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:

Desc	Feature
A unique identifier for the proposed project. Example: p0	project_id
Title of the project. Exa	
• Art Will Make You H • First Grad	project_title
Grade level of students for which the project is targeted. One of the fo enumerated $\boldsymbol{\nu}$	
 Grades P Grade Grade Grades 	project_grade_category

Feature

DC30	1 Catalo
One or more (comma-separated) subject categories for the project fr following enumerated list of v	
 Applied Lea	project_subject_categories
Exan	
 Music & The Literacy & Language, Math & Sc 	
State where school is located (<u>Two-letter U.S. posta (https://en.wikipedia.org/wiki/List_of_U.Sstate_abbreviations#Postal_c_Example</u>	school_state
One or more (comma-separated) subject subcategories for the parameters.	
• Lit	<pre>project_subject_subcategories</pre>
• Literature & Writing, Social Sci	
An explanation of the resources needed for the project. Exa	
• My students need hands on literacy materials to make sensory needs!<	<pre>project_resource_summary</pre>
First application	project_essay_1
Second application	project_essay_2
Third application	project_essay_3
Fourth application	project_essay_4
Datetime when project application was submitted. Example: 2016-0 12:43:5	<pre>project_submitted_datetime</pre>
A unique identifier for the teacher of the proposed project. Ex ibdf8baa8fedef6bfeec7ae4ff1c	teacher_id
Teacher's title. One of the following enumerated $\boldsymbol{\nu}$	
• • • • • Tea	teacher_prefix
Number of project applications previously submitted by the same to Exam	teacher_number_of_previously_posted_projects

^{*} 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

Desc

quantity Quantity of the resource required. Example	

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

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

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

Notes on the Essay Data

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

- project_essay_1: "Introduce us to your classroom"
- project_essay_2: "Tell us more about your students"
- project_essay_3: "Describe how your students will use the materials you're requesting"
- project_essay_4: "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 [3]:

```
%matplotlib inline
import warnings
warnings.filterwarnings("ignore")
import sqlite3
import pandas as pd
import numpy as np
import nltk
import string
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.feature_extraction.text import TfidfTransformer
from sklearn.feature_extraction.text import TfidfVectorizer
from sklearn.feature_extraction.text import CountVectorizer
from sklearn.metrics import confusion matrix
from sklearn import metrics
from sklearn.metrics import roc_curve, auc
from nltk.stem.porter import PorterStemmer
import re
# Tutorial about Python regular expressions: https://pymotw.com/2/re/
import string
from nltk.corpus import stopwords
from nltk.stem import PorterStemmer
from nltk.stem.wordnet import WordNetLemmatizer
from gensim.models import Word2Vec
from gensim.models import KeyedVectors
import pickle
from tqdm import tqdm
import os
from plotly import plotly
import plotly.offline as offline
import plotly.graph_objs as go
offline.init_notebook_mode()
from collections import Counter
```

1.1 Reading Data

```
In [4]:
```

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

In [5]:

```
print("Number of data points in train data", project_data.shape)
print('-'*50)
print("The attributes of data :", project_data.columns.values)
Number of data points in train data (109248, 17)
```

The attributes of data: ['Unnamed: 0' 'id' 'teacher_id' 'teacher_prefix' 's chool_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 [6]:

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

Number of data points in train data (1541272, 4) ['id' 'description' 'quantity' 'price']

Out[6]:

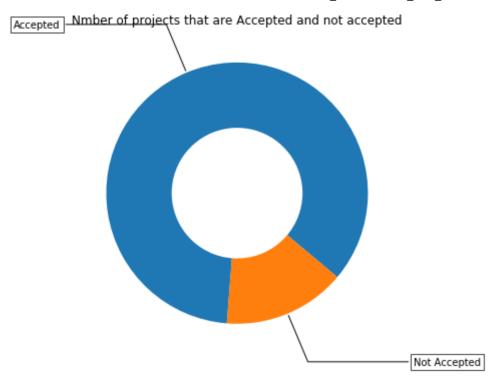
	id	description	quantity	price
0	p233245	LC652 - Lakeshore Double-Space Mobile Drying Rack	1	149.00
1	p069063	Bouncy Bands for Desks (Blue support pipes)	3	14.95
2	p069063	Cory Stories: A Kid's Book About Living With Adhd	1	8.45
3	p069063	Dixon Ticonderoga Wood-Cased #2 HB Pencils, Bo	2	13.59
4	p069063	EDUCATIONAL INSIGHTS FLUORESCENT LIGHT FILTERS	3	24.95

1.2 Data Analysis

In [7]:

```
# PROVIDE CITATIONS TO YOUR CODE IF YOU TAKE IT FROM ANOTHER WEBSITE.
# https://matplotlib.org/gallery/pie_and_polar_charts/pie_and_donut_labels.html#sphx-glr-gd
y_value_counts = project_data['project_is_approved'].value_counts()
print("Number of projects that are approved for funding ", y_value_counts[1], ", (", (y_val
print("Number of projects that are not approved for funding ", y_value_counts[0], '
fig, ax = plt.subplots(figsize=(6, 6), subplot_kw=dict(aspect="equal"))
recipe = ["Accepted", "Not Accepted"]
data = [y_value_counts[1], y_value_counts[0]]
wedges, texts = ax.pie(data, wedgeprops=dict(width=0.5), startangle=-40)
bbox_props = dict(boxstyle="square,pad=0.3", fc="w", ec="k", lw=0.72)
kw = dict(xycoords='data', textcoords='data', arrowprops=dict(arrowstyle="-"),
          bbox=bbox_props, zorder=0, va="center")
for i, p in enumerate(wedges):
    ang = (p.theta2 - p.theta1)/2. + p.theta1
    y = np.sin(np.deg2rad(ang))
    x = np.cos(np.deg2rad(ang))
    horizontalalignment = {-1: "right", 1: "left"}[int(np.sign(x))]
    connectionstyle = "angle, angleA=0, angleB={}".format(ang)
    kw["arrowprops"].update({"connectionstyle": connectionstyle})
    ax.annotate(recipe[i], xy=(x, y), xytext=(1.35*np.sign(x), 1.4*y),
                 horizontalalignment=horizontalalignment, **kw)
ax.set_title("Nmber of projects that are Accepted and not accepted")
plt.show()
```

Number of projects that are approved for funding 92706, (84.85830404217927) Number of projects that are not approved for funding 16542, (15.141695957820739%)



Observation -

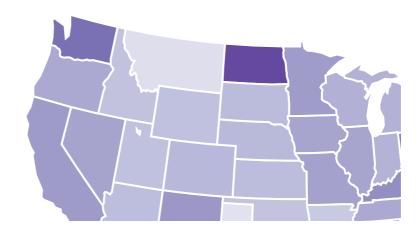
1. Around 85% of Projects are Approved for Funding & rest 15% of Projects are not Approved for Funding.

1.2.1 Univariate Analysis: School State

In [8]:

```
# Pandas dataframe groupby count, mean: https://stackoverflow.com/a/19385591/4084039
temp = pd.DataFrame(project_data.groupby("school_state")["project_is_approved"].apply(np.me
# if you have data which contain only 0 and 1, then the mean = percentage (think about it)
temp.columns = ['state_code', 'num_proposals']
# How to plot US state heatmap: https://datascience.stackexchange.com/a/9620
scl = [[0.0, 'rgb(242,240,247)'], [0.2, 'rgb(218,218,235)'], [0.4, 'rgb(188,189,220)'], \]
            [0.6, 'rgb(158,154,200)'], [0.8, 'rgb(117,107,177)'], [1.0, 'rgb(84,39,143)']]
data = [ dict(
        type='choropleth',
        colorscale = scl,
        autocolorscale = False,
        locations = temp['state_code'],
        z = temp['num_proposals'].astype(float),
        locationmode = 'USA-states',
        text = temp['state_code'],
        marker = dict(line = dict (color = 'rgb(255,255,255)', width = 2)),
        colorbar = dict(title = "% of pro")
    ) ]
layout = dict(
        title = 'Project Proposals % of Acceptance Rate by US States',
        geo = dict(
            scope='usa',
            projection=dict( type='albers usa' ),
            showlakes = True,
            lakecolor = 'rgb(255, 255, 255)',
        ),
    )
fig = go.Figure(data=data, layout=layout)
offline.iplot(fig, filename='us-map-heat-map')
```

Project Proposals % of Acceptance Rat



In [9]:

```
# https://www.csi.cuny.edu/sites/default/files/pdf/administration/ops/2letterstabbrev.pdf
temp.sort_values(by=['num_proposals'], inplace=True)
print("States with lowest % approvals")
print(temp.head(5))
print('='*50)
print("States with highest % approvals")
print(temp.tail(5))
```

```
States with lowest % approvals
   state_code num_proposals
46
           VT
                     0.800000
7
           DC
                     0.802326
43
           TX
                     0.813142
26
           MΤ
                     0.816327
                     0.831245
18
           ΙΔ
```

States with highest % approvals

```
state_code num_proposals
30
           NH
                     0.873563
35
           OH
                     0.875152
47
           WΑ
                     0.876178
28
           ND
                     0.888112
           DE
                     0.897959
8
```

