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		
project_subject_subcategories	One or more (comma-separated) su subcategories for the project. Exam • Literacy • Literature & Writing, Social Sciences		
project_resource_summary	An explanation of the resources nee the project. Example: • My students need hands of 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
	A binary flag indicating whether DonorsChoose approved the project. A value of θ indicates the project was not approved, and a value of 1 indicates the project was approved.

Notes on the Essay Data

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

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

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

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

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

```
In [49]: %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 [50]: project data = pd.read csv('train data.csv')
         resource data = pd.read csv('resources.csv')
In [51]:
         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 (32414, 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 [52]: 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 (32414, 4)
         ['id' 'description' 'quantity' 'price']
Out[52]:
                                                    description quantity
                 id
                                                                       price
          0 p253737 LC652 - Lakeshore Double-Space Mobile Drying Rack 1
                                                                       149.00
          1 p258326 Bouncy Bands for Desks (Blue support pipes)
                                                               3
                                                                       14.95
```

1.2 preprocessing of project_subject_categories

In [53]: 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 [54]: 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 [56]: project_data.head(2)

Out[56]:

	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

1.4.2.3 Using Pretrained Models: TFIDF weighted W2V

```
In [57]: # printing some random reviews
    print(project_data['essay'].values[0])
    print("="*50)
    print(project_data['essay'].values[150])
    print(project_data['essay'].values[1000])
    print("="*50)
    print(project_data['essay'].values[20000])
    print("="*50)
```

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

lo to halm their child learn mhanetics, 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

ante to do dock work and mayo at the same time. These steels 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

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 [58]: # 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"\'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"\'ve", " have", phrase)
    return phrase
```

```
In [59]: 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 [60]: # \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 [61]: #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 [63]: # 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%|
                 | 32414/32414 [00:17<00:00, 1886.24it/s]
In [64]: # after preprocesing
         preprocessed essays[20000]
```

Out[64]: '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'

1.4 Preprocessing of `project_title`

```
In [65]: # printing some project titles.
      for i in range (0,21):
          print(project data['project title'].values[i])
          print("="*50)
      Educational Support for English Learners at Home
      Wanted: Projector for Hungry Learners
       _____
      Soccer Equipment for AWESOME Middle School Students
      ______
      Techie Kindergarteners
      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 [66]: 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
      moving stopwords
         preprocessed titles.append(data.lower().strip()) # Creating array i
      n all the lower cases.
      100%|
            32414/32414 [00:00<00:00, 33274.17it/s]
In [67]: 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 [68]: 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[68]:
           Unnamed:
                         id
                                               teacher_id | teacher_prefix | school_state
```

	Unnamed: 0	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.	тх

1.5 Preparing data for models

```
In [69]: project data.columns
Out[69]: Index(['Unnamed: 0', 'id', 'teacher id', 'teacher prefix', 'school stat
         е',
                 'project_submitted_datetime', 'project_grade_category', 'project
         title',
                 'project_essay_1', 'project_essay_2', 'project_essay_3',
                 'project_essay_4', 'project_resource_summary',
                'teacher number of previously_posted_projects', 'project_is_appr
         oved',
                 'clean categories', 'clean subcategories', 'essay',
                'preprocessed titles', 'title word count'],
               dtype='object')
         we are going to consider
                - school state : categorical data
                - clean categories : categorical data
                - clean subcategories : categorical data
                - project grade category : categorical data
                - teacher prefix : categorical data
```

```
- project_title : text data
- text : text data
- project_resource_summary: text data (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 [70]: # 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 (32414, 9)
In [71]: # 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'].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', 'Literacv'l
         Shape of matrix after one hot encodig (32414, 30)
In [72]: 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 (32414, 51)
         the type of count vectorizer <class 'scipy.sparse.csr.csr matrix'>
In [73]: # https://www.geeksforgeeks.org/pvthon-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)
         ['Mr', 'Mrs', 'Ms', 'No Prefix', 'Teacher']
         Shape of matrix after one hot encodig (32414, 5)
In [74]: 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 (32414, 4)
         1.5.2 Vectorizing Text data
```

3

1.5.2.1 Bag of words

```
In [75]: # 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 (32414, 10299)
In [76]: 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', 'about', 'acad
         emic', 'access', 'accessible', 'accessing', 'accessories', 'achieve',
         'achievement', 'achieving']
         Shape of matrix after one hot encodig (32414, 1601)
         the type of count vectorizer <class 'scipy.sparse.csr.csr matrix'>
         the number of unique words 1601
         1.5.2.2 TFIDF vectorizer
In [77]: 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 (32414, 10299)
```

1.5.2.3 Using Pretrained Models: Avg W2V

```
In [78]:
         # 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 = [1]
         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
         ur coupus", \
               len(inter_words), "(", np.round(len(inter_words)/len(words)*100,
```

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

# stronging variables into pickle files python: http://www.jessicayung.
com/how-to-use-pickle-to-save-and-load-variables-in-python/
import pickle
with open('glove_vectors', 'wb') as f:
    pickle.dump(words_courpus, f)
```

Out[78]: '\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 model[word] = embedding\n splitLine[1:]])\n print ("Done.",le n(model)." words loaded!")\n return model\nmodel = loadGloveModel (\'alove.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 words.extend(i.split(\' \'))\n\nfor i in preproce preproced texts:\n words.extend(i.split(\' \'))\nprint("all the words in th 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 len(inter words),"(",np.round(len(inter wor tors and our coupus", ds)/len(words)*100,3),"%)")\n\nwords courpus = {}\nwords glove = set(mo 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'
In [79]: # 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 [80]: # 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
             avq w2v vectors.append(vector)
         print(len(avg w2v vectors))
         print(len(avg w2v vectors[0]))
         100%|
                  | 32414/32414 [00:08<00:00, 3719.79it/s]
         32414
         300
         1.5.2.3 Using Pretrained Models: TFIDF weighted W2V
```

