!")\n return model\nmodel = loadGloveModel (\'qlove.42B.300d.txt\')\n\n# =============\n0utput:\n \nLoading Glove Model\n1917495it [06:32, 4879.69it/s]\nDone. 1917495 words loaded!\n\n# ===========\n\nwords = []\nfor i in preproced texts:\n words.extend(i.split(\' \'))\n\nfor i in preproce words.extend(i.split(\' \'))\nprint("all the words in th d titles:\n e coupus", len(words))\nwords = set(words)\nprint("the unique words in the coupus", len(words))\n\ninter words = set(model.keys()).intersectio n(words)\nprint("The number of words that are present in both glove vec tors and our coupus", len(inter words),"(",np.round(len(inter wor ds)/len(words)*100,3),"%)")\n\nwords courpus = {}\nwords glove = set(mo words c

```
ourpus[i] = model[i]\nprint("word 2 vec length", len(words courpus))\n
          \n\n# stronging variables into pickle files python: http://www.jessicay
          ung.com/how-to-use-pickle-to-save-and-load-variables-in-python/\n\nimpo
          rt pickle\nwith open(\'glove vectors\', \'wb\') as f:\n
                                                                     pickle.dump
          (words courpus, f)\n\n\n'
In [118]: # stronging variables into pickle files python: http://www.jessicayung.
          com/how-to-use-pickle-to-save-and-load-variables-in-python/
          # 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 [119]: # 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/re
          view
              for word in sentence.split(): # for each word in a review/sentence
                  if word in glove words:
                      vector += model[word]
                      cnt words += 1
              if cnt words != 0:
                  vector /= cnt words
              avg w2v vectors.append(vector)
          print(len(avg w2v vectors))
          print(len(avg w2v vectors[0]))
          100%
                  109248/109248 [01:02<00:00, 1736.72it/s]
```

300

109248

1.5.2.3 Using Pretrained Models: TFIDF weighted W2V

```
In [120]: \# S = ["abc \ def \ pgr", "def \ def \ def \ abc", "pgr \ pgr \ def"]
          tfidf model = TfidfVectorizer()
          tfidf model.fit(preprocessed essays)
          # we are converting a dictionary with word as a key, and the idf as a v
          alue
          dictionary = dict(zip(tfidf model.get feature names(), list(tfidf model
           .idf )))
          tfidf words = set(tfidf model.get feature names())
In [121]: # average Word2Vec
          # compute average word2vec for each review.
          tfidf w2v vectors = []; # the avg-w2v for each sentence/review is store
          d 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
          e/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 t
          he tf value((sentence.count(word)/len(sentence.split())))
                      tf idf = dictionary[word]*(sentence.count(word)/len(sentenc
          e.split())) # getting the tfidf value for each word
                      vector += (vec * tf idf) # calculating tfidf weighted w2v
                      tf idf weight += tf idf
              if tf idf weight != 0:
                  vector /= tf idf weight
              tfidf w2v vectors.append(vector)
          print(len(tfidf w2v vectors))
          print(len(tfidf w2v vectors[0]))
          100%
                   109248/109248 [08:18<00:00, 219.13it/s]
          109248
          300
```

In [122]: # TFIDF on project titles titles tfidf vectorizer = TfidfVectorizer(min df=10) titles tfidf = titles tfidf vectorizer.fit transform(preprocessed title print("some sample features(unique words in the corpus)", titles tfidf v ectorizer.get feature names()[10:21]) print("Shape of matrix after one hot encodig ",titles tfidf.shape) some sample features(unique words in the corpus) ['academics', 'academ y', 'acceptance', 'access', 'accessibility', 'accessible', 'accessing', 'accessories', 'ace', 'achieve', 'achievement'] Shape of matrix after one hot encodig (109248, 3290) In [123]: # AVG W2V on project title avg w2v titles vectors = []; for sentence in tqdm(preprocessed titles): vector titles = np.zeros(300) cnt words titles = 0; for word in sentence.split(): if word in glove words: vector += model[word] cnt words titles += 1 if cnt words titles != 0: vector titles /= cnt words titles avg w2v titles vectors.append(vector titles) print(len(avg w2v titles vectors)) print(len(avg w2v titles vectors[0]))

100%|

```
109248/109248 [00:02<00:00, 40270.97it/s]
          109248
          300
In [124]: # TFIDF weighted W2V on project_title
          titles tfidf model = TfidfVectorizer()
          titles tfidf model.fit(preprocessed titles)
          titles dictionary = dict(zip(titles tfidf model.get feature names(), li
          st(titles tfidf model.idf )))
          titles tfidf words = set(titles tfidf model.get feature names())
In [125]: titles tfidf w2v vectors = [];
          for titles sentence in tqdm(preprocessed_titles):
              titles vector = np.zeros(300)
              titles tfidf weight = 0;
              for word in titles sentence.split():
                  if (word in glove words) and (word in titles tfidf words):
                      titles vec = model[word]
                      titles tf idf = titles dictionary[word]*(titles sentence.co
          unt(word)/len(titles sentence.split()))
                      titles vector += (titles vec * titles tf idf)
                      titles tfidf weight += titles tf idf
              if titles tfidf weight != 0:
                  titles vector /= titles tfidf weight
              titles tfidf w2v vectors.append(titles vector)
          print(len(titles tfidf w2v vectors))
          print(len(titles tfidf w2v vectors[0]))
```

```
100%|
      109248/109248 [00:06<00:00, 15934.72it/s]
109248
300
```

1.5.3 Vectorizing Numerical features

```
price data = resource data.groupby('id').agg({'price':'sum', 'quantity'
In [126]:
          :'sum'}).reset index()
          project data = pd.merge(project data, price data, on='id', how='left')
In [127]: # check this one: https://www.youtube.com/watch?v=0H0g0cln3Z4&t=530s
          # standardization sklearn: https://scikit-learn.org/stable/modules/gene
          rated/sklearn.preprocessing.StandardScaler.html
          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 21
          3.03 329. ... 399. 287.73 5.5 1.
          # 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 data
          print(f"Mean : {price scalar.mean [0]}, Standard deviation : {np.sqrt(p
          rice scalar.var [0])}")
          # Now standardize the data with above maen and variance.
          price standardized = price scalar.transform(project data['price'].value
          s.reshape(-1, 1))
          Mean: 298.1193425966608, Standard deviation: 367.49634838483496
In [128]: price standardized
Out[128]: array([[-0.3905327],
```

```
[ 0.00239637],
[ 0.59519138],
...,
[-0.15825829],
[-0.61243967],
[-0.51216657]])
```

1.5.4 Merging all the above features

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

```
In [129]: print(categories one hot.shape)
          print(sub categories one hot.shape)
          print(text bow.shape)
          print(price standardized.shape)
          (109248, 9)
          (109248, 30)
          (109248, 16623)
          (109248, 1)
In [130]: # merge two sparse matrices: https://stackoverflow.com/a/19710648/40840
          39
          from scipy.sparse import hstack
          # with the same hstack function we are concatinating a sparse matrix an
          d a dense matirx :)
          X = hstack((categories one hot, sub categories one hot, text bow, price
           standardized))
          X.shape
Out[130]: (109248, 16663)
          Computing Sentiment Scores
In [131]: import nltk
          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 te
ach the smallest students with the biggest enthusiasm \
for learning my students learn in many different ways using all of our
senses and multiple intelligences i use a wide range\
of techniques to help all my students succeed students in my class come
from a variety of different backgrounds which makes\
for wonderful sharing of experiences and cultures including native amer
icans our school is a caring community of successful \
learners which can be seen through collaborative student project based
learning in and out of the classroom kindergarteners \
in my class love to work with hands on materials and have many differen
t opportunities to practice a skill before it is\
mastered having the social skills to work cooperatively with friends is
a crucial aspect of the kindergarten curriculum\
montana is the perfect place to learn about agriculture and nutrition m
y students love to role play in our pretend kitchen\
in the early childhood classroom i have had several kids ask me can we
try cooking with real food i will take their idea \
and create common core cooking lessons where we learn important math an
d writing concepts while cooking delicious healthy \
food for snack time my students will have a grounded appreciation for t
he work that went into making the food and knowledge \
of where the ingredients came from as well as how it is healthy for the
ir bodies this project would expand our learning of \
nutrition and agricultural cooking recipes by having us peel our own ap
ples to make homemade applesauce make our own bread \
and mix up healthy plants from our classroom garden in the spring we wi
ll also create our own cookbooks to be printed and \
shared with families students will gain math and literature skills as w
ell as a life long enjoyment for healthy cooking \
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)
# neg: 0.0, neu: 0.753, pos: 0.247, compound: 0.93
```

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

Assignment 5: Logistic Regression

- [Task-1] Logistic Regression(either SGDClassifier with log loss, or LogisticRegression) on these feature sets
 - Set 1: categorical, numerical features + project_title(BOW) + preprocessed_eassay (`BOW with bi-grams` with `min_df=10` and `max_features=5000`)
 - Set 2: categorical, numerical features + project_title(TFIDF)+ preprocessed_eassay (`TFIDF with bi-grams` with `min_df=10` and `max_features=5000`)
 - Set 3: categorical, numerical features + project_title(AVG W2V)+ preprocessed_eassay (AVG W2V)
 - Set 4: categorical, numerical features + project_title(TFIDF W2V)+ preprocessed_essay (TFIDF W2V)
- 2. Hyper paramter tuning (find best hyper parameters corresponding the algorithm that you choose)
 - Find the best hyper parameter which will give the maximum <u>AUC</u> value
 - 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. 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 confusion matrix with predicted and original labels of test data points. Please visualize your confusion matrices using seaborn heatmaps.