In [10]:

```
#stacked bar plots matplotlib: https://matplotlib.org/gallery/lines_bars_and_markers/bar_st
def stack_plot(data, xtick, col2='project_is_approved', col3='total'):
    ind = np.arange(data.shape[0])

plt.figure(figsize=(20,5))
    p1 = plt.bar(ind, data[col3].values)
    p2 = plt.bar(ind, data[col2].values)

plt.ylabel('Projects')
    plt.title('Number of projects aproved vs rejected')
    plt.xticks(ind, list(data[xtick].values))
    plt.legend((p1[0], p2[0]), ('total', 'accepted'))
    plt.show()
```

In [11]:

```
def univariate_barplots(data, col1, col2='project_is_approved', top=False):
    # Count number of zeros in dataframe python: https://stackoverflow.com/a/51540521/40840
    temp = pd.DataFrame(project_data.groupby(col1)[col2].agg(lambda x: x.eq(1).sum())).rese

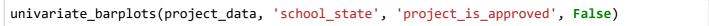
# Pandas dataframe grouby count: https://stackoverflow.com/a/19385591/4084039
    temp['total'] = pd.DataFrame(project_data.groupby(col1)[col2].agg({'total':'count'})).r
    temp['Avg'] = pd.DataFrame(project_data.groupby(col1)[col2].agg({'Avg':'mean'})).reset_

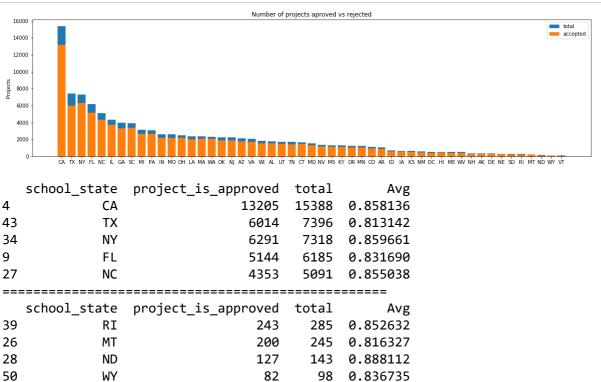
temp.sort_values(by=['total'],inplace=True, ascending=False)

if top:
    temp = temp[0:top]

stack_plot(temp, xtick=col1, col2=col2, col3='total')
    print(temp.head(5))
    print("="*50)
    print(temp.tail(5))
```

In [12]:





Observation -

VT

46

- 1. Every state has greater than 80% success rate in approval.
- 2. California(CA) has largest approval rate, while Vermont(VT) has lowest approval rate.

64

80

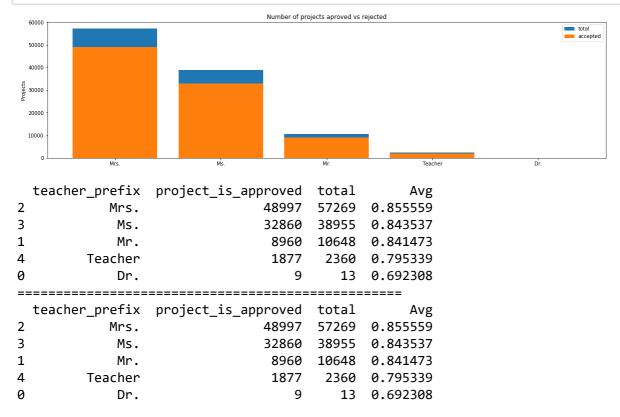
0.800000

3. There is Sudden drop in the Plot for rate of Approval of Projects.

1.2.2 Univariate Analysis: teacher_prefix

In [13]:





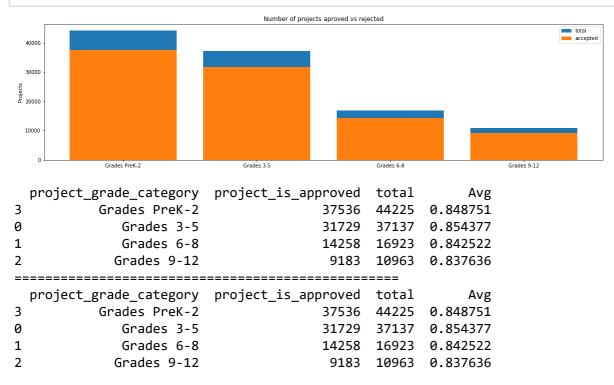
Observation -

- 1. Project by the Teacher who has prefix "Mrs." has the 85% chance of being Approved (Which is the Highest Approval Rate).
- 2. While Project by the Teacher who has prefix "Dr." has only 69% chance of being Approved (Which is Lowest Approval Rate).
- 3. There is lot of projects submissions for Teacher with prefix "Mrs", "Mr", "Ms' and also have high Approval rates than the others

1.2.3 Univariate Analysis: project_grade_category

In [14]:

univariate_barplots(project_data, 'project_grade_category', 'project_is_approved', top=Fals



Observation -

- 1. For All the Grades, Average rate of Projects being Approved is more than 80%.
- 2. Lower Grades has more no. of Project submitted while Higher Grades has less no. of Project submitted.

1.2.4 Univariate Analysis: project_subject_categories

In [15]:

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

In [16]:

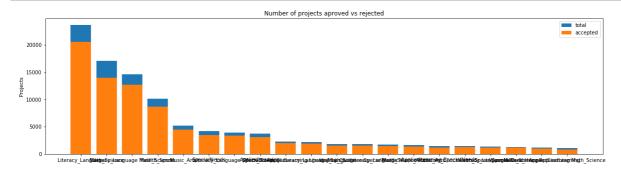
```
project_data['clean_categories'] = cat_list
project_data.drop(['project_subject_categories'], axis=1, inplace=True)
project_data.head(2)
```

Out[16]:

	Unnamed: 0	id	teacher_id	teacher_prefix	school_state	project
0	160221	p253737	c90749f5d961ff158d4b4d1e7dc665fc	Mrs.	IN	
1	140945	p258326	897464ce9ddc600bced1151f324dd63a	Mr.	FL	
4						•

In [17]:

univariate_barplots(project_data, 'clean_categories', 'project_is_approved', top=20)



	clean_categories	project_is_approved	total	Avg
24	Literacy_Language	20520	23655	0.867470
32	Math_Science	13991	17072	0.819529
28	Literacy_Language Math_Science	12725	14636	0.869432
8	Health_Sports	8640	10177	0.848973
40	Music_Arts	4429	5180	0.855019
===		=========		
				_

	clean_categories	<pre>project_is_approved</pre>	total	Avg
19	<pre>History_Civics Literacy_Language</pre>	1271	1421	0.894441
14	Health_Sports SpecialNeeds	1215	1391	0.873472
50	Warmth Care_Hunger	1212	1309	0.925898
33	Math_Science AppliedLearning	1019	1220	0.835246
4	AppliedLearning Math_Science	855	1052	0.812738

Observation -

- 1. For All different Subjects, Average rate of Projects being Approved is more than 80%.
- 2. When Math_Science is combined with Literacy_Language as a Single Subject, Project Approval rate is Highest i.e 86% & Lowest when combined with AppliedLearning i.e 81%.
- 3. Subject Warmth Care_Hunger has the Highest Approval rate i.e 92%.

In [18]:

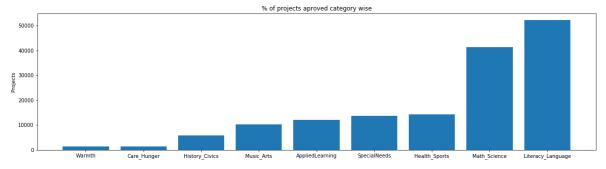
```
# count of all the words in corpus python: https://stackoverflow.com/a/22898595/4084039
from collections import Counter
my_counter = Counter()
for word in project_data['clean_categories'].values:
    my_counter.update(word.split())
```

In [19]:

```
# dict sort by value python: https://stackoverflow.com/a/613218/4084039
cat_dict = dict(my_counter)
sorted_cat_dict = dict(sorted(cat_dict.items(), key=lambda kv: kv[1]))

ind = np.arange(len(sorted_cat_dict))
plt.figure(figsize=(20,5))
p1 = plt.bar(ind, list(sorted_cat_dict.values()))

plt.ylabel('Projects')
plt.title('% of projects aproved category wise')
plt.xticks(ind, list(sorted_cat_dict.keys()))
plt.show()
```



In [20]:

```
for i, j in sorted_cat_dict.items():
    print("{:20} :{:10}".format(i,j))
```

1388 Warmth Care Hunger 1388 History_Civics 5914 Music Arts 10293 12135 AppliedLearning SpecialNeeds 13642 : Health Sports 14223 Math Science 41421 Literacy_Language 52239

Observation -

- 1. For Subject "Literacy_Language" maximum no. of Projects are being Submitted while for Subject "Warmth" lowest no. of Projects are being Submitted.
- 2. For each Subject there is High Variability among the Projects Submitted.