```
In [81]: \# S = ["abc def pqr", "def def def abc", "pqr pqr def"]
         tfidf model = TfidfVectorizer()
         tfidf model.fit(preprocessed essays)
         # we are converting a dictionary with word as a key, and the idf as a v
         alue
         dictionary = dict(zip(tfidf model.get feature names(), list(tfidf model
         .idf )))
         tfidf words = set(tfidf model.get feature names())
In [82]: # 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 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
             tfidf w2v vectors.append(vector)
         print(len(tfidf w2v vectors))
         print(len(tfidf w2v vectors[0]))
         100%
                   32414/32414 [01:04<00:00, 500.10it/s]
         32414
         300
```

```
In [83]: # TFIDF on project titles
         titles tfidf vectorizer = TfidfVectorizer(min df=10)
         titles tfidf = titles tfidf vectorizer.fit transform(preprocessed title
         s)
         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) ['across', 'act', 'act
         ion', 'active', 'activities', 'activity', 'add', 'adding', 'adventure',
         'adventures', 'after']
         Shape of matrix after one hot encodig (32414, 1601)
In [84]: # 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%
                  32414/32414 [00:00<00:00, 73238.03it/s]
```

```
32414
         300
In [85]: # 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 [86]: 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%
                  32414/32414 [00:00<00:00, 37230.40it/s]
```

32414 300

1.5.3 Vectorizing Numerical features

```
In [87]: price data = resource data.groupby('id').agg({'price':'sum', 'quantity'
         :'sum'}).reset index()
         project data = pd.merge(project data, price data, on='id', how='left')
In [88]: # check this one: https://www.youtube.com/watch?v=0H0q0cln3Z4&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: 51.22791725797495, Standard deviation: 149.72149781041014
In [89]: price standardized
Out[89]: array([[ 0.65302635],
                [-0.24230266],
                [-0.2857166],
```

```
[-0.10978996],
[-0.23014676],
[-0.10250978]])
```

1.5.4 Merging all the above features

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

```
In [90]: print(categories one hot.shape)
         print(sub categories one hot.shape)
         print(text bow.shape)
         print(price standardized.shape)
         (32414, 9)
         (32414, 30)
         (32414, 10299)
         (32414, 1)
In [91]: # merge two sparse matrices: https://stackoverflow.com/a/19710648/40840
         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[91]: (32414, 10339)
         Computing Sentiment Scores
         import nltk
In [92]:
         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
v 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, comp
ound)
# 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 10: Clustering

- step 1: Choose any vectorizer (data matrix) that you have worked in any of the assignments, and got the best AUC value.
- step 2: Choose any of the <u>feature selection/reduction algorithms</u> ex: selectkbest features, pretrained word vectors, model based feature selection etc and reduce the number of features to 5k features
- step 3: Apply all three kmeans, Agglomerative clustering, DBSCAN
 - K-Means Clustering:
 - Find the best 'k' using the elbow-knee method (plot k vs inertia_)
 - Agglomerative Clustering:
 - Apply <u>agglomerative algorithm</u> and try a different number of clusters like 2,5 etc.
 - You can take less data points (as this is very computationally expensive one) to perform hierarchical clustering because they do take a considerable amount of time to run.
 - DBSCAN Clustering:
 - Find the best 'eps' using the elbow-knee method.
 - You can take a smaller sample size for this as well.
- step 4: Summarize each cluster by manually observing few points from each cluster.
- step 5: You need to plot the word cloud with essay text for each cluster for each of algorithms mentioned in step 3.

2. Clustering

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

```
In [93]: approved_project = project_data['project_is_approved'].values
    project_data.drop(['project_is_approved'], axis=1, inplace=True)
    project_data.head(1)
```

Out[93]:

	Unnamed: 0	id	teacher_id	teacher_prefix	school_state	k
0	160221	p253737	c90749f5d961ff158d4b4d1e7dc665fc	Mrs.	ΙZ	С

1 rows × 21 columns

```
In [94]: # 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)
```

2.1 Choose the best data matrix on which you got the

best AUC

Naive Bayes Model with bow/tfidf feature had best AUC value 0.71. For this assignment we will use bow feature.

2.2 Make Data Model Ready: encoding numerical, categorical features

```
In [95]: # Vectorizing Categories on Train, Test and CV data
         from sklearn.feature extraction.text import CountVectorizer
         ccvectorizer = CountVectorizer(vocabulary=list(sorted cat dict.keys()),
          lowercase=False, binary=True)
         # Fit only to train data
         ccvectorizer.fit(X train['clean categories'].values)
         # Transform to train, test and CV data
         X Train categories one hot = ccvectorizer.transform(X_train['clean_cate
         gories'l.values)
         X Test categories one hot = ccvectorizer.transform(X test['clean catego
         ries'].values)
         X CV categories one hot = ccvectorizer.transform(X cv['clean categorie
         s'l.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 (14550, 9)
         Shape of test matrix after one hot encodig (10697, 9)
         Shape of cv matrix after one hot encodig (7167, 9)
```

```
In [96]: # Vectorizing subcategories on train, test and cv
         csvectorizer = CountVectorizer(vocabulary=list(sorted sub cat dict.keys
         ()), lowercase=False, binary=True)
         csvectorizer.fit(X train['clean subcategories'].values)
         X Train sub categories one hot = csvectorizer.transform(X train['clean
         subcategories'l.values)
         X Test sub categories one hot = csvectorizer.transform(X test['clean su
         bcategories'l.values)
         X CV sub categories one hot = csvectorizer.transform(X cv['clean subcat
         egories'].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 (14550, 30)
         Shape of test matrix after one hot encodig (10697, 30)
         Shape of cv matrix after one hot encodig (7167, 30)
In [97]: # 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 (14550, 51)
         Shape of test matrix after one hot encodig (10697, 51)
         Shape of cv matrix after one hot encodig (7167, 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 [98]: # 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
         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)
         ['Mr', 'Mrs', 'Ms', 'Teacher']
         Shape of train matrix after one hot encodig (14550, 4)
         Shape of test matrix after one hot encodig (10697, 4)
         Shape of cv matrix after one hot encodig (7167, 4)
In [99]: # 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 q
         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['pro
         ject grade category'l.values)
         X Test grade cat one hot = grade cat vectorizer.transform(X test['proje
         ct grade category'l.values)
```

```
X CV grade cat one hot = grade cat vectorizer.transform(X cv['project g
rade category'l.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 on
e hot.shape)
print("Shape of cv matrix after one hot encodig ", X CV grade cat one ho
t.shape)
['Grades~9-12', 'Grades~6-8', 'Grades~3-5', 'Grades~PreK-2']
Shape of train matrix after one hot encodig (14550, 4)
```