- 4. [Task-2] Apply Logistic Regression on the below feature set Set 5 by finding the best hyper parameter as suggested in step 2 and step 3.
- 5. Consider these set of features Set 5:
 - school state: categorical data
 - clean categories: categorical data
 - clean_subcategories : categorical data
 - project grade category :categorical data
 - teacher prefix : categorical data
 - quantity : numerical data
 - teacher number of previously posted projects : numerical data
 - price : numerical data
 - sentiment score's of each of the essay : numerical data
 - number of words in the title : numerical data
 - number of words in the combine essays : numerical data

And apply the Logistic regression on these features by finding the best hyper paramter as suggested 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



Note: Data Leakage

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

2. Logistic Regression

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

```
In [132]: approved_project = project_data['project_is_approved'].values
    project_data.drop(['project_is_approved'], axis=1, inplace=True)
    project_data.head(1)
```

Out[132]:

| | Unnamed: 0 | id | teacher_id | teacher_prefix | school_state | ķ |
|---|---------------|---------|----------------------------------|----------------|--------------|---|
| 0 | 160221 | p253737 | c90749f5d961ff158d4b4d1e7dc665fc | Mrs. | IN | 2 |

```
In [246]: # Data splitting
    from sklearn.model_selection import train_test_split

# Splitting in train and test
X_train, X_test, y_train, y_test = train_test_split(project_data, approved_project, test_size=0.33, stratify=approved_project)

# Splitting in Train Test and Cross Validation
X_train, X_cv, y_train, y_cv = train_test_split(X_train, y_train, test_size=0.33, stratify=y_train)
```

2.2 Make Data Model Ready: encoding numerical, categorical features

```
In [134]: # Vectorizing Categories on Train, Test and CV data
          from sklearn.feature extraction.text import CountVectorizer
          vectorizer = CountVectorizer(vocabulary=list(sorted cat dict.keys()), l
          owercase=False, binary=True)
          # Fit only to train data
          vectorizer.fit(X train['clean categories'].values)
          # Transform to train, test and CV data
          X Train categories one hot = vectorizer.transform(X train['clean catego
          ries'l.values)
          X Test categories one hot = vectorizer.transform(X test['clean categori
          es'l.values)
          X CV categories one hot = vectorizer.transform(X cv['clean categories']
          .values)
          print("Shape of train matrix after one hot encodig ",X Train categories
          one hot.shape)
          print("Shape of test matrix after one hot encodig ",X Test categories o
          ne hot.shape)
```

```
print("Shape of cv matrix after one hot encodig ", X CV categories one h
          ot.shape)
          Shape of train matrix after one hot encodig (49041, 9)
          Shape of test matrix after one hot encodig (36052, 9)
          Shape of cv matrix after one hot encodig (24155, 9)
In [135]: # Vectorizing subcategories on train, test and cv
          vectorizer = CountVectorizer(vocabulary=list(sorted sub cat dict.keys
          ()), lowercase=False, binary=True)
          vectorizer.fit(X train['clean subcategories'].values)
          X Train sub categories one hot = vectorizer.transform(X train['clean su
          bcategories'].values)
          X Test sub categories one hot = vectorizer.transform(X test['clean subc
          ategories'l.values)
          X CV sub categories one hot = vectorizer.transform(X_cv['clean_subcateg
          ories'l.values)
          print("Shape of train matrix after one hot encodig ",X Train sub catego
          ries one hot.shape)
          print("Shape of test matrix after one hot encodig ",X Test sub categori
          es one hot.shape)
          print("Shape of cv matrix after one hot encodig ",X CV sub categories o
          ne hot.shape)
          Shape of train matrix after one hot encodig (49041, 30)
          Shape of test matrix after one hot encodig (36052, 30)
          Shape of cv matrix after one hot encodig (24155, 30)
In [136]: # Vectorizing school state on train , test and cv
          school state vectorizer = CountVectorizer(lowercase=False, binary=True)
          school state vectorizer.fit(X train['school state'].values)
          print(school state vectorizer.get feature names())
          X Train school state one hot = school state vectorizer.transform(X trai
```

```
n['school state'].values)
          X Test school state one hot = school state vectorizer.transform(X test[
          'school state'].values)
          X CV school state one hot = school state vectorizer.transform(X cv['sch
          ool state'l.values)
          print("Shape of train matrix after one hot encodig ",X Train school sta
          te one hot shape)
          print("Shape of test matrix after one hot encodig ",X Test school state
          one hot.shape)
          print("Shape of cv matrix after one hot encodig ", X CV school state one
          hot.shape)
          print("the type of count vectorizer ",type(X Train school state one hot
          ))
          print("the type of count vectorizer ",type(X Test school state one hot
          print("the type of count vectorizer ", type(X CV school state one hot))
          ['AK', 'AL', 'AR', 'AZ', 'CA', 'CO', 'CT', 'DC', 'DE', 'FL', 'GA', 'H
          I', 'IA', 'ID', 'IL', 'IN', 'KS', 'KY', 'LA', 'MA', 'MD', 'ME', 'MI',
          'MN', 'MO', 'MS', 'MT', 'NC', 'ND', 'NE', 'NH', 'NJ', 'NM', 'NV', 'NY',
          'OH', 'OK', 'OR', 'PA', 'RI', 'SC', 'SD', 'TN', 'TX', 'UT', 'VA', 'VT',
          'WA', 'WI', 'WV', 'WY']
          Shape of train matrix after one hot encodig (49041, 51)
          Shape of test matrix after one hot encodig (36052, 51)
          Shape of cv matrix after one hot encodig (24155, 51)
          the type of count vectorizer <class 'scipy.sparse.csr.csr matrix'>
          the type of count vectorizer <class 'scipy.sparse.csr.csr matrix'>
          the type of count vectorizer <class 'scipy.sparse.csr.csr matrix'>
In [137]: # Vectorizing teacher prefix on train , test and cv
          project data["teacher prefix"].fillna("No Prefix", inplace = True)
          teacher prefix vectorizer = CountVectorizer(lowercase=False, binary=Tru
          e)
          teacher prefix vectorizer.fit(X train['teacher prefix'].values)
```

```
print(teacher prefix vectorizer.get feature names())
          X Train teacher prefix one hot = teacher prefix vectorizer.transform(X
          train['teacher prefix'].values)
          X Test teacher prefix_one_hot = teacher_prefix_vectorizer.transform(X_t
          est['teacher prefix'].values)
          X CV teacher prefix one hot = teacher prefix vectorizer.transform(X cv[
          'teacher prefix'].values)
          print("Shape of train matrix after one hot encodig ",X Train teacher pr
          efix one hot.shape)
          print("Shape of test matrix after one hot encodig ",X Test teacher pref
          ix one hot.shape)
          print("Shape of cv matrix after one hot encodig ",X CV teacher prefix o
          ne hot.shape)
          ['Dr', 'Mr', 'Mrs', 'Ms', 'No Prefix', 'Teacher']
          Shape of train matrix after one hot encodig (49041, 6)
          Shape of test matrix after one hot encodig (36052, 6)
          Shape of cv matrix after one hot encodig (24155, 6)
In [138]: # Vectorizing grade category on train , test and cv
          my grade counter = Counter()
          for project grade in project data['project grade category'].values:
              if (' ' in project grade):
                  project grade = project grade.replace(" ", "~")
              my grade counter.update(project grade.split())
          project grade cat dict = dict(my grade counter)
          sorted project grade cat dict = dict(sorted(project grade cat dict.item
          s(), key=lambda kv: kv[1]))
          grade cat vectorizer = CountVectorizer(vocabulary=list(sorted project g
          rade cat dict.keys()), lowercase=False, binary=True)
```

```
grade_cat_vectorizer.fit(X_train['project_grade_category'].values)
print(grade_cat_vectorizer.get_feature_names())

X_Train_grade_cat_one_hot = grade_cat_vectorizer.transform(X_train['project_grade_category'].values)

X_Test_grade_cat_one_hot = grade_cat_vectorizer.transform(X_test['project_grade_category'].values)

X_CV_grade_cat_one_hot = grade_cat_vectorizer.transform(X_cv['project_grade_category'].values)

print("Shape of train matrix after one hot encodig ",X_Train_grade_cat_one_hot.shape)
print("Shape of test matrix after one hot encodig ",X_Test_grade_cat_one_hot.shape)
print("Shape of cv matrix after one hot encodig ",X_CV_grade_cat_one_hot.shape)
```

['Grades~9-12', 'Grades~6-8', 'Grades~3-5', 'Grades~PreK-2'] Shape of train matrix after one hot encodig (49041, 4) Shape of test matrix after one hot encodig (36052, 4) Shape of cv matrix after one hot encodig (24155, 4)