1.2.5 Univariate Analysis: project_subject_subcategories

In [21]:

```
sub catogories = list(project data['project subject subcategories'].values)
# remove special characters from list of strings python: https://stackoverflow.com/a/473019
# https://www.geeksforgeeks.org/removing-stop-words-nltk-python/
# https://stackoverflow.com/questions/23669024/how-to-strip-a-specific-word-from-a-string
# https://stackoverflow.com/questions/8270092/remove-all-whitespace-in-a-string-in-python
sub_cat_list = []
for i in sub_catogories:
   temp = ""
    # consider we have text like this "Math & Science, Warmth, Care & Hunger"
    for j in i.split(','): # it will split it in three parts ["Math & Science", "Warmth",
        if 'The' in j.split(): # this will split each of the catogory based on space "Math
            j=j.replace('The','') # if we have the words "The" we are going to replace it w
                         ','') # we are placeing all the ' '(space) with ''(empty) ex:"Math
        j = j.replace('
        temp +=j.strip()+" "#" abc ".strip() will return "abc", remove the trailing spaces
        temp = temp.replace('&','_')
    sub cat list.append(temp.strip())
```

In [22]:

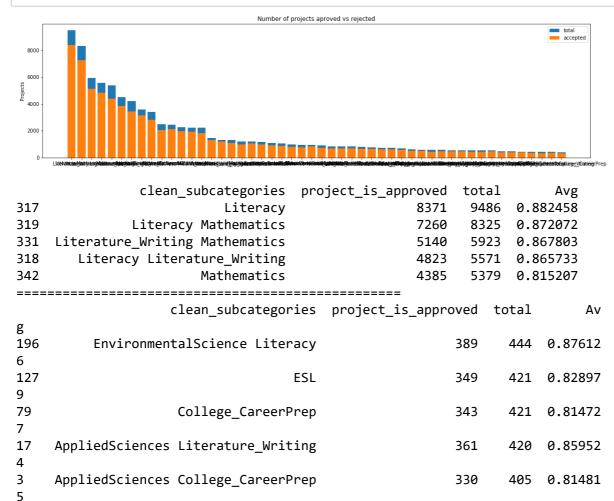
```
project_data['clean_subcategories'] = sub_cat_list
project_data.drop(['project_subject_subcategories'], axis=1, inplace=True)
project_data.head(2)
```

Out[22]:

	Unnamed: 0	id	teacher_id	teacher_prefix	school_state	project _.
0	160221	p253737	c90749f5d961ff158d4b4d1e7dc665fc	Mrs.	IN	
1	140945	p258326	897464ce9ddc600bced1151f324dd63a	Mr.	FL	
4						N .

In [23]:

univariate_barplots(project_data, 'clean_subcategories', 'project_is_approved', top=50)



Observation -

- 1. For All Subcategories, Average rate of Projects being Approved is above 80%.
- 2. The Subject sub_category "Literacy" has the highest Project approval rate i.e 88%

In [24]:

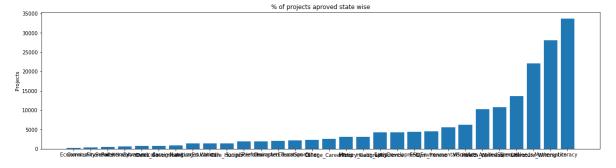
```
# count of all the words in corpus python: https://stackoverflow.com/a/22898595/4084039
from collections import Counter
my_counter = Counter()
for word in project_data['clean_subcategories'].values:
    my_counter.update(word.split())
```

In [25]:

```
# dict sort by value python: https://stackoverflow.com/a/613218/4084039
sub_cat_dict = dict(my_counter)
sorted_sub_cat_dict = dict(sorted(sub_cat_dict.items(), key=lambda kv: kv[1]))

ind = np.arange(len(sorted_sub_cat_dict))
plt.figure(figsize=(20,5))
p1 = plt.bar(ind, list(sorted_sub_cat_dict.values()))

plt.ylabel('Projects')
plt.title('% of projects aproved state wise')
plt.xticks(ind, list(sorted_sub_cat_dict.keys()))
plt.show()
```



In [26]:

```
for i, j in sorted_sub_cat_dict.items():
    print("{:20} :{:10}".format(i,j))
```

Economics 269 CommunityService 441 568 FinancialLiteracy ParentInvolvement 677 Extracurricular 810 Civics_Government 815 ForeignLanguages 890 NutritionEducation : 1355 Warmth 1388 Care_Hunger 1388 SocialSciences 1920 PerformingArts 1961 CharacterEducation : 2065 TeamSports 2192 **Other** 2372 College_CareerPrep 2568 Music 3145 History_Geography 3171 Health_LifeScience 4235 EarlyDevelopment 4254 **ESL** 4367 Gym Fitness 4509 EnvironmentalScience : 5591 VisualArts 6278 Health_Wellness :
AppliedSciences :
SpecialNeeds : 10234 10816 13642 Literature_Writing : 22179 28074 Mathematics Literacy 33700

Observation -

- 1. "Literacy" has the highest no. of Project Submission i.e. 33700 & "Economics" has Lowest i.e. 269.
- 2. There is Variation in No. of Projects Submitted among all.

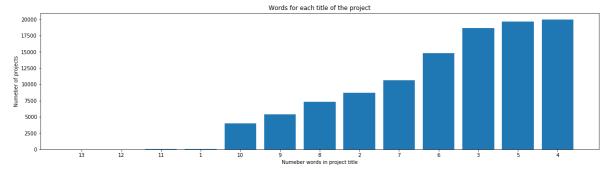
1.2.6 Univariate Analysis: Text features (Title)

In [27]:

```
#How to calculate number of words in a string in DataFrame: https://stackoverflow.com/a/374
word_count = project_data['project_title'].str.split().apply(len).value_counts()
word_dict = dict(word_count)
word_dict = dict(sorted(word_dict.items(), key=lambda kv: kv[1]))

ind = np.arange(len(word_dict))
plt.figure(figsize=(20,5))
p1 = plt.bar(ind, list(word_dict.values()))

plt.ylabel('Numeber of projects')
plt.xlabel('Numeber words in project title')
plt.title('Words for each title of the project')
plt.xticks(ind, list(word_dict.keys()))
plt.show()
```



Observation -

- 1. The Higher no. of Projects are there which has 3,4 or 5 words in their Title.
- 2. Projects with 13 words in their Title are Very Less in Number.

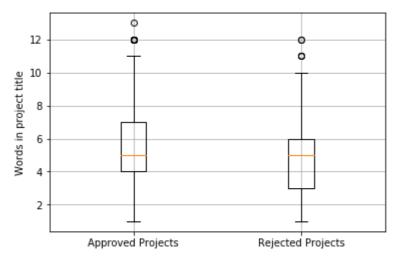
In [28]:

```
approved_title_word_count = project_data[project_data['project_is_approved']==1]['project_t
approved_title_word_count = approved_title_word_count.values

rejected_title_word_count = project_data[project_data['project_is_approved']==0]['project_t
rejected_title_word_count = rejected_title_word_count.values
```

In [29]:

```
# https://glowingpython.blogspot.com/2012/09/boxplot-with-matplotlib.html
plt.boxplot([approved_title_word_count, rejected_title_word_count])
plt.xticks([1,2],('Approved Projects','Rejected Projects'))
plt.ylabel('Words in project title')
plt.grid()
plt.show()
```

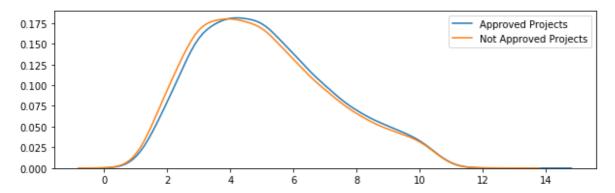


Observation -

- 1. Medians are almost same for both "Approved Projects" and "Rejected Projects".
- 2. Number of Words in the title for the "Approved Projects" are slightly more.
- 3. The minimum number of words are same for both "Approved Projects" and "Rejected Projects" but the maximum number of words in case of "Approved Projects" are slightly more.

In [30]:

```
plt.figure(figsize=(10,3))
sns.kdeplot(approved_title_word_count,label="Approved Projects", bw=0.6)
sns.kdeplot(rejected_title_word_count,label="Not Approved Projects", bw=0.6)
plt.legend()
plt.show()
```



Observation -

1. From the 2 PDF's we can see that PDF for "Approved Projects" is Slightly more than "Not Approved Projects" i.e. No. of words in Title for "Approved Projects" is More.

1.2.7 Univariate Analysis: Text features (Project Essay's)

In [31]:

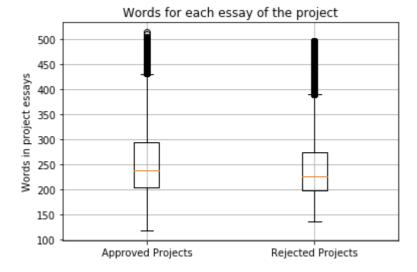
In [32]:

```
approved_word_count = project_data[project_data['project_is_approved']==1]['essay'].str.spl
approved_word_count = approved_word_count.values

rejected_word_count = project_data[project_data['project_is_approved']==0]['essay'].str.spl
rejected_word_count = rejected_word_count.values
```

In [33]:

```
# https://glowingpython.blogspot.com/2012/09/boxplot-with-matplotlib.html
plt.boxplot([approved_word_count, rejected_word_count])
plt.title('Words for each essay of the project')
plt.xticks([1,2],('Approved Projects','Rejected Projects'))
plt.ylabel('Words in project essays')
plt.grid()
plt.show()
```

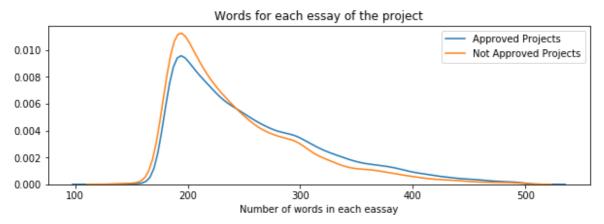


Observation -

- 1. Medians are almost same for both "Approved Projects" and "Rejected Projects".
- 2. For "Approved Projects" no. of words in the Essay are slightly more.
- 3. For "Approve Projects" distribution is also slightly more.