Shape of test matrix after one hot encodig (10697, 4) Shape of cv matrix after one hot encodig (7167, 4)

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

```
In [100]: # merge two column text dataframe:
          X train["essay"] = X train["project essay 1"].map(str) +\
                                  X train["project essay 2"].map(str) + \
                                  X train["project essay 3"].map(str) + \
                                  X train["project essay 4"].map(str)
In [101]: # preprocessing essay train data
          from tqdm import tqdm
          X Train preprocessed essays = []
          # tqdm is for printing the status bar
          for sentence in tgdm(X train['essay'].values):
              X Train essay sent = decontracted(sentance)
              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%
                    14550/14550 [00:12<00:00, 1212.25it/s]
In [102]: # preprocessing essay test data
          from tqdm import tqdm
          X Test preprocessed essays = []
          # tgdm is for printing the status bar
          for sentence in tgdm(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%|
                    10697/10697 [00:05<00:00, 1889.83it/s]
In [103]: # preprocessing essay cv data
          from tqdm import tqdm
          X CV preprocessed essays = []
          # tqdm is for printing the status bar
          for sentence in tgdm(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}=ssay_{sent} = re.sub('[^A-Za-z0-9]+', ' ', X_{CV}=ssay_{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%|
                     | 7167/7167 [00:03<00:00, 1805.17it/s]
In [104]: # 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%|
                    14550/14550 [00:00<00:00, 31590.38it/s]
In [105]: # preprocessing project title test data
          X Test preprocessed titles = []
           for dataset in tqdm(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
          moving stopwords
               X Test preprocessed titles.append(data.lower().strip()) # Creating
           array in all the lower cases.
          100%
                   10697/10697 [00:00<00:00, 33760.97it/s]
In [106]: # 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%|
                    | 7167/7167 [00:00<00:00, 27793.03it/s]
```

```
In [107]: # BOW Essay train, test and cv data
          # We are considering only the words which appeared in at least 10 docum
          ents(rows or projects).
          bow essay vectorizer = CountVectorizer(min df=10, max features = 5000)
          bow essay vectorizer.fit(X Train preprocessed_essays)
          X Train essay bow = bow essay vectorizer.transform(X Train preprocessed
          essavs)
          X Test essay bow = bow essay vectorizer.transform(X Test preprocessed e
          ssays)
          X CV essay bow = bow essay vectorizer.transform(X CV preprocessed essay
          s)
          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 train matrix after one hot encodig (14550, 133)
          Shape of test matrix after one hot encodig (10697, 133)
          Shape of CV matrix after one hot encodig (7167, 133)
In [108]: # BOW title train, test and cv data
          titles vectorizer = CountVectorizer(min df=10, max features = 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
          .shape)
          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', 'academic',
          'access', 'action', 'active', 'activities', 'activity', 'add', 'addin
          g', 'adventure']
          Shape of train matrix after one hot encodig (14550, 879)
          Shape of test matrix after one hot encodig (10697, 879)
          Shape of CV matrix after one hot encodig (7167, 879)
In [111]: # Vectorizing numerical feature
          # Merging price data with train, test and cv
          X train = pd.merqe(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 [112]: # 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))
          X train price norm = normalizer.transform(X train['price'].values.resha
          pe(-1,1))
          X test price norm = normalizer.transform(X test['price'].values.reshape
          (-1,1))
          X cv price norm = normalizer.transform(X cv['price'].values.reshape(-1,
          1))
          print("After vectorizations")
          print(X train price norm.shape, y train.shape)
```

```
print(X test price norm.shape, y test.shape)
          print(X cv price norm.shape, y cv.shape)
          print("="*100)
          After vectorizations
          (14550, 1) (14550,)
          (10697, 1) (10697,)
          (7167, 1) (7167,)
In [113]: normalizer.fit(X train['teacher number of previously posted projects'].
          values.reshape(-1,1))
          X train price norm = normalizer.transform(X train['teacher number of pr
          eviously posted projects'].values.reshape(-1,1))
          X test price norm = normalizer.transform(X test['teacher number of prev
          iously posted projects'].values.reshape(-1,1))
          X_cv_price_norm = normalizer.transform(X_cv['teacher_number of previous
          ly_posted_projects'].values.reshape(-1,1)
          print("After vectorizations")
          print(X train price norm.shape, y train.shape)
          print(X test price norm.shape, y test.shape)
          print(X cv price norm.shape, y cv.shape)
          print("="*100)
          After vectorizations
          (14550, 1) (14550,)
          (10697, 1) (10697,)
          (7167, 1) (7167,)
In [114]: # merge two sparse matrices: https://stackoverflow.com/a/19710648/40840
          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
(14550, 1111) (14550,)
(7167, 1111) (7167,)
(10697, 1111) (10697.)
```

2.4 Dimensionality Reduction on the selected features

```
ection.f classif.html#sklearn.feature selection.f classif
from sklearn.feature selection import SelectKBest, chi2, f classif
new_feature = SelectKBest(f classif, k=1000) # reduced to 1000 fea
tures as total features are 1111.
new feature.fit(X tr,y train)
X tr new = new feature.transform(X tr)
X te new = new feature.transform(X te)
X cr new = new feature.transform(X cr)
print("After Dimensionality Reduction")
print(X tr new.shape, y train.shape)
print(X cr new.shape, y cv.shape)
print(X te new.shape, y test.shape)
print("="*100)
After Dimensionality Reduction
(14550, 1000) (14550,)
(7167, 1000) (7167,)
(10697, 1000) (10697,)
_____
```

