

Assignment on Donor Choose

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

Attributes in the train.csv- 1) project_id 2) project_title 3) project_grade_category 4) project_subject_categories 5) school_state 6) project_subject_subcategories 7) project_resource_summary 8) project_essay_1 9) project_essay_2 10) project_essay_3 11) project_essay_4 12) project_submitted_datetime 13) teacher_id 14) teacher_prefix 15) teacher_number_of_previously_posted_project Attributes in resource.csv - 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: 1) id - A project_id value from the train.csv file. 2) description - Description of the resource. 3) quantity - Quantity of the resource required. 4) price - Price of the resource required. The data set contains the following label: project_is_approved - A binary flag indicating whether DonorsChoose approved the project. A value of 0 indicates the project was not approved, and a value of 1 indicates the project was approved.

In [1]: Here we importing bunch of libraries to perform Univariate analysis

```
%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

#https://plot.ly/python/getting-started-with-chart-studio/
import chart_studio.plotly as py
```

```
from collections import Counter
```

1.1 Reading the Data

```
In [2]: #we are loading two datas in pandas dataframe  
project_data = pd.read_csv(r'C:\Users\SAI\Downloads\Assignment_donorch  
ose 2018\train_data.csv')  
resource_data = pd.read_csv('resources.csv')
```

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

Number of data points in train data (109248, 17)

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

Out[3]:

	Unnamed: 0	id		teacher_id	teacher_prefix	school_state	project_s
0	160221	p253737	c90749f5d961ff158d4b4d1e7dc665fc		Mrs.	IN	2
1	140945	p258326	897464ce9ddc600bcfd1151f324dd63a		Mr.	FL	2

Unnamed: 0	id	teacher_id	teacher_prefix	school_state	project_s
2	21895 p182444 3465aa82da834c0582ebd0ef8040ca0		Ms.	AZ	2

here we have 109248 projects and 17 different attributes i.e.'teacher_id' 'teacher_prefix'
'school_state' 'project_submitted_datetime' 'project_grade_category' and others

In [4]: `project_data.dtypes`

Out[4]:

Unnamed: 0		int64
id		object
teacher_id		object
teacher_prefix		object
school_state		object
project_submitted_datetime		object
project_grade_category		object
project_subject_categories		object
project_subject_subcategories		object
project_title		object
project_essay_1		object
project_essay_2		object
project_essay_3		object
project_essay_4		object
project_resource_summary		object
teacher_number_of_previously_posted_projects		int64
project_is_approved		int64
dtype:	object	

In [5]: `project_data.info()`

<class 'pandas.core.frame.DataFrame'>

```
RangeIndex: 109248 entries, 0 to 109247
Data columns (total 17 columns):
Unnamed: 0                               109248 non-null int64
id                                         109248 non-null object
teacher_id                                 109248 non-null object
teacher_prefix                             109245 non-null object
school_state                               109248 non-null object
project_submitted_datetime                109248 non-null object
project_grade_category                   109248 non-null object
project_subject_categories               109248 non-null object
project_subject_subcategories            109248 non-null object
project_title                             109248 non-null object
project_essay_1                           109248 non-null object
project_essay_2                           109248 non-null object
project_essay_3                           3758 non-null object
project_essay_4                           3758 non-null object
project_resource_summary                 109248 non-null object
teacher_number_of_previously_posted_projects 109248 non-null int64
project_is_approved                      109248 non-null int64
dtypes: int64(3), object(14)
memory usage: 14.2+ MB
```

```
In [6]: print(list(project_data['project_is_approved'].unique()))
[0, 1]
```

1.2 Data Analysis

```
In [7]: y_value_counts = project_data['project_is_approved'].value_counts()
print("Number of projects thar are approved for funding ", y_value_counts[1], ", (", (y_value_counts[1]/(y_value_counts[1]+y_value_counts[0]))*100, "%)")
print("Number of projects thar are not approved for funding ", y_value_counts[0], ", (", (y_value_counts[0]/(y_value_counts[1]+y_value_counts[0]))*100, "%)")

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(arrowsyle="-"),
          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)

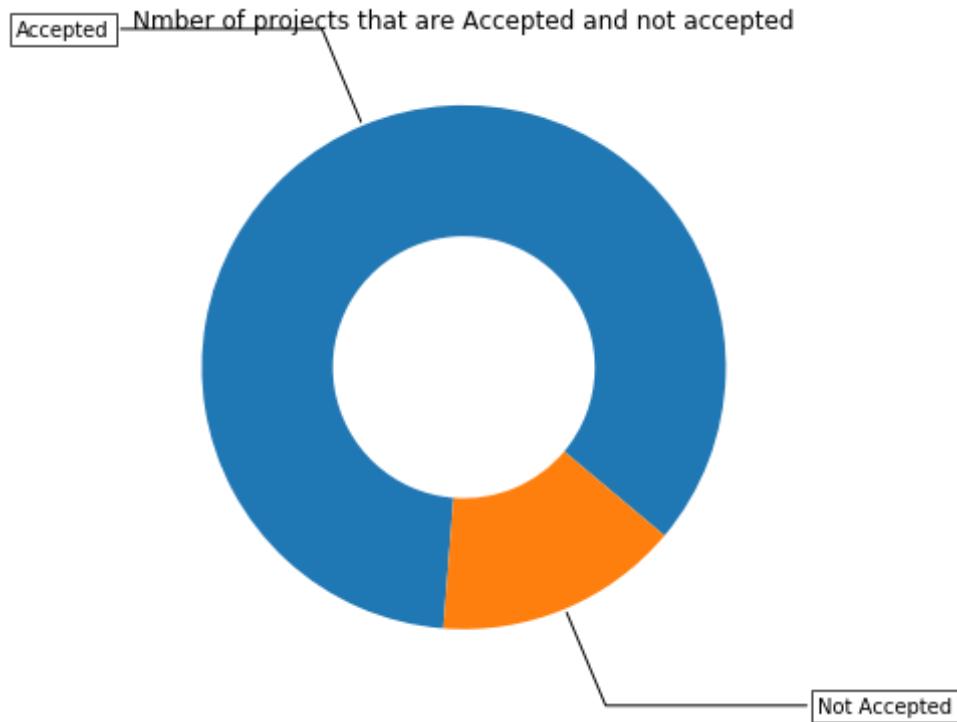
ax.set_title("Number of projects that are Accepted and not accepted")

plt.show()