In [34]:

```
plt.figure(figsize=(10,3))
sns.distplot(approved_word_count, hist=False, label="Approved Projects")
sns.distplot(rejected_word_count, hist=False, label="Not Approved Projects")
plt.title('Words for each essay of the project')
plt.xlabel('Number of words in each eassay')
plt.legend()
plt.show()
```



Observation -

1. From the 2 PDF's we can see that PDF for "Approved Projects" is Slightly more than "Not Approved Projects" i.e. No. of words in Each essay for "Approved Projects" is More.

1.2.8 Univariate Analysis: Cost per project

In [35]:

```
# we get the cost of the project using resource.csv file
resource_data.head(2)
```

Out[35]:

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

In [36]:

```
# https://stackoverflow.com/questions/22407798/how-to-reset-a-dataframes-indexes-for-all-gr
price_data = resource_data.groupby('id').agg({'price':'sum', 'quantity':'sum'}).reset_index
price_data.head(2)
```

Out[36]:

	id	price	quantity
0	p000001	459.56	7
1	p000002	515.89	21

In [37]:

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

In [38]:

```
approved_price = project_data[project_data['project_is_approved']==1]['price'].values
rejected_price = project_data[project_data['project_is_approved']==0]['price'].values
```

In [39]:

```
# https://glowingpython.blogspot.com/2012/09/boxplot-with-matplotlib.html
plt.boxplot([approved_price, rejected_price])
plt.title('Box Plots of Cost per approved and not approved Projects')
plt.xticks([1,2],('Approved Projects','Rejected Projects'))
plt.ylabel('Price')
plt.grid()
plt.show()
```

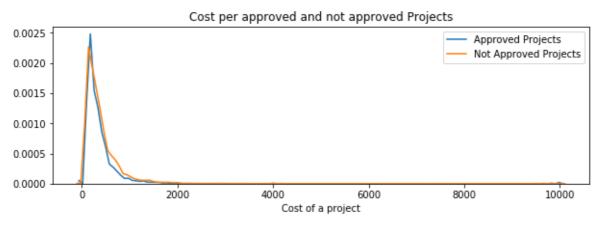


Observation -

1. All points in the box plots are overlapping therefore it does not make any sense & we can't conclude anything.

In [40]:

```
plt.figure(figsize=(10,3))
sns.distplot(approved_price, hist=False, label="Approved Projects")
sns.distplot(rejected_price, hist=False, label="Not Approved Projects")
plt.title('Cost per approved and not approved Projects')
plt.xlabel('Cost of a project')
plt.legend()
plt.show()
```



Observation -

1. We can notice that Both the PDF's are overlapping but Cost per "Not Approved Projects" is slightly more than Cost per "Approved Projects".

In [41]:

```
# http://zetcode.com/python/prettytable/
from prettytable import PrettyTable

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

x = PrettyTable()
x.field_names = ["Percentile", "Approved Projects", "Not Approved Projects"]

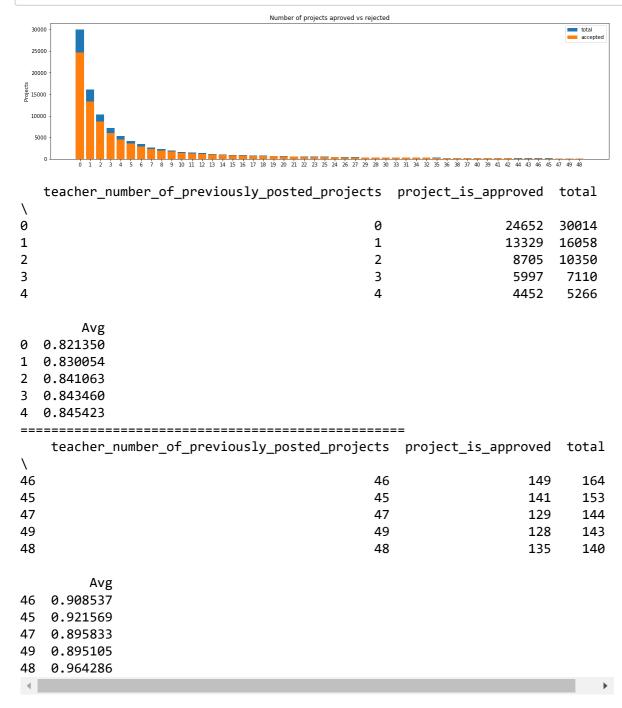
for i in range(0,101,5):
    x.add_row([i,np.round(np.percentile(approved_price,i), 3), np.round(np.percentile(rejector));
    results the provided price of the price of the price of the provided price of the price of the price of the provided price of the price of th
```

Percentile	+ Approved Projects	Not Approved Projects		
+	+	+		
0	0.66	1.97		
5	13.59	41.9		
10	33.88	73.67		
15	58.0	99.109		
20	77.38	118.56		
25	99.95	140.892		
30	116.68	162.23		
35	137.232	184.014		
40	157.0	208.632		
45	178.265	235.106		
50	198.99	263.145		
55	223.99	292.61		
60	255.63	325.144		
65	285.412	362.39		
70	321.225	399.99		
75	366.075	449.945		
80	411.67	519.282		
85	479.0	618.276		
90	593.11	739.356		
95	801.598	992.486		
100	9999.0	9999.0		

1.2.9 Univariate Analysis: teacher_number_of_previously_posted_projects

In [42]:

univariate_barplots(project_data, 'teacher_number_of_previously_posted_projects', 'project_



Observation -

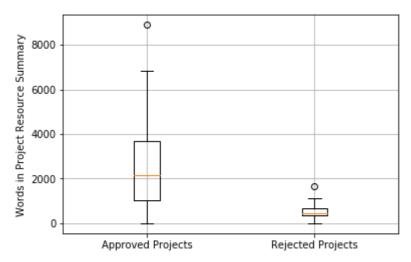
- Teacher who posted less no. of projects before or has never posted project before has high no. of projects
 High approval rate.
- 2. As Teacher previously posted projects increases, No, pf Projects & Approval rate decreases.
- 3. Average Approval rate is Higher where Teacher previously posted projects is High.

1.2.10 Univariate Analysis: project_resource_summary

In [43]:

```
approved_summ_word_count = project_data[project_data['project_is_approved']==1]['project_re
approved_summ_word_count = approved_summ_word_count.value_counts()
rejected_summ_word_count = project_data[project_data['project_is_approved']==0]['project_re
rejected_summ_word_count = rejected_summ_word_count.value_counts()

plt.boxplot([approved_summ_word_count, rejected_summ_word_count])
plt.xticks([1,2],('Approved Projects','Rejected Projects'))
plt.ylabel('Words in Project Resource Summary')
plt.grid()
plt.show()
```



Observation -

- 1. Median for "Approved Projects" is Higher than the "Rejected Projects".
- 2. For "Approved Projects" no. of Words in the Title are more than "rejected Projects".
- 3. Distribution of "Rejected Project" is approx. 20% of Distribution of "Approved Projects" hence we can conclude that No. of word in the Title of "Approved Projects" is more that "Rejected Project".

1.3 Text preprocessing

1.3.1 Essay Text

```
In [44]:
```

project_data.head(2)

Out[44]:

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

In [45]:

```
# printing some random essays.
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)
print(project_data['essay'].values[99999])
print(project_data['essay'].values[99999])
print("="*50)
```

My students are English learners that are working on English as their second 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 hav e over 24 languages represented in our English Learner program with students at every level of mastery. We also have over 40 countries represented with the families within our school. Each student brings a wealth of knowledge a nd experiences to us that open our eyes to new cultures, beliefs, and respec t.\"The limits of your language are the limits of your world.\"-Ludwig Wittg enstein Our English learner's have a strong support system at home that beg s for more resources. Many times our parents are learning to read and speak English along side of their children. Sometimes this creates barriers for p arents to be able to help their child learn phonetics, letter recognition, a nd other reading skills.\r\n\r\nBy providing these dvd's and players, studen ts are 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 pr oficiency status, will be a offered to be a part of this program. These edu cational videos will be specially chosen by the English Learner Teacher and will be sent home regularly to watch. The videos are to help the child deve lop early reading skills.\r\n\r\nParents that do not have access to a dvd pl ayer will have the opportunity to check out a dvd player to use for the yea r. The plan is to use these videos and educational dvd's for the years to c ome for other EL students.\r\nnannan

The 51 fifth grade students that will cycle through my classroom this year a 11 love learning, at least most of the time. At our school, 97.3% of the stu dents receive free or reduced price lunch. Of the 560 students, 97.3% are mi nority students. \r\nThe school has a vibrant community that loves to get to gether and celebrate. Around Halloween there is a whole school parade to sho w off the beautiful costumes that students wear. On Cinco de Mayo we put on a big festival with crafts made by the students, dances, and games. At the e nd of the year the school hosts a carnival 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 five brightly colored Hokki stools in place of regul ar, stationary, 4-legged chairs. As I will only have a total of ten in the c lassroom and not enough for each student to have an individual one, they wil 1 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 th em in place of chairs at my small group tables during math and reading time s. The rest of the day they will be used by the students who need the highes t amount of movement in their life in order to stay focused on school.\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. Whe n the students are sitting in group with me on the Hokki Stools, they are al ways moving, but at the same time doing their work. Anytime the students get to pick where they can sit, the Hokki Stools are the first to be taken. Ther e are always students who head over to the kidney table to get one of the st ools who are disappointed as there 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 comprom ise that allow my students to do desk work and move at the same time. These stools will help students to meet their 60 minutes a day of movement by allo wing them to activate their core muscles for balance while they sit. For man y of my students, these chairs will take away the barrier that exists in sch ools for a child who can't sit still.nannan