2.5 Apply Kmeans

```
In [116]: # https://scikit-learn.org/stable/modules/generated/sklearn.cluster.KMe
    ans.html
    # https://www.geeksforgeeks.org/elbow-method-for-optimal-value-of-k-in-
    kmeans/

    from sklearn.cluster import KMeans
    hyperparameter = [2, 4, 8, 16]
    inertias = []

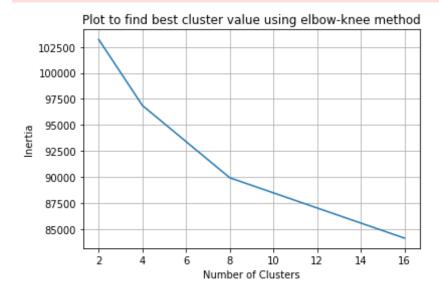
    for n_cluster in tqdm(hyperparameter):
```

```
kmeans = KMeans(n_clusters=n_cluster, random_state=0).fit(X_tr_new)
inertias.append(kmeans.inertia_)

plt.plot(hyperparameter, inertias)
plt.xlabel("Number of Clusters")
plt.ylabel("Inertia")
plt.title("Plot to find best cluster value using elbow-knee method")
plt.grid()
plt.show()
```

100%|

4/4 [02:51<00:00, 42.28s/it]



```
In [117]: kmeans_best = KMeans(n_clusters=8, random_state=0).fit(X_tr_new)
```

In [118]: kmeans_best.labels_

Out[118]: array([5, 5, 5, ..., 4, 3, 3])

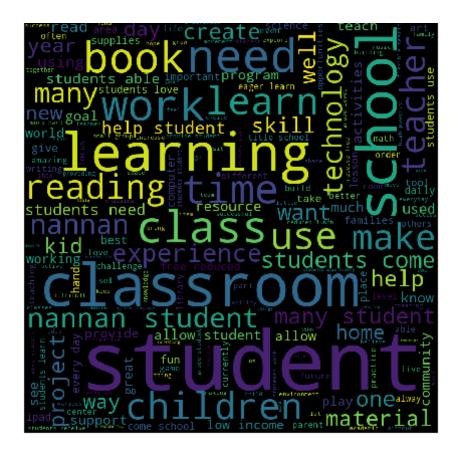
In [119]: len(kmeans_best.labels_)

```
Out[119]: 14550
In [218]: # Creating different clusters
          cluster 0 = []
          cluster 1 = []
          cluster 2 = [1]
          cluster 3 = []
          cluster 4 = []
          cluster 5 = []
          cluster 6 = []
          cluster 7 = []
          for val in range(0, len(kmeans_best.labels_)):
              if (kmeans_best.labels_[val] == 0):
                  cluster 0.append(preprocessed essays[val])
              if (kmeans best.labels [val] == 1):
                  cluster 1.append(preprocessed essays[val])
              if kmeans best.labels [val] == 2:
                  cluster 2.append(preprocessed essays[val])
              if kmeans best.labels [val] == 3:
                  cluster 3.append(preprocessed essays[val])
              if kmeans best.labels [val] == 4:
                  cluster 4.append(preprocessed essays[val])
              if kmeans best.labels [val] == 5:
                  cluster 5.append(preprocessed essays[val])
              if kmeans best.labels [val] == 6:
                  cluster 6.append(preprocessed essays[val])
              if kmeans_best.labels_[val] == 7:
                  cluster 7.append(preprocessed essays[val])
```

```
In [220]: from wordcloud import WordCloud
          def create wordcloud(cluster, cluster num, algo name):
              unique_string=(" ").join(cluster)
              wordcloud = WordCloud(width = 800, height = 800, background color =
           'black', min font size = 6).generate(unique_string)
              print("Word cloud plot for essay text for cluster {} for {} algorit
          hm".format(cluster num, algo name))
              plt.figure(figsize = (8, 6))
              plt.imshow(wordcloud)
              plt.axis("off")
              plt.tight layout(pad = 0)
              plt.show()
              plt.close()
In [221]: # https://www.geeksforgeeks.org/python-program-to-count-words-in-a-sent
          ence/
          # https://www.geeksforgeeks.org/find-k-frequent-words-data-set-python/
          def wordcloud analysis(cluster, cluster num):
              from collections import Counter
              word data = []
              final data = []
              for i in range(len(cluster)):
                  word data.append(cluster[i])
              word data = " ".join(word data)
              cluster word split = word data.split()
              #print(cluster word split)
```

```
tot_words = len(cluster_word_split)
              print("Total number of words in cluster {} are {}".format(cluster n
          um, tot_words))
              Counter = Counter(cluster word split)
              cluster most common words = Counter.most common(10)
              print("Top 10 most common words in cluster {} are {}".format(cluste
          r num, cluster most common words))
In [222]: cluster = [cluster 0, cluster 1, cluster 2, cluster 3, cluster 4, clust
          er_5, cluster_6, cluster 7]
          count = 0
          for val in cluster:
              create_wordcloud(val, count, "K-Means")
              wordcloud analysis(val, count)
              count +=1
```

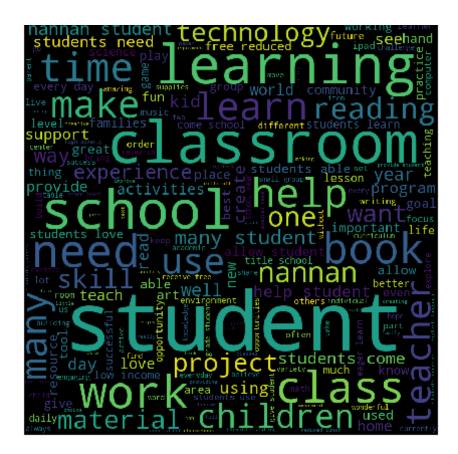
Word cloud plot for essay text for cluster 0 for K-Means algorithm