2.3 Make Data Model Ready: encoding eassay, and project title

```
X Train essay sent = X Train essay sent.replace('\\r', ' '
              X Train essay sent = X Train essay sent.replace('\\"', ' ')
              X Train essay sent = X Train essay sent.replace('\\n', ' ')
              X Train essay sent = re.sub('[^A-Za-z0-9]+', ' ', X_Train_essay_sen
          t)
              X_Train_essay_sent = ' '.join(e for e in X_Train_essay_sent.split()
           if e.lower() not in stopwords)
              X Train preprocessed essays.append(X Train essay sent.lower().strip
          ())
          100%
                     49041/49041 [01:00<00:00, 804.00it/s]
In [141]: # preprocessing essay test data
          from tqdm import tqdm
          X Test preprocessed essays = []
          # tqdm is for printing the status bar
          for sentence in tqdm(X test['essay'].values):
              X Test essay sent = decontracted(sentence)
              X Test essay sent = X Test essay sent.replace('\\r', ' ')
              X_Test_essay_sent = X_Test_essay_sent.replace('\\"', ' ')
              X Test essay sent = X Test essay sent.replace('\\n', ' ')
              X Test essay sent = re.sub('[^A-Za-z0-9]+', ' ', X Test essay sent)
              X Test essay sent = ' '.join(e for e in X Test essay sent.split() i
          f e.lower() not in stopwords)
              X Test preprocessed essays.append(X Test essay sent.lower().strip
          ())
          100%
                     36052/36052 [00:44<00:00, 809.19it/s]
In [142]: # preprocessing essay cv data
          from tadm import tadm
          X CV preprocessed essays = []
          # tqdm is for printing the status bar
          for sentence in tqdm(X cv['essay'].values):
              X CV essay sent = decontracted(sentence)
```

```
X CV essay sent = X CV essay sent.replace('\\r', ' ')
              X_CV_essay_sent = X_CV essay sent.replace('\\"', ' ')
              X CV essay sent = X CV essay sent.replace('\\n', ' ')
              X = CV = say sent = re.sub('[^A-Za-z0-9]+', ' ', X_CV_essay_sent)
              X CV essay sent = ' '.join(e for e in X CV essay sent.split() if e.
          lower() not in stopwords)
              X CV preprocessed essays.append(X CV essay sent.lower().strip())
          100%|
                     24155/24155 [00:29<00:00, 826.27it/s]
In [239]: normalizer = Normalizer()
          normalizer.fit(X train['essay word count'].values.reshape(-1,1))
          essay word count train = normalizer.transform(X train['essay word coun
          t'l.values.reshape(-1,1))
          essay word count cv = normalizer.transform(X cv['essay word count'].val
          ues.reshape(-1,1))
          essay word count test = normalizer.transform(X test['essay word count']
           .values.reshape(-1,1))
          print("After vectorizations")
          print(essay word count train.shape, y train.shape)
          print(essay word count cv.shape, y cv.shape)
          print(essay word count test.shape, y test.shape)
          print("="*100)
          After vectorizations
          (49041, 1) (49041,)
          (24155, 1) (24155,)
          (36052, 1) (36052,)
In [247]: normalizer = Normalizer()
          normalizer.fit(X_train['pos'].values.reshape(-1,1))
```

```
essay sent pos train = normalizer.transform(X train['pos'].values.resha
pe(-1,1))
essay sent pos cv = normalizer.transform(X cv['pos'].values.reshape(-1,
1))
essay sent pos test = normalizer.transform(X test['pos'].values.reshape
(-1,1)
print("After vectorizations")
print(essay sent pos train.shape, y train.shape)
print(essay sent pos cv.shape, y cv.shape)
print(essay sent pos test.shape, y test.shape)
print("="*100)
normalizer.fit(X train['neg'].values.reshape(-1,1))
essay sent neg train = normalizer.transform(X train['neg'].values.resha
pe(-1,1))
essay sent neg cv = normalizer.transform(X cv['neg'].values.reshape(-1,
1))
essay sent neg test = normalizer.transform(X test['neg'].values.reshape
(-1,1)
print("After vectorizations")
print(essay sent neg train.shape, y train.shape)
print(essay sent neg cv.shape, y cv.shape)
print(essay sent neg test.shape, y test.shape)
print("="*100)
normalizer.fit(X train['neu'].values.reshape(-1,1))
essay sent neu train = normalizer.transform(X train['neu'].values.resha
pe(-1,1))
essay sent neu cv = normalizer.transform(X cv['neu'].values.reshape(-1,
essay sent neu test = normalizer.transform(X test['neu'].values.reshape
(-1,1)
print("After vectorizations")
```

```
print(essay sent neu train.shape, y train.shape)
print(essay sent neu cv.shape, y cv.shape)
print(essay sent neu test.shape, y test.shape)
print("="*100)
normalizer.fit(X train['compound'].values.reshape(-1,1))
essay sent comp train = normalizer.transform(X train['compound'].values
.reshape(-1,1))
essay sent comp cv = normalizer.transform(X cv['compound'].values.resha
pe(-1,1))
essay sent comp test = normalizer.transform(X test['compound'].values.r
eshape(-1,1))
print("After vectorizations")
print(essay sent comp train.shape, y train.shape)
print(essay sent comp cv.shape, y cv.shape)
print(essay sent comp test.shape, y test.shape)
print("="*100)
After vectorizations
(49041, 1) (49041,)
(24155, 1) (24155,)
(36052. 1) (36052.)
After vectorizations
(49041, 1) (49041,)
(24155, 1) (24155,)
(36052, 1) (36052,)
After vectorizations
(49041, 1) (49041,)
(24155, 1) (24155,)
(36052, 1) (36052,)
After vectorizations
(49041, 1) (49041,)
```

```
(24155, 1) (24155,)
           (36052, 1) (36052,)
In [143]: # preprocessing project title train data
          X Train preprocessed titles = []
           for dataset in tqdm(X train['project title'].values):
               data = decontracted(dataset) # Replacing some specific and general
           short form into proper word/stopword.
               data = re.sub(r"it's", "it is", data) # Replacing it's with it is a
           s it is not part of function decontracted
               data = data.replace('\\r', ' ') # Replacing \r with space
               data = data.replace('\\"', ' ') # Replacing \ with space
data = data.replace('\\n', ' ') # Replacing \ n with space
               data = re.sub('[^A-Za-z0-9]+', '', data) # Replacing special chara
           cters with space
               data = re.sub("\S*\d\S*", "", data).strip() # Trimming numbers cont
           aining digits
               data = ' '.join(e for e in data.split() if e not in stopwords) # Re
          moving stopwords
               X Train preprocessed titles.append(data.lower().strip()) # Creating
           array in all the lower cases.
           100%
                    49041/49041 [00:03<00:00, 15175.54it/s]
In [144]: # preprocessing project title test data
          X Test preprocessed titles = []
           for dataset in tgdm(X test['project title'].values):
               data = decontracted(dataset) # Replacing some specific and general
           short form into proper word/stopword.
               data = re.sub(r"it's", "it is", data) # Replacing it's with it is a
           s it is not part of function decontracted
               data = data.replace('\\r', ' ') # Replacing \r with space
               data = data.replace('\\"', ' ') # Replacing \ with space
```

```
data = data.replace('\\n', ' ') # Replacing \n with space
              data = re.sub('[^A-Za-z0-9]+', ' ', data) # Replacing special chara
          cters with space
              data = re.sub("\S*\d\S*", "", data).strip() # Trimming numbers cont
          aining digits
              data = ' '.join(e for e in data.split() if e not in stopwords) # Re
          movina stopwords
              X Test preprocessed titles.append(data.lower().strip()) # Creating
           array in all the lower cases.
          100%|
                   36052/36052 [00:02<00:00, 17036.39it/s]
In [145]: # preprocessing project title cv data
          X CV preprocessed titles = []
          for dataset in tqdm(X cv['project title'].values):
              data = decontracted(dataset) # Replacing some specific and general
           short form into proper word/stopword.
              data = re.sub(r"it's", "it is", data) # Replacing it's with it is a
          s it is not part of function decontracted
              data = data.replace('\\r', ' ') # Replacing \r with space
              data = data.replace('\\"', ' ') # Replacing \ with space
              data = data.replace('\\n', ' ') # Replacing \n with space
              data = re.sub('[^A-Za-z0-9]+', ' ', data) # Replacing special chara
          cters with space
              data = re.sub("\S*\d\S*", "", data).strip() # Trimming numbers cont
          aining digits
              data = ' '.join(e for e in data.split() if e not in stopwords) # Re
          moving stopwords
              X CV preprocessed titles.append(data.lower().strip()) # Creating ar
          ray in all the lower cases.
          100%|
                  24155/24155 [00:01<00:00, 16979.16it/s]
In [240]: normalizer = Normalizer()
```

```
normalizer.fit(X train['title word count'].values.reshape(-1,1))
          title word count train = normalizer.transform(X train['title word coun
          t'l.values.reshape(-1,1))
          title word count cv = normalizer.transform(X cv['title word count'].val
          ues.reshape(-1,1))
          title word count test = normalizer.transform(X test['title word count']
           .values.reshape(-1,1))
          print("After vectorizations")
          print(title word count train.shape, y train.shape)
          print(title word count cv.shape, y cv.shape)
          print(title word count test.shape, y test.shape)
          print("="*100)
          After vectorizations
          (49041, 1) (49041,)
          (24155, 1) (24155,)
          (36052, 1) (36052,)
In [146]: # BOW Essay train, test and cv data
          # We are considering only the words which appeared in at least 10 docum
          ents(rows or projects).
          vectorizer = CountVectorizer(ngram range=(2,2), min df=10, max features
           = 5000)
          vectorizer.fit(X Train preprocessed essays)
          X Train essay bow = vectorizer.transform(X Train preprocessed essays)
          X Test essay bow = vectorizer.transform(X Test preprocessed essays)
          X CV essay bow = vectorizer.transform(X CV preprocessed essays)
          print("Shape of train matrix after one hot encodig ",X Train essay bow.
          shape)
          print("Shape of test matrix after one hot encodig ",X Test essay bow.sh
          ape)
          print("Shape of CV matrix after one hot encodig ",X CV essay bow.shape)
```

```
Shape of test matrix after one hot encodig (36052, 147)
          Shape of CV matrix after one hot encodig (24155, 147)
In [147]: # BOW title train, test and cv data
          titles vectorizer = CountVectorizer(ngram range=(2,2), min df=10, max f
          eatures = 5000)
          titles vectorizer.fit(X Train preprocessed titles)
          X Train titles bow = titles vectorizer.transform(X Train preprocessed t
          itles)
          X Test titles bow = titles vectorizer.transform(X Test preprocessed tit
          les)
          X CV titles bow = titles vectorizer.transform(X CV preprocessed titles)
          print("some sample features(unique words in the corpus)",titles vectori
          zer.get feature names()[0:10])
          print("Shape of train matrix after one hot encodig ",X Train titles bow
          print("Shape of test matrix after one hot encodig ",X Test titles bow.s
          hape)
          print("Shape of CV matrix after one hot encodig ",X CV titles bow.shape
          some sample features(unique words in the corpus) ['about it', 'about ou
          r', 'about reading', 'about science', 'about that', 'about the', 'about
          world', 'academic success', 'access technology', 'active bodies']
          Shape of train matrix after one hot encodig (49041, 1659)
          Shape of test matrix after one hot encodig (36052, 1659)
          Shape of CV matrix after one hot encodig (24155, 1659)
In [148]: #TFIDF essay train, test and cv data
          vectorizer = TfidfVectorizer(ngram range=(2,2), min df=10, max features
           = 5000)
          vectorizer.fit(X Train preprocessed essays)
          X Train essay tfidf = vectorizer.transform(X Train preprocessed essays)
```