How do you remember your days of school? Was it in a sterile environment wit h 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 each day.\r\n\r\nMy cl ass is made up of 28 wonderfully unique boys and girls of mixed races in Ark ansas.\r\nThey attend a Title I school, which means there is a high enough p ercentage of free and reduced-price lunch to qualify. Our school is an \"ope n 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; th ey are like sponges, absorbing all the information and experiences and keep on wanting more. With these resources such as the comfy red throw pillows and the whimsical nautical hanging decor and the blue fish nets, I will be able to help create the mood in our classroom setting to be one of a themed nauti cal environment. Creating a classroom environment is very important in the s uccess in each and every child's education. The nautical photo props will 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 child with them, hav e them developed, and then hung in our classroom ready for their first day o f 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 cards to their team groups.\r\n\r\nYou r generous donations will help me to help make our classroom a fun, invitin g, learning environment from day one.\r\n\r\nIt costs lost of money out of m y own pocket on resources to get our classroom ready. Please consider helpin g with this project to make our new school year a very successful one. Thank you!nannan

My kindergarten students have varied disabilities ranging from speech and la nguage delays, cognitive delays, gross/fine motor delays, to autism. They ar e eager beavers and always strive to work their hardest working past their 1 imitations. \r\n\r\nThe materials we have are the ones I seek out for my stu dents. I teach in a Title I school where most of the students receive free o r reduced price lunch. Despite their disabilities and limitations, my stude nts 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 yo u were in a meeting? This is how my kids feel all the time. The want to be a ble 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 d on't want to sit and do worksheets. They want to learn to count by jumping a nd playing. Physical engagement is the key to our success. The number toss a nd color and shape mats can make that happen. My students will forget they a re doing work and just have the fun a 6 year old deserves.nannan

The mediocre teacher tells. The good teacher explains. The superior teacher demonstrates. The great teacher inspires. -William A. Ward\r\n\r\nMy school has 803 students which is makeup is 97.6% African-American, making up the la rgest segment of the student body. A typical school in Dallas is made up of 23.2% African-American students. Most of the students are on free or reduced lunch. We aren't receiving doctors, lawyers, or engineers children from rich backgrounds or neighborhoods. As an educator I am inspiring minds of young c hildren and we focus not only on academics but one smart, effective, efficie

nt, and disciplined students with good character. In our classroom we can uti lize the Bluetooth for swift transitions during class. I use a speaker which doesn't amplify the sound enough to receive the message. Due to the volume of my speaker my students can't hear videos or books clearly and it isn't making the lessons as meaningful. But with the bluetooth speaker my students will be able to hear and I can stop, pause and replay it at any time.\r\nThe cart will allow me to have more room for storage of things that are needed for the day and has an extra part to it I can use. The table top chart has all of the letter, words and pictures for students to learn about different letters and it is more accessible.nannan

In [46]:

```
# 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"\'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"\'t", " have", phrase)
    phrase = re.sub(r"\'ve", " have", phrase)
    phrase = re.sub(r"\'ve", " have", phrase)
    phrase = re.sub(r"\'m", " am", phrase)
    return phrase
```

In [47]:

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

My kindergarten students have varied disabilities ranging from speech and la nguage delays, cognitive delays, gross/fine motor delays, to autism. They ar e eager beavers and always strive to work their hardest working past their l imitations. \r\n\r\nThe materials we have are the ones I seek out for my stu dents. I teach in a Title I school where most of the students receive free o r reduced price lunch. Despite their disabilities and limitations, my stude nts 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 yo u were in a meeting? This is how my kids feel all the time. The want to be a ble 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 d o 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 _____

localhost:8888/notebooks/Desktop/Applied Machine Learning/Assignments/Assignment/2 DonorsChoose EDA TSNE.ipynb#

In [48]:

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

My kindergarten students have varied disabilities ranging from speech and la nguage delays, cognitive delays, gross/fine motor delays, to autism. They ar e eager beavers and always strive to work their hardest working past their 1 The materials we have are the ones I seek out for my student s. I teach in a Title I school where most of the students receive free or re duced 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 wer e in a meeting? This is how my kids feel all the time. The want to be able t o 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. They also want to learn through games, my kids do not w ant to sit and do worksheets. They want to learn to count by jumping and pla ying. Physical engagement is the key to our success. The number toss and col or and shape mats can make that happen. My students will forget they are doi ng work and just have the fun a 6 year old deserves.nannan

In [49]:

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

My kindergarten students have varied disabilities ranging from speech and la nguage delays cognitive delays gross fine motor delays to autism They are ea ger beavers and always strive to work their hardest working past their limit ations 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 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 a nts 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 le arn 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 They also want to learn through games my kids do not want to sit and do work sheets They want to learn to count by jumping and playing Physical engagemen t 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

In [50]:

In [51]:

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

```
100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%|
```

In [52]:

```
# after preprocesing
preprocessed_essays[20000]
```

Out[52]:

'my kindergarten students varied disabilities ranging speech language delays cognitive delays gross fine motor delays autism they eager beavers always st rive work hardest working past limitations the materials ones i seek student s i teach title i school students receive free reduced price lunch despite d isabilities limitations students love coming school come eager learn explore have ever felt like ants pants needed groove move meeting this kids feel tim e 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 engagemen t key success the number toss color shape mats make happen my students forge t work fun 6 year old deserves nannan'

1.3.2 Project title Text

In [53]:

```
# similarly we can preprocess the titles also

preprocessed_titles = []
# tqdm is for printing the status bar
for title in tqdm(project_data['project_title'].values):
    title = decontracted(title)
    title = title.replace('\\r', '')
    title = title.replace('\\r', '')
    title = title.replace('\\r', '')
    title = title.replace('\\r', '')
    title = re.sub('[^A-Za-z0-9]+', '', title)
    title = ''.join(e for e in title.split() if e not in stopwords)
    preprocessed_titles.append(title.lower().strip())
print(preprocessed_titles[23])
```

```
100%| 109248/109248 [00:03<00:0
0, 35938.55it/s]
```

instrumental power conquering steam

1. 4 Preparing data for models

```
In [54]:
```

```
- 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
- quantity : numerical
- teacher_number_of_previously_posted_projects : numerical
```

1.4.1 Vectorizing Categorical data

- price : numerical

https://www.appliedaicourse.com/course-online/lessons/handling-categorical-and-numerical-features/)

In [55]:

```
# we use count vectorizer to convert the values into one hot encoded features
from sklearn.feature_extraction.text import CountVectorizer
vectorizer = CountVectorizer(vocabulary=list(sorted_cat_dict.keys()), lowercase=False, bina
vectorizer.fit(project_data['clean_categories'].values)
print(vectorizer.get_feature_names())

categories_one_hot = vectorizer.transform(project_data['clean_categories'].values)
print("Shape of matrix after one hot encodig ",categories_one_hot.shape)

['Warmth', 'Care_Hunger', 'History_Civics', 'Music_Arts', 'AppliedLearning',
'SpecialNeeds', 'Health_Sports', 'Math_Science', 'Literacy_Language']
Shape of matrix after one hot encodig (109248, 9)
```

In [56]:

```
# we use count vectorizer to convert the values into one hot encoded features
vectorizer = CountVectorizer(vocabulary=list(sorted_sub_cat_dict.keys()), lowercase=False,
vectorizer.fit(project_data['clean_subcategories'].values)
print(vectorizer.get_feature_names())
sub_categories_one_hot = vectorizer.transform(project_data['clean_subcategories'].values)
print("Shape of matrix after one hot encodig ",sub_categories_one_hot.shape)
['Economics', 'CommunityService', 'FinancialLiteracy', 'ParentInvolvement',
'Extracurricular', 'Civics_Government', 'ForeignLanguages', 'NutritionEducat
ion', 'Warmth', 'Care_Hunger', 'SocialSciences', 'PerformingArts', 'Characte
rEducation', 'TeamSports', 'Other', 'College_CareerPrep', 'Music', 'History_
Geography', 'Health_LifeScience', 'EarlyDevelopment', 'ESL', 'Gym_Fitness',
'EnvironmentalScience', 'VisualArts', 'Health_Wellness', 'AppliedSciences',
'SpecialNeeds', 'Literature_Writing', 'Mathematics', 'Literacy']
Shape of matrix after one hot encodig (109248, 30)
In [57]:
# We can do the similar feature encoding with state
vectorizer = CountVectorizer(vocabulary=set(project_data.school_state),lowercase=False, bir
vectorizer.fit(project_data['school_state'].values)
print(vectorizer.get_feature_names())
school_state_one_hot = vectorizer.transform(project_data['school_state'].values)
print("Shape of matrix after one hot encodig ",school_state_one_hot.shape)
['AK', 'AL', 'AR', 'AZ', 'CA', 'CO', 'CT', 'DC', 'DE', 'FL', 'GA', 'HI', 'I
Å', 'ÍD', 'ÍL', 'ÍN', 'KS', 'KY', 'ĹA', 'MA', 'MD', 'ME', 'MI', 'MN', 'MO',
                 'ND', 'NE', 'NH', 'NJ', 'NM', 'NV', 'NY', 'OH', 'OK', 'O
'MS', 'MT', 'NC',
R', 'PA', 'RI', 'SC', 'SD', 'TN', 'TX', 'UT', 'VA', 'VT', 'WA', 'WI', 'WV',
'WY']
Shape of matrix after one hot encodig (109248, 51)
In [58]:
# We can do the similar feature encoding with teacher prefix
# removing special characters
import re
Clean_prefix = []
for prefix in (project_data['teacher_prefix'].values):
    prefix = re.sub('[^A-Za-z0-9]+', ' ', str(prefix))
    Clean prefix.append(prefix)
vectorizer = CountVectorizer(vocabulary=set(Clean_prefix),lowercase=False, binary=True)
vectorizer.fit(Clean_prefix)
print(vectorizer.get_feature_names())
teacher prefix one hot = vectorizer.transform(Clean prefix)
print("Shape of matrix after one hot encodig ",teacher_prefix_one_hot.shape)
['Dr ', 'Mr ', 'Mrs ', 'Teacher', 'nan']
```