Total number of words in cluster 0 are 275009

Top 10 most common words in cluster 0 are [('students', 13165), ('i', 5 410), ('school', 4202), ('my', 3036), ('learning', 2924), ('classroom', 2802), ('they', 2210), ('not', 2209), ('the', 2162), ('learn', 2082)]

Word cloud plot for essay text for cluster 1 for K-Means algorithm

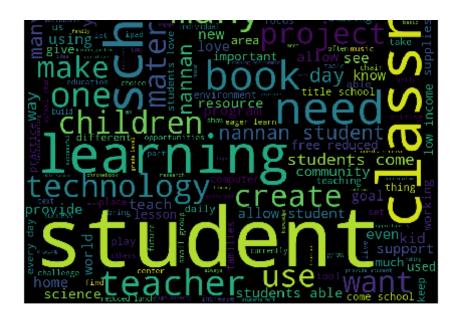


Total number of words in cluster 1 are 233192

Top 10 most common words in cluster 1 are [('students', 11201), ('i', 4 741), ('school', 3596), ('my', 2658), ('learning', 2485), ('classroom', 2461), ('learn', 1885), ('the', 1854), ('they', 1823), ('help', 1765)]

Word cloud plot for essay text for cluster 2 for K-Means algorithm

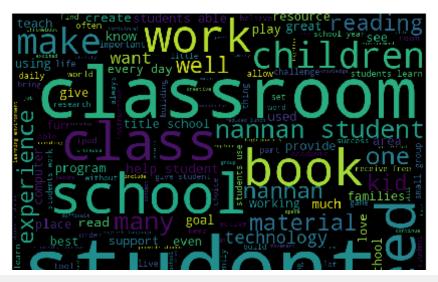


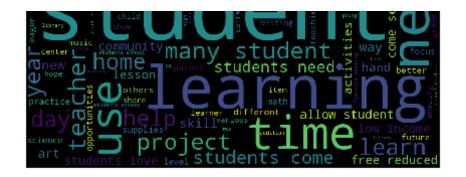


Total number of words in cluster 2 are 242928

Top 10 most common words in cluster 2 are [('students', 11806), ('i', 5 000), ('school', 3776), ('my', 2662), ('learning', 2599), ('classroom', 2481), ('the', 2032), ('not', 1971), ('learn', 1878), ('they', 1855)]

Word cloud plot for essay text for cluster 3 for K-Means algorithm

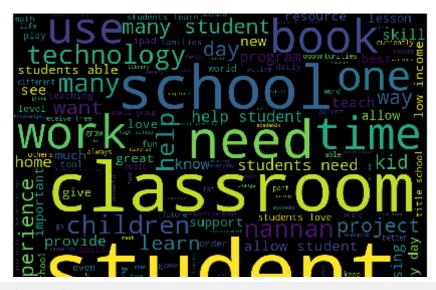




Total number of words in cluster 3 are 335322

Top 10 most common words in cluster 3 are [('students', 15916), ('i', 6 826), ('school', 5133), ('my', 3715), ('learning', 3624), ('classroom', 3555), ('the', 2813), ('they', 2653), ('not', 2601), ('help', 2494)]

Word cloud plot for essay text for cluster 4 for K-Means algorithm

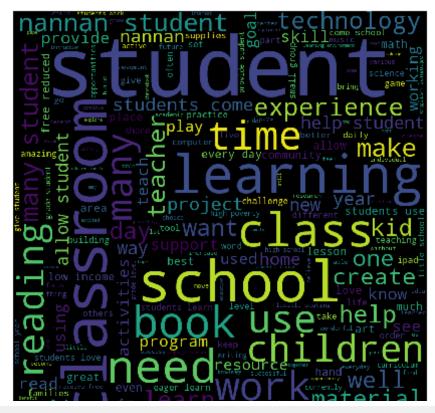




Total number of words in cluster 4 are 326325

Top 10 most common words in cluster 4 are [('students', 15621), ('i', 6 452), ('school', 5047), ('my', 3684), ('learning', 3491), ('classroom', 3370), ('the', 2648), ('they', 2635), ('not', 2606), ('learn', 2520)]

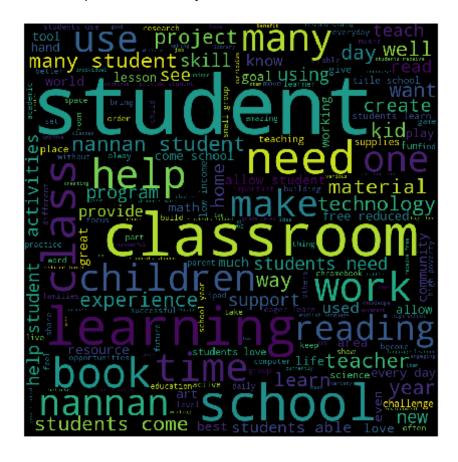
Word cloud plot for essay text for cluster 5 for K-Means algorithm



Total number of words in cluster 5 are 319401

Top 10 most common words in cluster 5 are [('students', 15312), ('i', 6 463), ('school', 4984), ('my', 3501), ('learning', 3494), ('classroom', 3334), ('the', 2655), ('not', 2513), ('learn', 2438), ('help', 2437)]

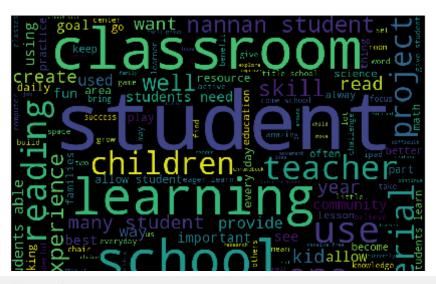
Word cloud plot for essay text for cluster 6 for K-Means algorithm