Shape of train matrix after one hot encodig (49041, 147)

```
X Test essay tfidf = vectorizer.transform(X Test preprocessed essays)
          X CV essay tfidf = vectorizer.transform(X CV preprocessed essays)
          print("Shape of train matrix after one hot encodig ",X Train essay tfid
          f.shape)
          print("Shape of test matrix after one hot encodig ",X Test essay tfidf.
          shape)
          print("Shape of CV matrix after one hot encodig ",X CV essay tfidf.shap
          Shape of train matrix after one hot encodig (49041, 147)
          Shape of test matrix after one hot encodig (36052, 147)
          Shape of CV matrix after one hot encodig (24155, 147)
In [149]: # TFIDF on project titles train, test and cv data
          titles tfidf vectorizer = TfidfVectorizer(ngram range=(2,2), min df=10,
           max features = 5000)
          titles tfidf vectorizer.fit(X Train_preprocessed_titles)
          X Train titles tfidf = titles vectorizer.transform(X Train preprocessed
          titles)
          X Test titles tfidf = titles vectorizer.transform(X Test preprocessed t
          itles)
          X CV titles tfidf = titles vectorizer.transform(X CV preprocessed title
          s)
          print("Shape of train matrix after one hot encodig ",X Train titles tfi
          df.shape)
          print("Shape of test matrix after one hot encodig ",X Test titles tfidf
          .shape)
          print("Shape of CV matrix after one hot encodig ",X CV titles tfidf.sha
          pe)
          Shape of train matrix after one hot encodig (49041, 1659)
          Shape of test matrix after one hot encodig (36052, 1659)
          Shape of CV matrix after one hot encodig (24155, 1659)
In [150]: # average Word2Vec essay on train
```

```
# compute average word2vec for each review.
          X Train avg w2v vectors = []; # the avg-w2v for each sentence/review is
           stored in this list
          for sentence in tqdm(X Train preprocessed essays): # for each review/se
          ntence
              vector = np.zeros(300) # as word vectors are of zero length
              cnt words =0; # num of words with a valid vector in the sentence/re
          view
              for word in sentence.split(): # for each word in a review/sentence
                  if word in glove words:
                      vector += model[word]
                      cnt words += 1
              if cnt words != 0:
                  vector /= cnt words
              X Train avg w2v vectors.append(vector)
          print(len(X Train avg w2v vectors))
          print(len(X Train avg w2v vectors[0]))
          100%
                    49041/49041 [00:25<00:00, 1933.71it/s]
          49041
          300
In [151]: # average Word2Vec essay on test
          # compute average word2vec for each review.
          X Test avg w2v vectors = []; # the avg-w2v for each sentence/review is
           stored in this list
          for sentence in tgdm(X Test preprocessed essays): # for each review/sen
          tence
              vector = np.zeros(300) # as word vectors are of zero length
              cnt words =0; # num of words with a valid vector in the sentence/re
          view
              for word in sentence.split(): # for each word in a review/sentence
                  if word in glove words:
                      vector += model[word]
                      cnt words += 1
              if cnt words != 0:
                  vector /= cnt words
```

```
X Test avg w2v vectors.append(vector)
          print(len(X Test avg w2v vectors))
          print(len(X Test avg w2v vectors[0]))
          100%|
                    36052/36052 [00:19<00:00, 1823.47it/s]
          36052
          300
In [152]: # average Word2Vec essay on cv
          # compute average word2vec for each review.
          X CV avg w2v vectors = []; # the avg-w2v for each sentence/review is st
          ored in this list
          for sentence in tqdm(X CV preprocessed essays): # for each review/sente
          nce
              vector = np.zeros(300) # as word vectors are of zero length
              cnt words =0; # num of words with a valid vector in the sentence/re
          view
              for word in sentence.split(): # for each word in a review/sentence
                  if word in glove words:
                      vector += model[word]
                      cnt words += 1
              if cnt words != 0:
                  vector /= cnt words
              X CV avg w2v vectors.append(vector)
          print(len(X CV avg w2v vectors))
          print(len(X CV avg w2v vectors[0]))
          100%
                   | 24155/24155 [00:13<00:00, 1854.35it/s]
          24155
          300
In [153]: # AVG W2V on project title train
          X Train avg w2v titles vectors = [];
```

```
for sentence in tqdm(X_Train_preprocessed_titles):
              vector titles = np.zeros(300)
              cnt_words_titles = 0;
              for word in sentence.split():
                  if word in glove words:
                      vector += model[word]
                      cnt words titles += 1
              if cnt words titles != 0:
                  vector titles /= cnt words titles
              X Train avg w2v titles vectors.append(vector titles)
          print(len(X Train avg w2v titles vectors))
          print(len(X Train avg w2v titles vectors[0]))
          100%|
                   49041/49041 [00:01<00:00, 33126.41it/s]
          49041
          300
In [154]: # AVG W2V on project title test
          X Test avg w2v titles vectors = [];
          for sentence in tqdm(X Test preprocessed titles):
              vector titles = np.zeros(300)
              cnt words titles = 0;
              for word in sentence.split():
                  if word in glove words:
                      vector += model[word]
```

```
cnt words titles += 1
              if cnt words titles != 0:
                  vector titles /= cnt words titles
              X_Test_avg_w2v_titles_vectors.append(vector_titles)
          print(len(X Test avg w2v titles vectors))
          print(len(X Test avg w2v titles vectors[0]))
          100%|
                   36052/36052 [00:01<00:00, 32520.73it/s]
          36052
          300
In [155]: # AVG W2V on project title cv
          X_CV_avg_w2v_titles_vectors = [];
          for sentence in tqdm(X CV preprocessed titles):
              vector titles = np.zeros(300)
              cnt words titles = 0;
              for word in sentence.split():
                  if word in glove words:
                      vector += model[word]
                      cnt words titles += 1
              if cnt words titles != 0:
                  vector_titles /= cnt_words_titles
              X CV avg w2v titles vectors.append(vector titles)
          print(len(X CV avg w2v titles vectors))
          print(len(X CV avg w2v titles vectors[0]))
```

```
100%|
                   24155/24155 [00:00<00:00, 36192.76it/s]
          24155
          300
In [156]: # TFIDF W2V
          # S = ["abc def pgr", "def def def abc", "pgr pgr def"]
          tfidf model = TfidfVectorizer()
          tfidf model.fit(X Train preprocessed essays)
          # we are converting a dictionary with word as a key, and the idf as a v
          alue
          dictionary = dict(zip(tfidf model.get feature names(), list(tfidf model
          .idf )))
          tfidf words = set(tfidf model.get feature names())
In [157]: # TFIDF w2v essay train
          # compute average word2vec for each review.
          X Train tfidf w2v vectors = []; # the avg-w2v for each sentence/review
           is stored in this list
          for sentence in tqdm(X Train preprocessed essays): # for each review/se
          ntence
              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
          e/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 t
          he tf value((sentence.count(word)/len(sentence.split())))
                      tf idf = dictionary[word]*(sentence.count(word)/len(sentenc
          e.split())) # getting the tfidf value for each word
                      vector += (vec * tf idf) # calculating tfidf weighted w2v
                      tf idf weight += tf idf
              if tf idf weight != 0:
                  vector /= tf idf weight
              X Train tfidf w2v vectors.append(vector)
```

```
print(len(X Train tfidf w2v vectors))
          print(len(X_Train tfidf w2v vectors[0]))
          100%|
                     49041/49041 [03:31<00:00, 232.11it/s]
          49041
          300
In [158]: # TFIDF w2v essav test
          # compute average word2vec for each review.
          X Test tfidf w2v vectors = []; # the avg-w2v for each sentence/review i
          s stored in this list
          for sentence in tgdm(X Test preprocessed essays): # for each review/sen
          tence
              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
          e/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 t
          he tf value((sentence.count(word)/len(sentence.split())))
                      tf idf = dictionary[word]*(sentence.count(word)/len(sentenc
          e.split())) # getting the tfidf value for each word
                      vector += (vec * tf idf) # calculating tfidf weighted w2v
                      tf idf weight += tf idf
              if tf idf weight != 0:
                  vector /= tf idf weight
              X Test tfidf w2v vectors.append(vector)
          print(len(X Test tfidf w2v vectors))
          print(len(X Test tfidf w2v vectors[0]))
          100%|
                    36052/36052 [00:13<00:00, 2682.08it/s]
          36052
          300
```

```
In [159]: # TFIDF w2v essay cv
          # compute average word2vec for each review.
          X CV tfidf w2v vectors = []; # the avg-w2v for each sentence/review is
           stored in this list
          for sentence in tqdm(X CV preprocessed essays): # for each review/sente
          nce
              vector = np.zeros(300) # as word vectors are of zero length
              tf idf weight =0: # num of words with a valid vector in the sentenc
          e/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 t
          he tf value((sentence.count(word)/len(sentence.split())))
                      tf idf = dictionary[word]*(sentence.count(word)/len(sentenc
          e.split())) # getting the tfidf value for each word
                      vector += (vec * tf idf) # calculating tfidf weighted w2v
                      tf idf weight += tf idf
              if tf idf weight != 0:
                  vector /= tf idf weight
              X CV tfidf w2v vectors.append(vector)
          print(len(X CV tfidf w2v vectors))
          print(len(X CV tfidf w2v vectors[0]))
          100%|
                  | 24155/24155 [00:08<00:00, 2764.84it/s]
          24155
          300
In [160]: # TFIDF weighted W2V on project title
          titles tfidf model = TfidfVectorizer()
          titles tfidf model.fit(X Train preprocessed titles)
          titles dictionary = dict(zip(titles tfidf model.get feature names(), li
          st(titles tfidf model.idf )))
          titles tfidf words = set(titles tfidf model.get feature names())
In [161]: # TFIDF w2v title train
```

```
X Train titles tfidf w2v vectors = [];
          for titles sentence in tqdm(X Train preprocessed titles):
              titles vector = np.zeros(300)
              titles tfidf weight = 0;
              for word in titles sentence.split():
                  if (word in glove words) and (word in titles tfidf words):
                      titles vec = model[word]
                      titles tf idf = titles dictionary[word]*(titles sentence.co
          unt(word)/len(titles sentence.split()))
                      titles vector += (titles vec * titles tf idf)
                      titles tfidf weight += titles tf idf
              if titles tfidf weight != 0:
                  titles vector /= titles tfidf weight
              X Train titles tfidf w2v vectors.append(titles vector)
          print(len(X Train titles tfidf w2v vectors))
          print(len(X Train titles tfidf w2v vectors[0]))
          100%|
                   49041/49041 [00:01<00:00, 37478.07it/s]
          49041
          300
In [162]: # TFIDF w2v title train
          X_Test_titles_tfidf_w2v_vectors = [];
          for titles sentence in tqdm(X Test preprocessed titles):
              titles vector = np.zeros(300)
              titles tfidf weight = 0;
```

```
for word in titles sentence.split():
                  if (word in glove_words) and (word in titles_tfidf_words):
                      titles vec = model[word]
                      titles tf idf = titles dictionary[word]*(titles sentence.co
          unt(word)/len(titles sentence.split()))
                      titles vector += (titles vec * titles tf idf)
                      titles tfidf weight += titles tf idf
              if titles tfidf weight != 0:
                  titles vector /= titles tfidf weight
              X Test titles tfidf w2v vectors.append(titles vector)
          print(len(X Test titles tfidf w2v vectors))
          print(len(X Test titles tfidf w2v vectors[0]))
          100%|
                   36052/36052 [00:00<00:00, 37631.02it/s]
          36052
          300
In [163]: # TFIDF w2v title cv
          X_CV_titles_tfidf_w2v_vectors = [];
          for titles sentence in tqdm(X CV preprocessed titles):
              titles vector = np.zeros(300)
              titles tfidf weight = 0;
              for word in titles sentence.split():
                  if (word in glove_words) and (word in titles_tfidf_words):
                      titles_vec = model[word]
```

```
titles tf idf = titles dictionary[word]*(titles sentence.co
          unt(word)/len(titles sentence.split()))
                      titles vector += (titles vec * titles tf idf)
                      titles tfidf weight += titles tf idf
              if titles tfidf weight != 0:
                  titles vector /= titles tfidf weight
              X CV titles tfidf w2v vectors.append(titles vector)
          print(len(X CV titles tfidf w2v vectors))
          print(len(X CV titles tfidf w2v vectors[0]))
          100%|
                   24155/24155 [00:00<00:00, 36639.22it/s]
          24155
          300
In [167]: # Vectorizing numerical feature
          # Merging price data with train, test and cv
          X train = pd.merge(X train, price data, on='id', how='left')
          X test = pd.merge(X test, price data, on='id', how='left')
          X cv = pd.merge(X cv, price data, on='id', how='left')
In [168]: # Standardizing price train test and cv data
          from sklearn.preprocessing import Normalizer
          normalizer = Normalizer()
          # normalizer.fit(X train['price'].values)
          # this will rise an error Expected 2D array, got 1D array instead:
          # array=[105.22 215.96 96.01 ... 368.98 80.53 709.67].
          # Reshape your data either using
          # array.reshape(-1, 1) if your data has a single feature
          \# array.reshape(1, -1) if it contains a single sample.
          normalizer.fit(X train['price'].values.reshape(-1,1))
```