Shape of matrix after one hot encodig (109248, 6)

In [59]:

```
# We can do the similar feature encoding with project_grade_category

vectorizer = CountVectorizer(vocabulary=set(project_data.project_grade_category),lowercase=
vectorizer.fit(project_data['project_grade_category'].values)
print(vectorizer.get_feature_names())

project_grade_category_one_hot = vectorizer.transform(project_data['project_grade_category'
print("Shape of matrix after one hot encodig ",project_grade_category_one_hot.shape)
```

```
['Grades 3-5', 'Grades 6-8', 'Grades 9-12', 'Grades PreK-2'] Shape of matrix after one hot encodig (109248, 4)
```

1.4.2 Vectorizing Text data

1.4.2.1 Bag of words

In [60]:

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

Shape of matrix after one hot encoding (109248, 16623)

1.4.2.2 Bag of Words on project_title

In [61]:

```
# you can vectorize the title also
# before you vectorize the title make sure you preprocess it

vectorizer = CountVectorizer(min_df=10)
text_bow_titles = vectorizer.fit_transform(preprocessed_titles)
print("Shape of matrix after one hot encoding ",text_bow_titles.shape)
```

Shape of matrix after one hot encoding (109248, 3329)

1.4.2.3 TFIDF vectorizer

In [62]:

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

Shape of matrix after one hot encodig (109248, 16623)

1.4.2.4 TFIDF Vectorizer on project_title

In [63]:

```
from sklearn.feature_extraction.text import TfidfVectorizer
vectorizer = TfidfVectorizer(min_df=10)
text_tfidf_titles = vectorizer.fit_transform(preprocessed_titles)
print("Shape of matrix after one hot encodig ",text_tfidf_titles.shape)
```

Shape of matrix after one hot encodig (109248, 3329)

1.4.2.5 Using Pretrained Models: Avg W2V

In [64]:

```
'''# Reading glove vectors in python: https://stackoverflow.com/a/38230349/4084039
def loadGloveModel(gloveFile):
    print ("Loading Glove Model")
   f = open(gloveFile,'r', encoding="utf8")
   model = \{\}
   for line in tqdm(f):
       splitLine = line.split()
       word = splitLine[0]
       embedding = np.array([float(val) for val in splitLine[1:]])
       model[word] = embedding
    print ("Done.",len(model)," words loaded!")
    return model
model = loadGloveModel('glove.42B.300d.txt')
Output:
Loading Glove Model
1917495it [06:32, 4879.69it/s]
Done. 1917495 words loaded!
words = []
for i in preproced texts:
   words.extend(i.split(' '))
for i in preproced titles:
   words.extend(i.split(' '))
print("all the words in the coupus", len(words))
words = set(words)
print("the unique words in the coupus", len(words))
inter words = set(model.keys()).intersection(words)
print("The number of words that are present in both glove vectors and our coupus", \
      len(inter_words),"(",np.round(len(inter_words)/len(words)*100,3),"%)")
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-pickl
import pickle
with open('glove_vectors', 'wb') as f:
   pickle.dump(words_courpus, f)
```

Out[64]:

```
'# Reading glove vectors in python: https://stackoverflow.com/a/38230349/408
4039\ndef (https://stackoverflow.com/a/38230349/4084039\ndef) loadGloveModel
(gloveFile):\n print ("Loading Glove Model")\n f = open(gloveFile,\'r
```

```
\', encoding="utf8")\n
                         model = {} \n
                                         for line in tqdm(f):\n
                                                                       spli
tLine = line.split()\n
                             word = splitLine[0]\n
                                                          embedding = np.ar
ray([float(val) for val in splitLine[1:]])\n
                                                   model[word] = embedding
      print ("Done.",len(model)," words loaded!")\n
                                                      return model\nmodel =
loadGloveModel(\'glove.42B.300d.txt\')\n\n# ================\nOu
          \nLoading Glove Model\n1917495it [06:32, 4879.69it/s]\nDone. 1917
495 words loaded!\n\n# ==============\n\nwords = []\nfor i in
preproced_texts:\n
                     words.extend(i.split(\' \'))\n\nfor i in preproced_tit
         words.extend(i.split(\' \'))\nprint("all the words in the coupus",
len(words))\nwords = set(words)\nprint("the unique words in the coupus", len
(words))\n\ninter_words = set(model.keys()).intersection(words)\nprint("The
number of words that are present in both glove vectors and our coupus",
len(inter_words),"(",np.round(len(inter_words)/len(words)*100,3),"%)")\n\nwo
rds_courpus = {}\nwords_glove = set(model.keys())\nfor i in words:\n
in words glove:\n
                        words_courpus[i] = model[i]\nprint("word 2 vec leng
th", len(words courpus))\n\n\ stronging variables into pickle files pytho
n: http://www.jessicayung.com/how-to-use-pickle-to-save-and-load-variables-i
n-python/\n\mimport (http://www.jessicayung.com/how-to-use-pickle-to-save-an
d-load-variables-in-python/\n\nimport) pickle\nwith open(\'glove_vectors\',
                   pickle.dump(words_courpus, f)\n'
\'wb\') as f:\n
```

In [65]:

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

In [66]:

```
# average Word2Vec
# compute average word2vec for each review.
avg_w2v_vectors = []; # the avg-w2v for each sentence/review is stored in this list
for sentence in tqdm(preprocessed_essays): # for each review/sentence
    vector = np.zeros(300) # as word vectors are of zero length
    cnt_words =0; # num of words with a valid vector in the sentence/review
    for word in sentence.split(): # for each word in a review/sentence
        if word in glove_words:
            vector += model[word]
            cnt_words += 1
    if cnt_words != 0:
        vector /= cnt_words
    avg_w2v_vectors.append(vector)

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

```
100%| 100%| 1009248/109248 [00:36<00:0 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/109248 | 109248/10924 | 109248/10924 | 109248/10924 | 109248/10924 | 109248/10924 | 109248/10924 | 109248/10924 | 109248/10924 | 109248/10924 | 109248/10924 | 109248/10924 | 109248/10924 | 109248/10924 | 109248/10924 | 109248/10924 | 109248/10924 | 109248/10924 | 109248/10924 | 109248/
```

1.4.2.6 Using Pretrained Models: AVG W2V on project_title

In [67]:

```
# Similarly we can vectorize for title also
avg_w2v_vectors = []; # the avg-w2v for each sentence/review is stored in this list
for sentence in tqdm(preprocessed_titles): # for each title
    vector = np.zeros(300) # as word vectors are of zero length
    cnt_words =0; # num of words with a valid vector in the sentence/review
    for word in sentence.split(): # for each word in a review/sentence
        if word in glove_words:
            vector += model[word]
            cnt_words += 1
    if cnt_words != 0:
        vector /= cnt_words
    avg_w2v_vectors.append(vector)

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

```
100%| 109248/109248 [00:02<00:0 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 109248 | 10924
```

1.4.2.7 Using Pretrained Models: TFIDF weighted W2V

In [68]:

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

In [69]:

```
# average Word2Vec
# compute average word2vec for each review.
tfidf_w2v_vectors = []; # the avg-w2v for each sentence/review is stored in this list
for sentence in tqdm(preprocessed_essays): # for each review/sentence
    vector = np.zeros(300) # as word vectors are of zero length
    tf_idf_weight =0; # num of words with a valid vector in the sentence/review
    for word in sentence.split(): # for each word in a review/sentence
        if (word in glove_words) and (word in tfidf_words):
            vec = model[word] # getting the vector for each word
            # here we are multiplying idf value(dictionary[word]) and the tf value((sentend
            tf_idf = dictionary[word]*(sentence.count(word)/len(sentence.split())) # gettir
            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]))
```