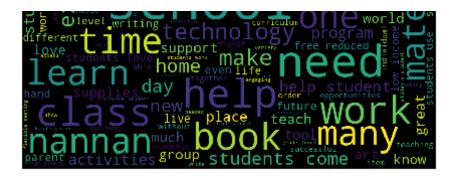


Total number of words in cluster 6 are 289543

Top 10 most common words in cluster 6 are [('students', 13771), ('i', 5 947), ('school', 4587), ('my', 3127), ('learning', 3057), ('classroom', 2949), ('the', 2317), ('they', 2315), ('not', 2312), ('learn', 2199)]

Word cloud plot for essay text for cluster 7 for K-Means algorithm





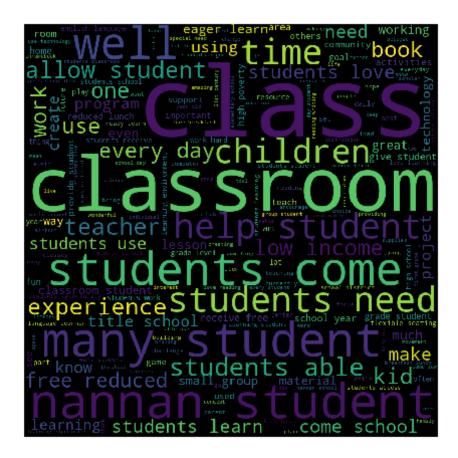
Total number of words in cluster 7 are 176070

Top 10 most common words in cluster 7 are [('students', 8489), ('i', 3560), ('school', 2700), ('my', 2023), ('learning', 1860), ('classroom', 1792), ('the', 1483), ('not', 1413), ('they', 1344), ('help', 1313)]

2.6 Apply AgglomerativeClustering

```
In [223]: # https://stackabuse.com/hierarchical-clustering-with-python-and-scikit
-learn/
# https://scikit-learn.org/stable/modules/generated/sklearn.cluster.Agg
lomerativeClustering.html
```

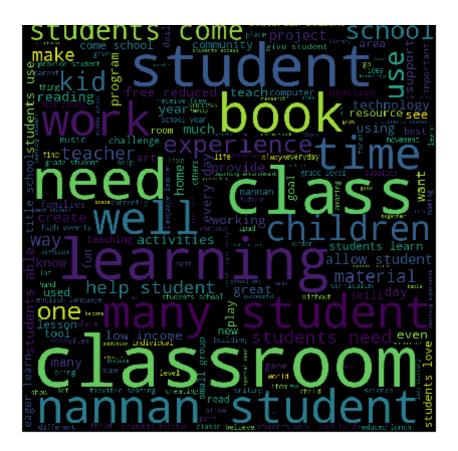
```
from sklearn.cluster import AgglomerativeClustering
          clustering = AgglomerativeClustering(n clusters=2).fit(X tr new.toarray
          ())
In [224]: clustering.labels
Out[224]: array([1, 1, 1, ..., 0, 0, 0], dtype=int64)
In [225]: # Creating different clusters
          cluster 0 = []
          cluster 1 = []
          for val in range(0, len(clustering.labels )):
              if (clustering.labels [val] == 0):
                  cluster 0.append(preprocessed essays[val])
              if (clustering.labels [val] == 1):
                  cluster 1.append(preprocessed essays[val])
In [226]: cluster = [cluster_0, cluster_1]
          count = 0
          for val in cluster:
              create wordcloud(val, count, "Agglomerative")
              wordcloud analysis(val, count)
              count +=1
          Word cloud plot for essay text for cluster 0 for Agglomerative algorith
```



Total number of words in cluster 0 are 1542836

Top 10 most common words in cluster 0 are [('students', 73906), ('i', 3 1235), ('school', 23899), ('my', 17300), ('learning', 16540), ('classro om', 15965), ('the', 12590), ('not', 12166), ('they', 12122), ('learn', 11893)]

Word cloud plot for essay text for cluster 1 for Agglomerative algorith

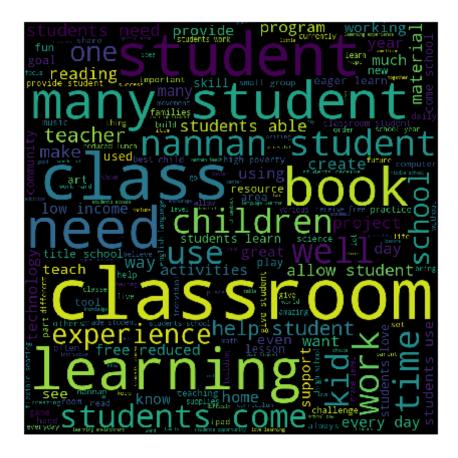


Total number of words in cluster 1 are 654954

Top 10 most common words in cluster 1 are [('students', 31375), ('i', 1 3164), ('school', 10126), ('my', 7106), ('learning', 6994), ('classroo m', 6779), ('the', 5374), ('not', 5218), ('they', 5117), ('help', 497 0)]

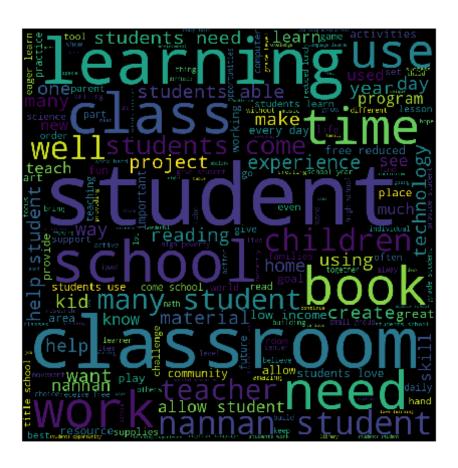
```
In [227]: # https://stackabuse.com/hierarchical-clustering-with-python-and-scikit
          -learn/
          # https://scikit-learn.org/stable/modules/generated/sklearn.cluster.Agg
          lomerativeClustering.html
          from sklearn.cluster import AgglomerativeClustering
          clustering = AgglomerativeClustering(n clusters=5).fit(X tr new.toarray
          ())
In [228]: clustering.labels
Out[228]: array([0, 0, 0, ..., 2, 2, 2], dtype=int64)
In [229]: # Creating different clusters
          cluster 0 = []
          cluster 1 = []
          cluster 2 = []
          cluster 3 = []
          cluster 4 = []
          for val in range(0, len(clustering.labels )):
              if (clustering.labels [val] == 0):
                  cluster 0.append(preprocessed essays[val])
              if (clustering.labels [val] == 1):
                  cluster 1.append(preprocessed essays[val])
              if (clustering.labels [val] == 2):
                  cluster 2.append(preprocessed essays[val])
              if (clustering.labels [val] == 3):
                  cluster 3.append(preprocessed essays[val])
              if (clustering.labels [val] == 4):
                  cluster 4.append(preprocessed essays[val])
```