2.4 Appling Logistic Regression on different kind of featurization as mentioned in the instructions

Apply Logistic Regression 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

Set 1: categorical, numerical features + project_title(BOW) + preprocessed_eassay (BOW with bi-grams with min_df=10 and max_features=5000)

```
In [169]: # merge two sparse matrices: https://stackoverflow.com/a/19710648/40840
39
    from scipy.sparse import hstack
# Train data stack
```

```
X tr = hstack((X Train categories one hot,X Train sub categories one ho
          t,X Train school state one hot,
                         X Train teacher prefix one hot, X Train grade cat one hot
           ,X Train essay bow,X Train titles bow,
                         X train price norm)).tocsr()
          # CV data Stack
          X cr = hstack((X CV categories one hot, X CV sub categories one hot, X CV
          school state one hot,
                         X CV teacher prefix one hot, X CV grade cat one hot, X CV
          essay bow, X CV titles bow,
                         X cv price norm)).tocsr()
          # Test Data Stack
          X te = hstack((X Test categories one hot, X Test sub categories one hot,
          X Test school state one hot,
                         X Test teacher prefix one hot, X Test grade cat one hot, X
          Test essay bow, X Test titles bow,
                         X test price norm)).tocsr()
          print("Final Data matrix")
          print(X tr.shape, y train.shape)
          print(X cr.shape, y cv.shape)
          print(X te.shape, y_test.shape)
          print("="*100)
          Final Data matrix
          (49041, 1907) (49041,)
          (24155, 1907) (24155,)
          (36052, 1907) (36052,)
In [170]: def batch predict(clf, data):
              y data pred = []
              # Changing the shape of predicted data in the multiple of 1000
              tr loop = data.shape[0] - data.shape[0]%1000
```

```
# Running the loop for each 1000th data
for i in range(0, tr_loop, 1000):
    # Predicting probability
    y_data_pred.extend(clf.predict_proba(data[i:i+1000])[:,1])

y_data_pred.extend(clf.predict_proba(data[tr_loop:])[:,1])
return y_data_pred
```

```
In [174]: # https://scikit-learn.org/stable/modules/generated/sklearn.model selec
         tion.GridSearchCV.html
         from sklearn.model selection import GridSearchCV
         neigh = LogisticRegression()
         100001}
         clf = GridSearchCV(neigh, parameters, cv=10, scoring='roc auc')
         clf.fit(X tr, y train)
         train auc= clf.cv results ['mean train score']
         train auc std= clf.cv results ['std train score']
         cv auc = clf.cv results ['mean test score']
         cv auc std= clf.cv results ['std test score']
         plt.plot(parameters['C'], train auc, label='Train AUC')
         # this code is copied from here: https://stackoverflow.com/a/48803361/4
         084039
         plt.gca().fill between(parameters['C'], train auc - train auc std, train
         auc + train auc std,alpha=0.2,color='darkblue')
         plt.plot(parameters['C'], cv auc, label='CV AUC')
         # this code is copied from here: https://stackoverflow.com/a/48803361/4
         084039
         plt.gca().fill between(parameters['C'],cv auc - cv auc std,cv auc + cv
         auc std,alpha=0.2,color='darkorange')
         plt.scatter(parameters['C'], train auc, label='Train AUC points')
```

```
plt.scatter(parameters['C'], cv_auc, label='CV AUC points')

plt.xscale("log")
plt.legend()
plt.xlabel("C: hyperparameter")
plt.ylabel("AUC")
plt.title("ERROR PLOTS")
plt.grid()
plt.show()
```

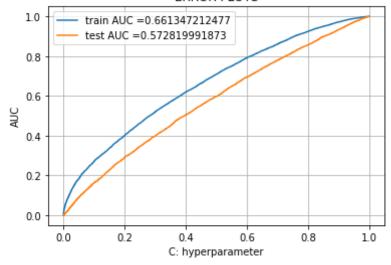
ERROR PLOTS 0.68 Train AUC CV AUC 0.66 Train AUC points CV AUC points 0.64 0.62 0.60 0.58 0.56 0.54 10^{-4} 10^{-2} 10° 10^{2} 10^{4} C: hyperparameter

```
y_test_pred = batch_predict(neigh, X_te)

train_fpr, train_tpr, tr_thresholds = roc_curve(y_train, y_train_pred)
test_fpr, test_tpr, te_thresholds = roc_curve(y_test, y_test_pred)

plt.plot(train_fpr, train_tpr, label="train AUC ="+str(auc(train_fpr, train_tpr)))
plt.plot(test_fpr, test_tpr, label="test AUC ="+str(auc(test_fpr, test_tpr)))
plt.legend()
plt.xlabel("C: hyperparameter")
plt.ylabel("AUC")
plt.title("ERROR PLOTS")
plt.grid()
plt.show()
```