1.4.2.9 Using Pretrained Models: TFIDF weighted W2V on project title

In [70]:

```
# Similarly you can vectorize for title also
tfidf w2v vectors = []; # the avg-w2v for each sentence/review is stored in this list
for sentence in tqdm(preprocessed_titles): # for each title
    vector = np.zeros(300) # as word vectors are of zero Length
    tf idf weight =0; # num of words with a valid vector in the sentence/review
    for word in sentence.split(): # for each word in a review/sentence
        if (word in glove_words) and (word in tfidf_words):
            vec = model[word] # getting the vector for each word
            # here we are multiplying idf value(dictionary[word]) and the tf value((sentend
            tf_idf = dictionary[word]*(sentence.count(word)/len(sentence.split())) # gettir
            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%| 100%| 1009248/109248 [00:05<00:0 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 1009248 | 10092
```

1.4.3 Vectorizing Numerical features (Price)

```
In [80]:
```

```
# check this one: https://www.youtube.com/watch?v=0HOqOcln3Z4&t=530s
# standardization sklearn: https://scikit-learn.org/stable/modules/generated/sklearn.prepro
from sklearn.preprocessing import StandardScaler

# price_standardized = standardScalar.fit(project_data['price'].values)
# this will rise the error
# ValueError: Expected 2D array, got 1D array instead: array=[725.05 213.03 329. ... 399.
# 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 standar
print(f"Mean : {price_scalar.mean_[0]}, Standard deviation : {np.sqrt(price_scalar.var_[0])}
# Now standardize the data with above maen and variance.
price_standardized = price_scalar.transform(project_data['price'].values.reshape(-1, 1))
```

Mean: 298.1193425966608, Standard deviation: 367.49634838483496

In [81]:

```
price_standardized
```

```
Out[81]:
```

1.4.4 Vectorizing Numerical features (Quantity)

In [82]:

```
from sklearn.preprocessing import StandardScaler
import warnings
warnings.filterwarnings("ignore")

# price_standardized = standardScalar.fit(project_data['price'].values)
# this will rise the error
# ValueError: Expected 2D array, got 1D array instead: array=[725.05 213.03 329. ... 399.
# Reshape your data either using array.reshape(-1, 1)

quantity_scalar = StandardScaler()
quantity_scalar.fit(project_data['quantity'].values.reshape(-1,1)) # finding the mean and s
print(f"Mean : {quantity_scalar.mean_[0]}, Standard deviation : {np.sqrt(quantity_scalar.va

# Now standardize the data with above maen and variance.
quantity_standardized = quantity_scalar.transform(project_data['quantity'].values.reshape(-1, 1))
```

Mean: 16.965610354422964, Standard deviation: 26.182821919093175

```
In [83]:
```

1.4.5 Vectorizing Numerical features (teacher_number_of_previously_posted_projects)

In [84]:

```
from sklearn.preprocessing import StandardScaler
import warnings
warnings.filterwarnings("ignore")

# price_standardized = standardScalar.fit(project_data['price'].values)
# this will rise the error
# ValueError: Expected 2D array, got 1D array instead: array=[725.05 213.03 329. ... 399.
# Reshape your data either using array.reshape(-1, 1)

teacher_number_of_previously_posted_projects_scalar = StandardScaler()
teacher_number_of_previously_posted_projects_scalar.fit(project_data['teacher_number_of_pre
print(f"Mean : {teacher_number_of_previously_posted_projects_scalar.mean_[0]}, Standard dev

# Now standardize the data with above maen and variance.
teacher_number_of_previously_posted_projects_standardized = teacher_number_of_previously_posted_projects_standardized_projects_standardized_projects_standardized_projects_standardized_projects_standardized_projects_standardized_projects_standardized_projects_standardized_projects_standardized_projects_standardized_projects_standardized_projects_standardized_projects_standardized_projects_standardized_projects_standardized_projects_standardized_projects_standardized_projects_standardized_projects_standardized_projects_standardized_projects_standardized_projects_standardized_projects_standardized_pr
```

Mean: 11.153165275336848, Standard deviation: 27.77702641477403

In [85]:

```
teacher_number_of_previously_posted_projects_standardized
```

Out[85]:

1.4.6 Merging all the above features

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

```
In [86]:
```

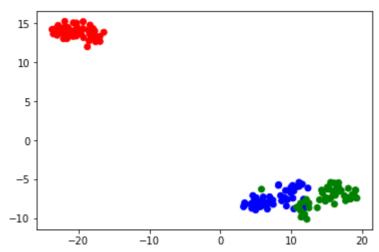
```
print(categories one hot.shape)
print(sub_categories_one_hot.shape)
print(text_bow.shape)
print(price_standardized.shape)
(109248, 9)
(109248, 30)
(109248, 16623)
(109248, 1)
In [87]:
# merge two sparse matrices: https://stackoverflow.com/a/19710648/4084039
from scipy.sparse import hstack
# with the same hstack function we are concatinating a sparse matrix and a dense matirx :)
X = hstack((categories_one_hot, sub_categories_one_hot, text_bow, price_standardized))
X.shape
Out[87]:
(109248, 16663)
```

Assignment 2: Apply TSNE

- 1. In the above cells we have plotted and analyzed many features. Please observe the plots and write the observations in markdown cells below every plot.
- 2. EDA: Please complete the analysis of the feature: teacher number of previously posted projects
- Build the data matrix using these features
 - school_state : categorical data (one hot encoding)
 - clean categories: categorical data (one hot encoding)
 - clean subcategories : categorical data (one hot encoding)
 - teacher prefix : categorical data (one hot encoding)
 - project grade category: categorical data (one hot encoding)
 - project_title: text data (BOW, TFIDF, AVG W2V, TFIDF W2V)
 - · price: numerical
 - teacher_number_of_previously_posted_projects : numerical
- 4. Now, plot FOUR t-SNE plots with each of these feature sets.
 - A. categorical, numerical features + project_title(BOW)
 - B. categorical, numerical features + project title(TFIDF)
 - C. categorical, numerical features + project_title(AVG W2V)
 - D. categorical, numerical features + project_title(TFIDF W2V)
- 5. Concatenate all the features and Apply TNSE on the final data matrix
- 6. Note 1: The TSNE accepts only dense matrices
- 7. Note 2: Consider only 5k to 6k data points to avoid memory issues. If you run into memory error issues, reduce the number of data points but clearly state the number of data-poins you are using

In [88]:

```
# this is the example code for TSNE
import numpy as np
from sklearn.manifold import TSNE
from sklearn import datasets
import pandas as pd
import matplotlib.pyplot as plt
iris = datasets.load_iris()
x = iris['data']
y = iris['target']
tsne = TSNE(n_components=2, perplexity=30, learning_rate=200)
X_embedding = tsne.fit_transform(x)
# if x is a sparse matrix you need to pass it as X_{embedding} = tsne.fit_transform(x.toarray)
for_tsne = np.hstack((X_embedding, y.reshape(-1,1)))
for_tsne_df = pd.DataFrame(data=for_tsne, columns=['Dimension_x', 'Dimension_y', 'Score'])
colors = {0:'red', 1:'blue', 2:'green'}
plt.scatter(for_tsne_df['Dimension_x'], for_tsne_df['Dimension_y'], c=for_tsne_df['Score'].
plt.show()
```



2.1 TSNE with BOW encoding of project_title feature (5000 Data Enteries)

```
In [89]:
```

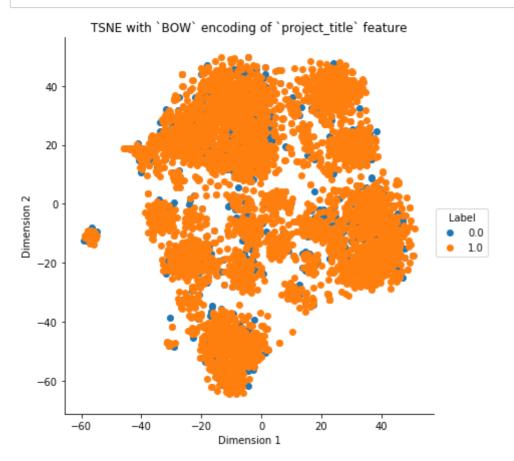
```
/_one_hot, price_standardized, quantity_standardized, teacher_number_of_previously_posted_pr

d
Out[89]:
```

(109248, 3432)

In [90]:

```
# please write all of the code with proper documentation and proper titles for each subsect
# when you plot any graph make sure you use
    # a. Title, that describes your plot, this will be very helpful to the reader
    # b. Legends if needed
    # c. X-axis label
    # d. Y-axis Label
from sklearn.manifold import TSNE
X = X.tocsr()
X_{\text{new}} = X[1000:6000,:]
random_5000 = X_new.toarray()
labels = project_data['project_is_approved']
labels_5000 = labels[1000:6000]
model = TSNE(n_components=2, random_state=0,perplexity=30,n_iter=1000,learning_rate=200)
tsne_data = model.fit_transform(random_5000)
# creating a new data frame which help us in ploting the result
tsne_data = np.vstack((tsne_data.T, labels_5000)).T
tsne_df = pd.DataFrame(data=tsne_data, columns=("Dimension 1", "Dimension 2", "Label"))
# Ploting the result of tsne
sns.FacetGrid(tsne_df, hue="Label", size=6).map(plt.scatter, 'Dimension 1', 'Dimension 2').a
plt.title('TSNE with `BOW` encoding of `project_title` feature')
plt.show()
```



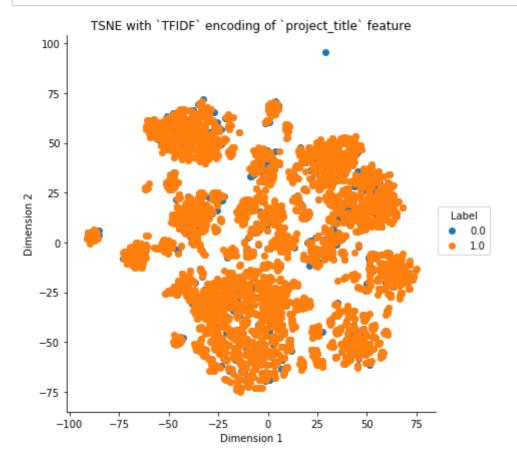
- 1. We can that their is No Clear separation between the "Approved Projects" & "Not Approved Projects" due overlapping of points.
- 2. Some points are scattered & some points group together to form Cluster.