Word cloud plot for essay text for cluster 0 for Agglomerative algorith m



Total number of words in cluster 0 are 654954

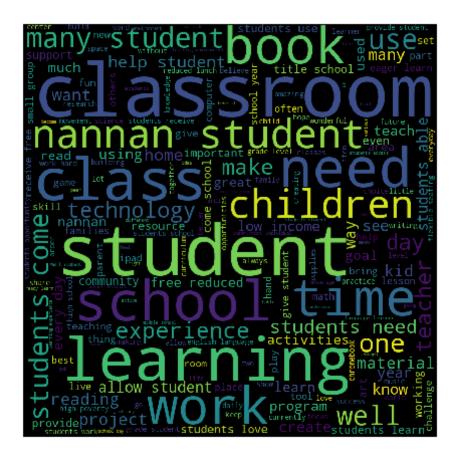
Top 10 most common words in cluster 0 are [('students', 31375), ('i', 1 3164), ('school', 10126), ('my', 7106), ('learning', 6994), ('classroo m', 6779), ('the', 5374), ('not', 5218), ('they', 5117), ('help', 497 0)]
Word cloud plot for essay text for cluster 1 for Agglomerative algorith



Total number of words in cluster 1 are 462009

Top 10 most common words in cluster 1 are [('students', 22110), ('i', 9 223), ('school', 7115), ('my', 5188), ('learning', 4964), ('classroom', 4703), ('the', 3748), ('they', 3641), ('not', 3620), ('learn', 3535)]

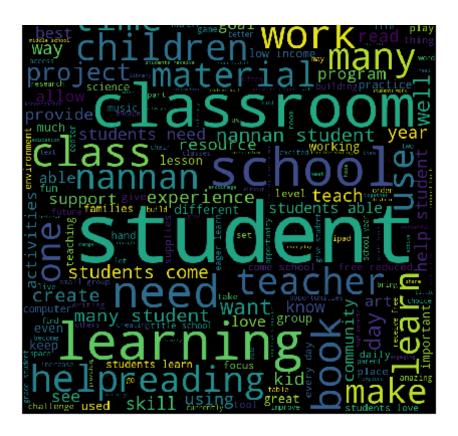
Word cloud plot for essay text for cluster 2 for Agglomerative algorith m



Total number of words in cluster 2 are 583018

Top 10 most common words in cluster 2 are [('students', 27791), ('i', 1 1746), ('school', 8997), ('my', 6507), ('learning', 6264), ('classroo m', 6069), ('the', 4761), ('not', 4664), ('they', 4655), ('learn', 442 8)]

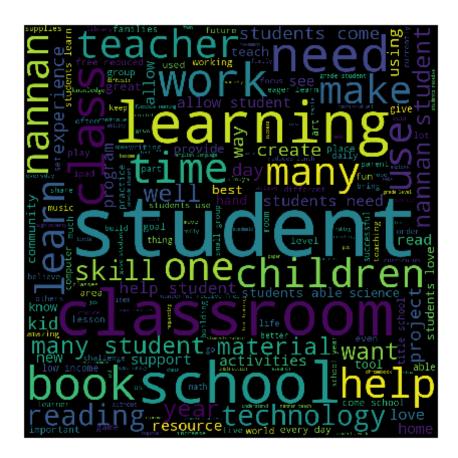
Word cloud plot for essay text for cluster 3 for Agglomerative algorith m



Total number of words in cluster 3 are 222219

Top 10 most common words in cluster 3 are [('students', 10757), ('i', 4587), ('school', 3438), ('my', 2445), ('learning', 2409), ('classroom', 2305), ('the', 1840), ('not', 1808), ('learn', 1723), ('they', 1676)]

Word cloud plot for essay text for cluster 4 for Agglomerative algorith m

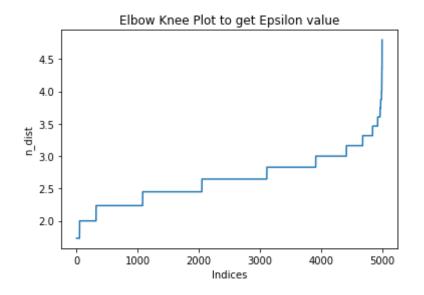


Total number of words in cluster 4 are 275590

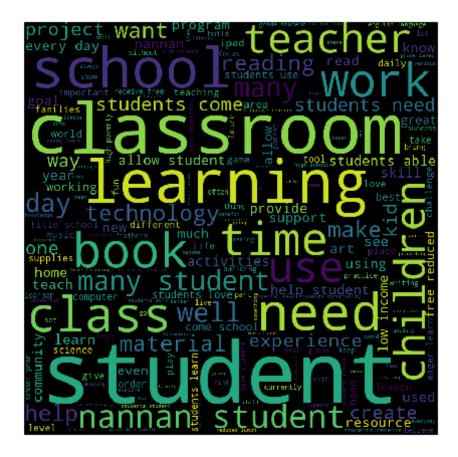
Top 10 most common words in cluster 4 are [('students', 13248), ('i', 5 679), ('school', 4349), ('my', 3160), ('learning', 2903), ('classroom', 2888), ('the', 2241), ('learn', 2207), ('they', 2150), ('help', 2125)]