ERROR PLOTS



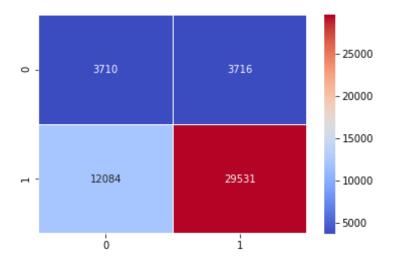
```
In [190]: # we are writing our own function for predict, with defined thresould
# we will pick a threshold that will give the least fpr
def predict(proba, threshould, fpr, tpr):
    t = threshould[np.argmax(tpr*(1-fpr))]
```

```
# (tpr*(1-fpr)) will be maximum if your fpr is very low and tpr is
           very high
              print("the maximum value of tpr*(1-fpr)", max(tpr*(1-fpr)), "for th
          reshold", np.round(t,3))
              predictions = []
              for i in proba:
                  if i>=t:
                      predictions.append(1)
                  else:
                       predictions.append(0)
              return predictions
In [191]: print("="*100)
          from sklearn.metrics import confusion matrix
          print("Train confusion matrix")
          print(confusion matrix(y train, predict(y train pred, tr thresholds, tr
          ain fpr, train fpr)))
          print("Test confusion matrix")
          print(confusion matrix(y test, predict(y test pred, tr thresholds, test
          fpr, test fpr)))
          Train confusion matrix
          the maximum value of tpr*(1-fpr) 0.249999836795 for threshold 0.833
          [[ 3710 3716]
           [12084 295311]
          Test confusion matrix
          the maximum value of tpr*(1-fpr) 0.249999991611 for threshold 0.847
          [[ 5438
                     211
                    178]]
           [30415
In [192]: # https://seaborn.pydata.org/generated/seaborn.heatmap.html
          # https://stackoverflow.com/questions/29647749/seaborn-showing-scientif
          ic-notation-in-heatmap-for-3-digit-numbers
          # Train Confusion Matrix Heatmap
          train confusion matrix = confusion matrix(y train, predict(y train pred
```

```
, tr_thresholds, train_fpr, train_fpr))
print("Train Confusion Matrix")
sns.heatmap(train_confusion_matrix,annot=True,linewidth = 0.1, cmap='coolwarm', fmt='g')
```

the maximum value of tpr*(1-fpr) 0.249999836795 for threshold 0.833 Train Confusion Matrix

Out[192]: <matplotlib.axes._subplots.AxesSubplot at 0x18a334c22b0>



```
In [193]: # https://seaborn.pydata.org/generated/seaborn.heatmap.html
    # Test Confusion Matrix Heatmap

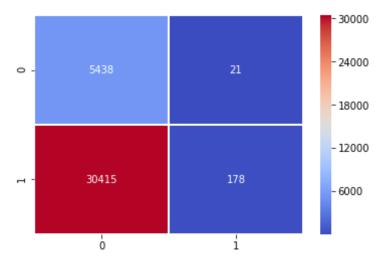
    test_confusion_matrix = confusion_matrix(y_test, predict(y_test_pred, t r_thresholds, test_fpr, test_fpr))

    print("Test Confusion Matrix")
    sns.heatmap(test_confusion_matrix,annot=True,linewidth = 0.5, cmap='coolwarm', fmt='g')

    the maximum value of tpr*(1-fpr) 0.249999991611 for threshold 0.847
```

Test Confusion Matrix





Set 2 : Categorical, Numerical features + Project_title(TFIDF) + Preprocessed_essay (TFIDF with bi-grams with min_df=10 and max_features=5000)

```
# Test Data Stack
         X te = hstack((X Test categories one hot, X Test sub categories one hot,
          X Test school state one hot,
                        X Test teacher prefix one hot, X Test grade cat one hot, X
          _Test_essay_tfidf,X Test titles tfidf,
                        X test price norm)).tocsr()
          print("Final Data matrix")
          print(X tr.shape, y train.shape)
         print(X cr.shape, y cv.shape)
          print(X_te.shape, y test.shape)
          print("="*100)
          Final Data matrix
          (49041, 1907) (49041,)
          (24155, 1907) (24155,)
          (36052, 1907) (36052,)
In [196]: neigh = LogisticRegression()
          100001}
          clf = GridSearchCV(neigh, parameters, cv=10, scoring='roc auc')
          clf.fit(X tr, y train)
          train auc= clf.cv results ['mean train score']
          train auc std= clf.cv results ['std train score']
          cv auc = clf.cv results ['mean test score']
          cv auc std= clf.cv results ['std test score']
          plt.plot(parameters['C'], train auc, label='Train AUC')
         # this code is copied from here: https://stackoverflow.com/a/48803361/4
          084039
          plt.gca().fill between(parameters['C'],train auc - train auc std,train
          auc + train auc std,alpha=0.2,color='darkblue')
          plt.plot(parameters['C'], cv auc, label='CV AUC')
          # this code is copied from here: https://stackoverflow.com/a/48803361/4
```

```
plt.gca().fill_between(parameters['C'],cv_auc - cv_auc_std,cv_auc + cv_auc_std,alpha=0.2,color='darkorange')

plt.scatter(parameters['C'], train_auc, label='Train AUC points')

plt.scatter(parameters['C'], cv_auc, label='CV AUC points')

plt.xscale("log")

plt.legend()

plt.xlabel("C: hyperparameter")

plt.ylabel("AUC")

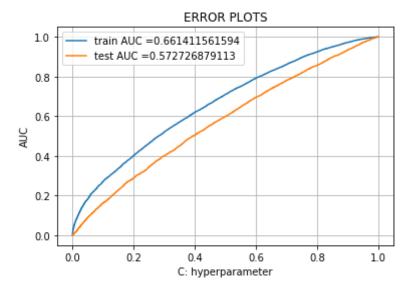
plt.title("ERROR PLOTS")

plt.grid()

plt.show()
```

ERROR PLOTS 0.68 Train AUC CV AUC 0.66 Train AUC points CV AUC points 0.64 0.62 ₩ 0.60 0.58 0.56 0.54 0.52 10^{-4} 10^{-2} 10^{2} 10° 10^{4} C: hyperparameter

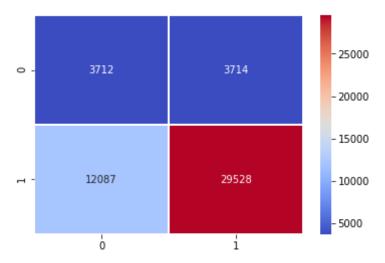
```
# roc_auc_score(y_true, y_score) the 2nd parameter should be probabilit
v estimates of the positive class
# not the predicted outputs
y train pred = batch predict(neigh, X tr)
y test pred = batch predict(neigh, X te)
train fpr, train tpr, tr thresholds = roc curve(y train, y train pred)
test fpr, test tpr, te thresholds = roc curve(y test, y test pred)
plt.plot(train fpr, train tpr, label="train AUC ="+str(auc(train fpr, t
rain tpr)))
plt.plot(test fpr, test tpr, label="test AUC ="+str(auc(test fpr, test
tpr)))
plt.legend()
plt.xlabel("C: hyperparameter")
plt.ylabel("AUC")
plt.title("ERROR PLOTS")
plt.grid()
plt.show()
```



In [200]: print("="*100)

```
from sklearn.metrics import confusion matrix
          print("Train confusion matrix")
          print(confusion matrix(y train, predict(y train pred, tr thresholds, tr
          ain fpr, train fpr)))
          print("Test confusion matrix")
          print(confusion matrix(y test, predict(y test pred, tr thresholds, test
          fpr, test fpr)))
          Train confusion matrix
          the maximum value of tpr*(1-fpr) 0.249999981866 for threshold 0.833
          [[ 3712 3714]
           [12087 29528]]
          Test confusion matrix
          the maximum value of tpr*(1-fpr) 0.249999991611 for threshold 0.847
          [[ 4845 614]
           [25221 5372]]
In [201]: # https://seaborn.pydata.org/generated/seaborn.heatmap.html
          # Train Confusion Matrix Heatmap
          train confusion matrix = confusion matrix(y train, predict(y train pred
          , tr thresholds, train fpr, train fpr))
          print("Train Confusion Matrix")
          sns.heatmap(train confusion matrix,annot=True,linewidth = 0.5, cmap='co
          olwarm', fmt='g')
          the maximum value of tpr*(1-fpr) 0.249999981866 for threshold 0.833
          Train Confusion Matrix
```

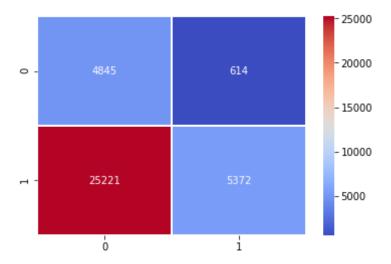
Out[201]: <matplotlib.axes. subplots.AxesSubplot at 0x18a05961860>



Out[202]: <matplotlib.axes. subplots.AxesSubplot at 0x18a610f2cf8>

```
In [202]: # https://seaborn.pydata.org/generated/seaborn.heatmap.html
    # Test Confusion Matrix Heatmap
    test_confusion_matrix = confusion_matrix(y_test, predict(y_test_pred, t r_thresholds, test_fpr, test_fpr))
    print("Test Confusion Matrix")
    sns.heatmap(test_confusion_matrix,annot=True,linewidth = 0.5, cmap='coolwarm', fmt='g')
    the maximum value of tpr*(1-fpr) 0.249999991611 for threshold 0.847
    Test Confusion Matrix
```