2.2 TSNE with TFIDF encoding of project_title feature (5000 Data Enteries)

In [91]: stack((categories_one_hot, sub_categories_one_hot, school_state_one_hot, teacher_prefix_one pe Out[91]: (109248, 16726)

In [92]:

```
# please write all the code with proper documentation, and proper titles for each subsection
# when you plot any graph make sure you use
    # a. Title, that describes your plot, this will be very helpful to the reader
    # b. Legends if needed
    # c. X-axis label
    # d. Y-axis Label
X = X.tocsr()
X_{new} = X[1000:6000,:]
random 5000 = X new.toarray()
labels = project_data['project_is_approved']
labels_5000 = labels[1000:6000]
model = TSNE(n_components=2, random_state=0,perplexity=30,n_iter=1000,learning_rate=200)
tsne data = model.fit transform(random 5000)
# creating a new data frame which help us in ploting the result
tsne_data = np.vstack((tsne_data.T, labels_5000)).T
tsne_df = pd.DataFrame(data=tsne_data, columns=("Dimension 1", "Dimension 2", "Label"))
# Ploting the result of tsne
sns.FacetGrid(tsne_df, hue="Label", size=6).map(plt.scatter, 'Dimension 1', 'Dimension 2').a
plt.title('TSNE with `TFIDF` encoding of `project_title` feature')
plt.show()
```



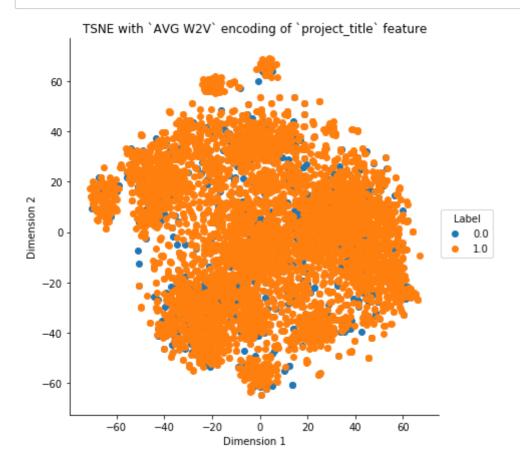
1. "Approved Projects" points are Grouped together very well but Overlapping "Not Approved Projects" points that's why we can't predict neither Approved Projects nor Rejected Projects.

2.3 TSNE with AVG W2V encoding of project_title feature (5000 Data Enteries)

In [93]: ((categories_one_hot, sub_categories_one_hot, school_state_one_hot, teacher_prefix_one_hot, Out[93]: (109248, 403)

In [94]:

```
# please write all the code with proper documentation, and proper titles for each subsection
# when you plot any graph make sure you use
    # a. Title, that describes your plot, this will be very helpful to the reader
    # b. Legends if needed
    # c. X-axis label
    # d. Y-axis Label
X = X.tocsr()
X_{new} = X[1000:6000,:]
random 5000 = X new.toarray()
labels = project_data['project_is_approved']
labels_5000 = labels[1000:6000]
model = TSNE(n_components=2, random_state=0,perplexity=30,n_iter=1000,learning_rate=200)
tsne data = model.fit transform(random 5000)
# creating a new data frame which help us in ploting the result
tsne_data = np.vstack((tsne_data.T, labels_5000)).T
tsne_df = pd.DataFrame(data=tsne_data, columns=("Dimension 1", "Dimension 2", "Label"))
# Ploting the result of tsne
sns.FacetGrid(tsne_df, hue="Label", size=6).map(plt.scatter, 'Dimension 1', 'Dimension 2').a
plt.title('TSNE with `AVG W2V` encoding of `project_title` feature')
plt.show()
```



- 1. For "Approved Projects" there are more no. of Points than "Not Approved Projects" but they are Overlapping with "Rejected Projects".
- 2. So, it is impossible to point out that which point belongs to which class.

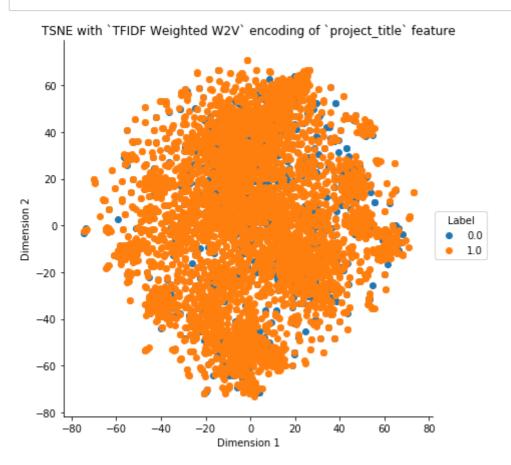
2.4 TSNE with TFIDF Weighted W2V encoding of project_title feature (5000 Data Enteries)

In [95]:

```
X = hstack((categories_one_hot, sub_categories_one_hot, school_state_one_hot, teacher_prefi
X.shape
Out[95]:
(109248, 403)
```

In [96]:

```
# please write all the code with proper documentation, and proper titles for each subsection
# when you plot any graph make sure you use
    # a. Title, that describes your plot, this will be very helpful to the reader
    # b. Legends if needed
    # c. X-axis label
    # d. Y-axis Label
X = X.tocsr()
X_{new} = X[1000:6000,:]
random 5000 = X new.toarray()
labels = project_data['project_is_approved']
labels_5000 = labels[1000:6000]
model = TSNE(n_components=2, random_state=0,perplexity=30,n_iter=1000,learning_rate=200)
tsne data = model.fit transform(random 5000)
# creating a new data frame which help us in ploting the result
tsne_data = np.vstack((tsne_data.T, labels_5000)).T
tsne_df = pd.DataFrame(data=tsne_data, columns=("Dimension 1", "Dimension 2", "Label"))
# Ploting the result of tsne
sns.FacetGrid(tsne_df, hue="Label", size=6).map(plt.scatter, 'Dimension 1', 'Dimension 2').a
plt.title('TSNE with `TFIDF Weighted W2V` encoding of `project_title` feature')
plt.show()
```



- 1. It is clearly seen that "Approved Projects" are very well seperated but overlapping with "Not Approved Points".
- 2. Therefore it is difficult to predict.

2.5 TSNE with BOW, TFIDF, AVG W2V, TFIDF Weighted W2V encoding of project_title feature (5000 Data Enteries)

In [97]:

ardized, quantity_standardized, teacher_number_of_previously_posted_projects_standardized,te

Out[97]:

(109248, 33649)

In [98]:

```
from sklearn.manifold import TSNE

X = X.tocsr()
X_new = X[1000:6000,:]
random_5000 = X_new.toarray()
labels = project_data['project_is_approved']
labels_5000 = labels[1000:6000]

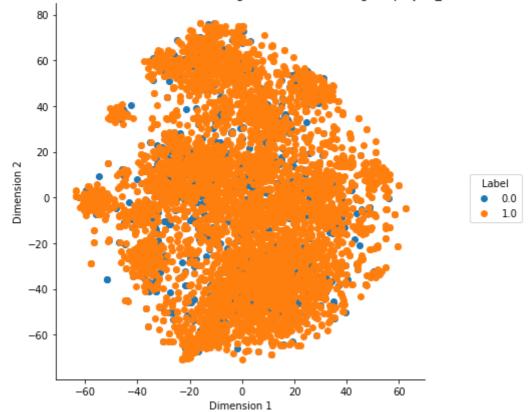
model = TSNE(n_components=2, random_state=0,perplexity=30,n_iter=1000,learning_rate=200)
tsne_data = model.fit_transform(random_5000)

# creating a new data frame which help us in ploting the result

tsne_data = np.vstack((tsne_data.T, labels_5000)).T
tsne_df = pd.DataFrame(data=tsne_data, columns=("Dimension 1", "Dimension 2", "Label"))

# Ploting the result of tsne
sns.FacetGrid(tsne_df, hue="Label", size=6).map(plt.scatter,'Dimension 1', 'Dimension 2').a
plt.title('TSNE with `BOW, TFIDF, AVG W2V, TFIDF Weighted W2V` encoding of `project_title`
plt.show()
```

TSNE with `BOW, TFIDF, AVG W2V, TFIDF Weighted W2V` encoding of `project_title` feature



Observation -

 For the Combine Tsne with Bag of Words, TF-IDF, Avg Word2Vec, TF-IDF Weighted Word2Vec also gives the Overlapping result of "Approved Projects" & "Rejected Projects", Hence we would have to try any other method.

2.6 Summary

- 1. We did not find any plot that separates "Approved Points" & "Not Approved Points" well from the above variants of T-SNE.
- 2. This is may be due to taking of fewer points beacuse of limited resources, T-SNE may Produce better result on taking Full dataset but not sure.

In []:			