2.7 Apply DBSCAN

```
In [231]: # https://stackoverflow.com/questions/12893492/choosing-eps-and-minpts-
          for-dbscan-r/48558030#48558030
          # https://scikit-learn.org/stable/modules/generated/sklearn.cluster.DBS
          CAN. html
          # https://scikit-learn.org/stable/modules/generated/sklearn.neighbors.N
          earestNeighbors.html
          # https://ipfs-sec.stackexchange.cloudflare-ipfs.com/datascience/A/ques
          tion/10162.html
          X \text{ tr new} = X \text{ tr new}[:5000]
          min pts=10
          n dist=[]
          distance = []
          for val in X tr new:
              val dist = np.sort(np.sum((X tr new.toarray()-val.toarray())**2, ax
          is=1), axis=None)
              distance.append(val dist[min pts])
          n dist = np.sort(np.sqrt(np.array(distance)))
          indices = [ind for ind in range(X tr new.shape[0])]
          plt.plot(indices, n_dist)
          plt.xlabel('Indices')
          plt.ylabel('n dist')
          plt.title('Elbow Knee Plot to get Epsilon value')
          plt.show()
```



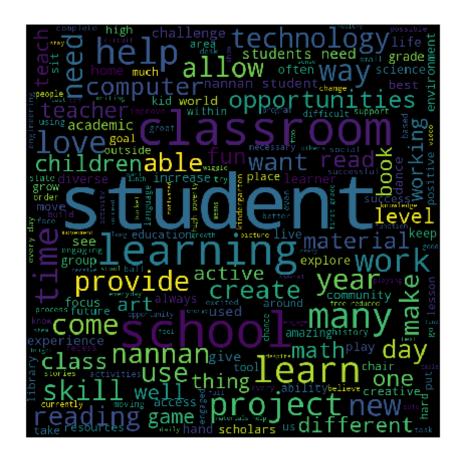
```
els else 0)
          n noise = list(clustering.labels ).count(-1)
          print('Estimated number of clusters: %d' % n clusters )
          print('Estimated number of noise points: %d' % n noise )
          Estimated number of clusters: 4
          Estimated number of noise points: 2177
In [286]: # Creating different clusters
          cluster 0 = []
          cluster 1 = []
          cluster 2 = []
          cluster 3 = []
          for val in range(0, len(clustering.labels )):
              if (clustering.labels [val] == 0):
                  cluster 0.append(preprocessed essays[val])
              if (clustering.labels [val] == 1):
                  cluster 1.append(preprocessed essays[val])
              if (clustering.labels [val] == 2):
                  cluster 2.append(preprocessed essays[val])
              if (clustering.labels [val] == 3):
                  cluster 3.append(preprocessed essays[val])
In [288]: cluster = [cluster 0, cluster 1, cluster 2, cluster 3]
          count = 0
          for val in cluster:
              create wordcloud(val, count, "DBSCAN")
              wordcloud analysis(val, count)
              count +=1
```



Total number of words in cluster 0 are 420306

Top 10 most common words in cluster 0 are [('students', 20005), ('i', 8370), ('school', 6655), ('my', 4675), ('classroom', 4486), ('learnin g', 4460), ('the', 3532), ('they', 3340), ('not', 3228), ('learn', 31 76)]

Word cloud plot for essay text for cluster 1 for DBSCAN algorithm



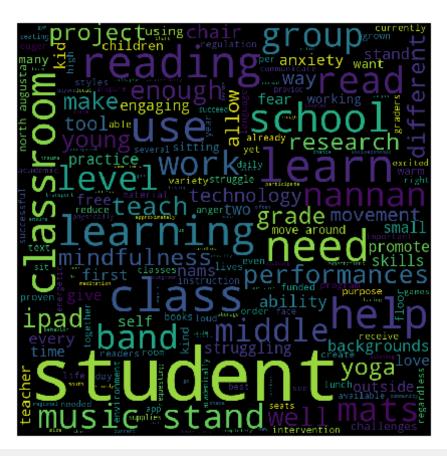
Total number of words in cluster 1 are 7191

Top 10 most common words in cluster 1 are [('students', 341), ('i', 1 43), ('school', 98), ('classroom', 81), ('my', 80), ('learning', 69), ('they', 65), ('the', 62), ('not', 55), ('learn', 55)]

Word cloud plot for essay text for cluster 2 for DBSCAN algorithm

Total number of words in cluster 2 are 593

Top 10 most common words in cluster 2 are [('students', 31), ('help', 11), ('technology', 10), ('learn', 10), ('allow', 9), ('the', 8), ('closed', 2), ('school', 2), ('comp', 3), ('mapp', 6)]



```
Total number of words in cluster 3 are 931

Top 10 most common words in cluster 3 are [('students', 43), ('i', 3 1), ('class', 11), ('learn', 11), ('reading', 10), ('learning', 9), ('use', 9), ('need', 9), ('help', 9), ('school', 8)]
```

3. Conclusions

Please write down few lines of your observations on this assignment.

```
In [19]: # http://zetcode.com/python/prettytable/
        from prettytable import PrettyTable
        x = PrettyTable()
        x.field names = ["Model Name", "Hyperparameters", "Hyperparameter Value
         1", "Hyperparameter Value 2", "Number of Clusters"]
        x.add_row(["Kmeans", "n_clusters/max_iter", 8, 300, 8])
        x.add_row(["AgglomerativeClustering", "n_clusters", 2, "na", 2])
        x.add row(["AgglomerativeClustering", "n clusters", 5, "na", 5])
        x.add row(["DBSCAN", "minpts/eps", 10, 2.3, 4])
        print(x)
                              | Hyperparameters
                 Model Name
                                                    | Hyperparameter Value
        1 | Hyperparameter Value 2 | Number of Clusters |
                   Kmeans | n clusters/max iter |
                    300
         | AgglomerativeClustering | n clusters
                                                                 2
                     na
```

+	Agg 	lomerativeCluster na DBSCAN 2.3	ring 	n_clusters 5 minpts/eps 4		5 10
	+	2.3		4	 +	