Create PDF in your applications with the Pdfcrowd HTML to PDF API



Set 3 : Categorical, Numerical features + Project_title(AVG W2V) + Preprocessed_essay (AVG W2V)

```
In [203]: # Train data stack
          X tr = hstack((X Train categories one hot, X Train sub categories one ho
          t,X Train school state one hot,
                         X Train_teacher_prefix_one_hot,X_Train_grade_cat_one_hot
          ,X Train avg w2v vectors,X Train avg w2v titles vectors,
                         X train price norm)).tocsr()
          # CV data Stack
          X cr = hstack((X CV categories one hot, X CV sub categories one hot, X CV
          school state one hot,
                         X CV teacher prefix one hot, X CV grade cat one hot, X CV
          avg w2v vectors, X CV avg w2v titles vectors,
                         X cv price norm)).tocsr()
          # Test Data Stack
          X_te = hstack((X_Test_categories_one_hot,X_Test_sub_categories_one_hot,
          X Test school state one hot,
                         X Test_teacher_prefix_one_hot,X_Test_grade_cat_one_hot,X
```

```
Test avg w2v vectors, X Test avg w2v titles vectors,
                       X test price norm)).tocsr()
          print("Final Data matrix")
         print(X tr.shape, y train.shape)
         print(X cr.shape, y cv.shape)
         print(X te.shape, y test.shape)
         print("="*100)
         Final Data matrix
         (49041, 701) (49041,)
         (24155, 701) (24155.)
         (36052, 701) (36052,)
In [204]: neigh = LogisticRegression()
         100001}
         clf = GridSearchCV(neigh, parameters, cv=10, scoring='roc auc')
         clf.fit(X tr, y train)
         train auc= clf.cv results ['mean train score']
         train auc std= clf.cv results ['std_train_score']
         cv auc = clf.cv results ['mean test score']
         cv auc std= clf.cv results ['std test score']
         plt.plot(parameters['C'], train auc, label='Train AUC')
         # this code is copied from here: https://stackoverflow.com/a/48803361/4
         084039
         plt.gca().fill between(parameters['C'],train auc - train auc std,train
         auc + train auc std,alpha=0.2,color='darkblue')
         plt.plot(parameters['C'], cv auc, label='CV AUC')
         # this code is copied from here: https://stackoverflow.com/a/48803361/4
          084039
         plt.gca().fill between(parameters['C'],cv auc - cv auc std,cv auc + cv
         auc std,alpha=0.2,color='darkorange')
```

```
plt.scatter(parameters['C'], train_auc, label='Train AUC points')
plt.scatter(parameters['C'], cv_auc, label='CV AUC points')

plt.xscale("log")
plt.legend()
plt.xlabel("C: hyperparameter")
plt.ylabel("AUC")
plt.title("ERROR PLOTS")
plt.grid()
plt.show()
```

ERROR PLOTS 0.57 0.56 NC NC 0.55 0.54 Train AUC CV AUC 0.53 Train AUC points CV AUC points 10^{-4} 10^{-2} 10° 10^{2} 10^{4} C: hyperparameter

```
y_train_pred = batch_predict(neigh, X_tr)
y_test_pred = batch_predict(neigh, X_te)

train_fpr, train_tpr, tr_thresholds = roc_curve(y_train, y_train_pred)
test_fpr, test_tpr, te_thresholds = roc_curve(y_test, y_test_pred)

plt.plot(train_fpr, train_tpr, label="train AUC ="+str(auc(train_fpr, train_tpr)))
plt.plot(test_fpr, test_tpr, label="test AUC ="+str(auc(test_fpr, test_tpr)))
plt.legend()
plt.xlabel("C: hyperparameter")
plt.ylabel("AUC")
plt.title("ERROR PLOTS")
plt.grid()
plt.show()
```

ERROR PLOTS 1.0 train AUC = 0.575070803929 test AUC = 0.555015680253 0.8 0.6 AUC 0.4 0.2 0.0 0.2 0.4 0.8 1.0 0.0 0.6 C: hyperparameter

```
In [217]: print("="*100)
    print("Train confusion matrix")
    print(confusion_matrix(y_train, predict(y_train_pred, tr_thresholds, tr
    ain_fpr, train_fpr)))
    print("Test confusion matrix")
```

```
print(confusion_matrix(y_test, predict(y_test_pred, tr_thresholds, test
           fpr, test fpr)))
          Train confusion matrix
          the maximum value of tpr*(1-fpr) 0.25 for threshold 0.844
          [[ 3713 3713]
           [16260 253551]
          Test confusion matrix
          the maximum value of tpr*(1-fpr) 0.249999991611 for threshold 0.834
          [[ 2043 3416]
           [ 9401 21192]]
In [218]: # https://seaborn.pydata.org/generated/seaborn.heatmap.html
          # Train Confusion Matrix Heatmap
          train confusion matrix = confusion matrix(y train, predict(y train pred
          , tr thresholds, train fpr, train fpr))
          print("Train Confusion Matrix")
          sns.heatmap(train confusion matrix,annot=True,linewidth = 0.5, cmap='Rd
          Pu', fmt='g')
          the maximum value of tpr*(1-fpr) 0.25 for threshold 0.844
          Train Confusion Matrix
Out[218]: <matplotlib.axes. subplots.AxesSubplot at 0x18a62d921d0>
```



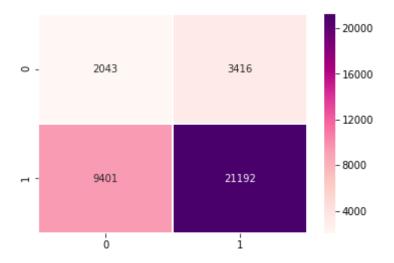
```
In [219]: # https://seaborn.pydata.org/generated/seaborn.heatmap.html
# Test Confusion Matrix Heatmap

test_confusion_matrix = confusion_matrix(y_test, predict(y_test_pred, t r_thresholds, test_fpr, test_fpr))

print("Test Confusion Matrix")
sns.heatmap(test_confusion_matrix,annot=True,linewidth = 0.5, cmap='RdPu', fmt='g')

the maximum value of tpr*(1-fpr) 0.249999991611 for threshold 0.834
Test Confusion Matrix
```

Out[219]: <matplotlib.axes._subplots.AxesSubplot at 0x18a32ee0278>



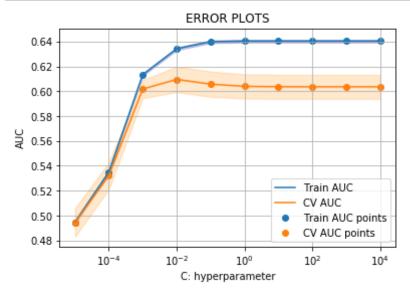
Set 4 : Categorical, Numerical features + Project_title(TFIDF W2V) + Preprocessed_essay (TFIDF W2V)

```
In [220]: # Train data stack
          X tr = hstack((X Train categories one hot, X Train sub categories one ho
          t,X Train school state one hot,
                         X_Train_teacher_prefix_one_hot,X_Train_grade_cat_one_hot
          ,X Train tfidf w2v vectors,X Train titles tfidf w2v vectors,
                         X train price norm)).tocsr()
          # CV data Stack
          X cr = hstack((X CV categories one hot, X CV sub categories one hot, X CV
          school state one hot,
                         X CV teacher prefix one hot, X CV grade cat one hot, X CV
          tfidf w2v vectors,X CV titles tfidf w2v vectors,
                         X cv price norm)).tocsr()
          # Test Data Stack
          X_te = hstack((X_Test_categories_one_hot,X_Test_sub_categories_one_hot,
          X Test school state one hot,
                         X Test teacher_prefix_one_hot,X_Test_grade_cat_one_hot,X
```

```
Test tfidf w2v vectors, X Test titles tfidf w2v vectors,
                       X test price norm)).tocsr()
          print("Final Data matrix")
         print(X tr.shape, y train.shape)
         print(X cr.shape, y cv.shape)
         print(X te.shape, y test.shape)
         print("="*100)
         Final Data matrix
         (49041, 701) (49041,)
         (24155, 701) (24155.)
         (36052, 701) (36052,)
In [221]: neigh = LogisticRegression()
         100001}
         clf = GridSearchCV(neigh, parameters, cv=10, scoring='roc auc')
         clf.fit(X tr, y train)
         train auc= clf.cv results ['mean train score']
         train auc std= clf.cv results ['std_train_score']
         cv auc = clf.cv results ['mean test score']
         cv auc std= clf.cv results ['std test score']
         plt.plot(parameters['C'], train auc, label='Train AUC')
         # this code is copied from here: https://stackoverflow.com/a/48803361/4
         084039
         plt.gca().fill between(parameters['C'],train auc - train auc std,train
         auc + train auc std,alpha=0.2,color='darkblue')
         plt.plot(parameters['C'], cv auc, label='CV AUC')
         # this code is copied from here: https://stackoverflow.com/a/48803361/4
          084039
         plt.gca().fill between(parameters['C'],cv auc - cv auc std,cv auc + cv
         auc std,alpha=0.2,color='darkorange')
```

```
plt.scatter(parameters['C'], train_auc, label='Train AUC points')
plt.scatter(parameters['C'], cv_auc, label='CV AUC points')

plt.xscale("log")
plt.legend()
plt.xlabel("C: hyperparameter")
plt.ylabel("AUC")
plt.title("ERROR PLOTS")
plt.grid()
plt.show()
```



```
y_train_pred = batch_predict(neigh, X_tr)
y_test_pred = batch_predict(neigh, X_te)

train_fpr, train_tpr, tr_thresholds = roc_curve(y_train, y_train_pred)
test_fpr, test_tpr, te_thresholds = roc_curve(y_test, y_test_pred)

plt.plot(train_fpr, train_tpr, label="train AUC ="+str(auc(train_fpr, train_tpr)))
plt.plot(test_fpr, test_tpr, label="test AUC ="+str(auc(test_fpr, test_tpr)))
plt.legend()
plt.xlabel("C: hyperparameter")
plt.ylabel("AUC")
plt.title("ERROR PLOTS")
plt.grid()
plt.show()
```

ERROR PLOTS 1.0 train AUC = 0.637505254698 test AUC = 0.62071870895 0.8 0.6 AUC 0.4 0.2 0.0 0.2 0.4 1.0 0.0 0.6 0.8 C: hyperparameter

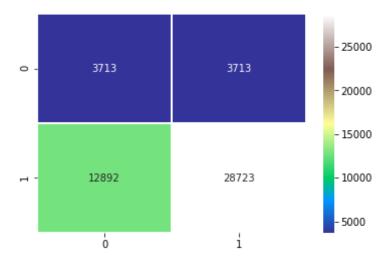
```
In [227]: # https://seaborn.pydata.org/generated/seaborn.heatmap.html
# Train Confusion Matrix Heatmap

train_confusion_matrix = confusion_matrix(y_train, predict(y_train_pred, tr_thresholds, train_fpr, train_fpr))
```

```
print("Train Confusion Matrix")
sns.heatmap(train_confusion_matrix,annot=True,linewidth = 0.5, cmap='te
rrain', fmt='g')
```

the maximum value of tpr*(1-fpr) 0.25 for threshold 0.832 Train Confusion Matrix

Out[227]: <matplotlib.axes._subplots.AxesSubplot at 0x18a60d92e80>



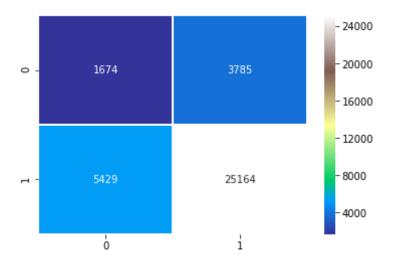
```
In [228]: # https://seaborn.pydata.org/generated/seaborn.heatmap.html
    # Test Confusion Matrix Heatmap

test_confusion_matrix = confusion_matrix(y_test, predict(y_test_pred, t r_thresholds, test_fpr, test_fpr))

print("Test Confusion Matrix")
sns.heatmap(test_confusion_matrix,annot=True,linewidth = 0.5, cmap='terrain',fmt='g')
```

the maximum value of tpr*(1-fpr) 0.249999991611 for threshold 0.854 Test Confusion Matrix

Out[228]: <matplotlib.axes._subplots.AxesSubplot at 0x18a31801400>



2.5 Logistic Regression with added Features `Set 5`

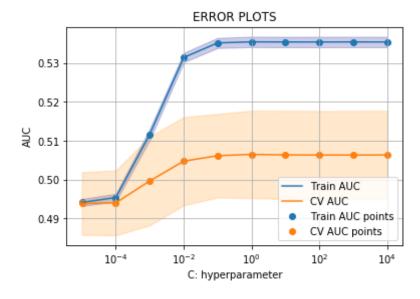
```
In [249]: # Train data stack
          X tr = hstack((X Train categories one hot, X Train sub categories one ho
          t,X Train school state one hot,
                         X Train teacher prefix one hot, X Train grade cat one hot
           ,title word count train,
                         essay word count train, essay sent pos train, essay sent
           neg train, essay sent neu train,
                         essay_sent_comp_train)).tocsr()
          # CV data Stack
          X cr = hstack((X CV categories one hot, X CV sub categories one hot, X CV
          school state one hot,
                         X CV teacher prefix one hot, X CV grade cat one hot, title
           _word_count_cv,
                         essay word count cv, essay sent pos cv, essay sent neg c
          v, essay sent neu cv,
                         essay_sent_comp_cv)).tocsr()
          # Test Data Stack
```

```
X te = hstack((X Test categories one hot, X Test sub categories one hot,
         X Test school state one hot,
                       X Test teacher prefix one hot, X Test grade cat one hot, t
         itle word count test,
                       essay word count test, essay sent pos test, essay sent n
         eq test, essay sent_neu_test,
                       essay sent comp test)).tocsr()
         print("Final Data matrix")
         print(X tr.shape, y train.shape)
         print(X_cr.shape, y cv.shape)
         print(X te.shape, y test.shape)
         print("="*100)
         Final Data matrix
         (49041, 106) (49041,)
         (24155, 106) (24155,)
         (36052, 106) (36052,)
         _____
In [250]: neigh = LogisticRegression()
         100001}
         clf = GridSearchCV(neigh, parameters, cv=10, scoring='roc auc')
         clf.fit(X tr, y train)
         train auc= clf.cv results ['mean train score']
         train auc std= clf.cv results ['std train score']
         cv auc = clf.cv results ['mean test score']
         cv auc std= clf.cv results ['std test score']
         plt.plot(parameters['C'], train auc, label='Train AUC')
         # this code is copied from here: https://stackoverflow.com/a/48803361/4
         084039
         plt.gca().fill between(parameters['C'],train auc - train auc std,train
         auc + train auc std,alpha=0.2,color='darkblue')
         plt.plot(parameters['C'], cv auc, label='CV AUC')
```

```
# this code is copied from here: https://stackoverflow.com/a/48803361/4
084039
plt.gca().fill_between(parameters['C'],cv_auc - cv_auc_std,cv_auc + cv_auc_std,alpha=0.2,color='darkorange')

plt.scatter(parameters['C'], train_auc, label='Train AUC points')
plt.scatter(parameters['C'], cv_auc, label='CV AUC points')

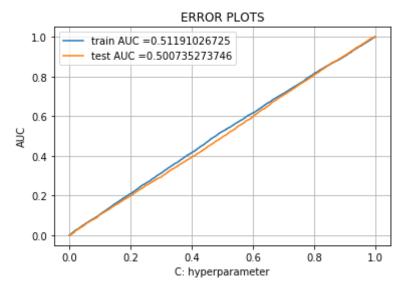
plt.xscale("log")
plt.legend()
plt.xlabel("C: hyperparameter")
plt.ylabel("AUC")
plt.title("ERROR PLOTS")
plt.grid()
plt.show()
```



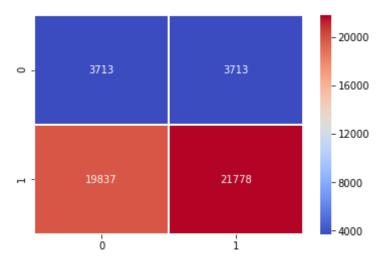
```
In [259]: # https://scikit-learn.org/stable/modules/generated/sklearn.metrics.roc
    _curve.html#sklearn.metrics.roc_curve
    from sklearn.metrics import roc_curve, auc

neigh = LogisticRegression(C=0.001)
```

```
neigh.fit(X_tr, y_train)
# roc_auc_score(y_true, y_score) the 2nd parameter should be probabilit
y estimates of the positive class
# not the predicted outputs
y train pred = batch predict(neigh, X tr)
y test pred = batch predict(neigh, X te)
train fpr, train tpr, tr thresholds = roc curve(y train, y train pred)
test_fpr, test_tpr, te_thresholds = roc_curve(y test, y test pred)
plt.plot(train fpr, train tpr, label="train AUC ="+str(auc(train fpr, t
rain tpr)))
plt.plot(test fpr, test tpr, label="test AUC ="+str(auc(test fpr, test
tpr)))
plt.legend()
plt.xlabel("C: hyperparameter")
plt.vlabel("AUC")
plt.title("ERROR PLOTS")
plt.grid()
plt.show()
```



```
In [260]: print("="*100)
          from sklearn.metrics import confusion matrix
          print("Train confusion matrix")
          print(confusion matrix(y train, predict(y train pred, tr thresholds, tr
          ain_fpr, train fpr)))
          print("Test confusion matrix")
          print(confusion matrix(y test, predict(y test pred, tr thresholds, test
          fpr, test fpr)))
          Train confusion matrix
          the maximum value of tpr*(1-fpr) 0.25 for threshold 0.845
          [[ 3713 3713]
           [19837 21778]]
          Test confusion matrix
          the maximum value of tpr*(1-fpr) 0.249999991611 for threshold 0.846
          [[ 3225 2234]
           [18320 12273]]
In [261]: # https://seaborn.pydata.org/generated/seaborn.heatmap.html
          # Train Confusion Matrix Heatmap
          train confusion matrix = confusion matrix(y train, predict(y train pred
           , tr thresholds, train fpr, train fpr))
          print("Train Confusion Matrix")
          sns.heatmap(train confusion matrix,annot=True,linewidth = 0.5, cmap='co
          olwarm', fmt='g')
          the maximum value of tpr*(1-fpr) 0.25 for threshold 0.845
          Train Confusion Matrix
Out[261]: <matplotlib.axes. subplots.AxesSubplot at 0x18a53278160>
```



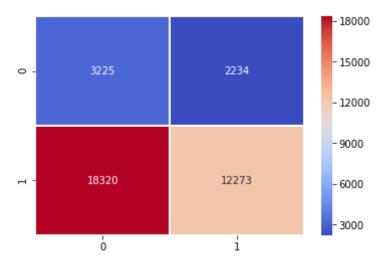
```
In [262]: # https://seaborn.pydata.org/generated/seaborn.heatmap.html
# Test Confusion Matrix Heatmap

test_confusion_matrix = confusion_matrix(y_test, predict(y_test_pred, t r_thresholds, test_fpr, test_fpr))

print("Test Confusion Matrix")
sns.heatmap(test_confusion_matrix,annot=True,linewidth = 0.5, cmap='coolwarm', fmt='g')

the maximum value of tpr*(1-fpr) 0.249999991611 for threshold 0.846
Test Confusion Matrix
```

Out[262]: <matplotlib.axes._subplots.AxesSubplot at 0x18a612eec50>



3. Conclusion

```
In [265]: # http://zetcode.com/python/prettytable/
    from prettytable import PrettyTable

x = PrettyTable()

x.field_names = ["Featurization", "Best C Value", "Train AUC", "Test AUC"]

x.add_row(["Logistic Regression on BOW", 0.5, 0.66, 0.57])
x.add_row(["Logistic Regression on TFIDF", 0.5, 0.66, 0.57])
x.add_row(["Logistic Regression on AVG W2V", 0.1, 0.57, 0.55])
x.add_row(["Logistic Regression on TFIDF W2V", 0.05, 0.64, 0.62])
x.add_row(["Logistic Regression on Categorical features, Numerical features & Essay Sentiments", 0.001, 0.51, 0.50])
print(x)
```