```

Number of projects that are approved for funding 92706 , (84.85830404
217927 %)

Number of projects that are not approved for funding 16542 , (15.1416
95957820739 %)



from above results we can see that by visually and numerically that there are 85% of projects got funding and 15% of projects are nonfunding projects

In [9]:

```
#for each projects we need a multiple resources
print("Number of data points in train data", resource_data.shape)
print(resource_data.columns.values)
resource_data.head(2)
```

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

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

Out[9]:

	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

here we can see the there are 1541272 resources which more than no of projects also there are 4 columns of resources 'id' 'description' 'quantity' 'price'

In [10]: `resource_data.info()`

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1541272 entries, 0 to 1541271
Data columns (total 4 columns):
id            1541272 non-null object
description    1540980 non-null object
quantity      1541272 non-null int64
price          1541272 non-null float64
dtypes: float64(1), int64(1), object(2)
memory usage: 47.0+ MB
```

In [11]: `resource_data.describe()`

Out[11]:

	quantity	price
count	1.541272e+06	1.541272e+06
mean	2.860509e+00	5.028398e+01
std	7.570345e+00	1.447326e+02
min	1.000000e+00	0.000000e+00
25%	1.000000e+00	7.900000e+00
50%	1.000000e+00	1.499000e+01
75%	2.000000e+00	3.980000e+01

	quantity	price
max	8.000000e+02	9.999000e+03

1.2.1 Univariate Analysis: School State

```
In [12]: #for univariate analysis we are trying to get codes.  
#by searching different code and modifying it we putting here by giving  
citation to all search query  
  
# Pandas dataframe groupby count, mean: https://stackoverflow.com/a/19385591/4084039  
  
temp = pd.DataFrame(project_data.groupby("school_state")["project_is_aproved"].apply(np.mean)).reset_index()  
# 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')
'''
```

Out[12]:

```

# How to plot US state heatmap: https://datascience.stackexchange.com/a/9620\nnscl = [[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)\']]n\ndata = [
dict(\n    type=\'choropleth\',\n    colorscale = scl,\n    autocolorscale = False,\n    locations = temp[\'state_code\'],\n    z = temp[\'num_proposals\'].astype(float),\n    locationmode =
\'USA-states\',\n    text = temp[\'state_code\'],\n    marker = dict(line = dict (color = \'rgb(255,255,255)\',width = 2)),\n    colorbar = dict(title = "% of pro")\n) ]\n\nlayout = dict(\n    title = \\'Project Proposals % of Acceptance Rate by US States\',\n    geo = dict(\n        scope=\\'usa\',\n        projection=dict(
type=\\'albers usa\' ),\n        showlakes = True,\n        lakecolor = \\'rgb(255, 255, 255)\' ),\n    ),\n)\n\nfig = go.Figure(\n    data=data,\n    layout=layout)\noffline.iplot(fig, filename=\\'us-map-heat-map\\')\n'''
```

In [13]:

```

# https://www.csi.cuny.edu/sites/default/files/pdf/administration/ops/2
letterstabbrev.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        MT      0.816327
18        LA      0.831245
=====
States with highest % approvals
   state_code  num_proposals
30        NH      0.873563
35        OH      0.875152
47        WA      0.876178
28        ND      0.888112
8         DE      0.897959

```

Summary

1-from above information we observe that Delaware(DE)has the highest percentage of project acceptance having approximately 90% acceptance rate sucseeded by North Dakota(ND)and Washington(WA) nearly y 89% & 88% respectively

2-Also we could obsererved that Vermount is the lowest approval rated state by exactly 80% followed by District of coloumbia (DC) and Texas(TX) with nearly 80% and 81% respectively 3-Avg approval rate is 855 but here we getting respective state percentages uper Five showing states with lower approval rate and bottom five showing states with higher approval rate

In [14]:

```

# now we are going to run some fuctions to get stacked bar plots
#stacked bar plots matplotlib: https://matplotlib.org/gallery/lines\_bars\_and\_markers/bar\_stacked.html
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 [15]:

```

def univariate_barplots(data, col1, col2='project_is_approved', top=False):
    # Count number of zeros in dataframe python: https://stackoverflow.
    com/a/51540521/4084039
    temp = pd.DataFrame(project_data.groupby(col1)[col2].agg(lambda x:
x.eq(1).sum())).reset_index()

    # Pandas dataframe grouby count: https://stackoverflow.com/a/193855
    91/4084039
    temp['total'] = pd.DataFrame(project_data.groupby(col1)[col2].agg({
        'total':'count'})).reset_index()['total']
    temp['Avg'] = pd.DataFrame(project_data.groupby(col1)[col2].agg({
        'Avg':'mean'})).reset_index()['Avg']

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

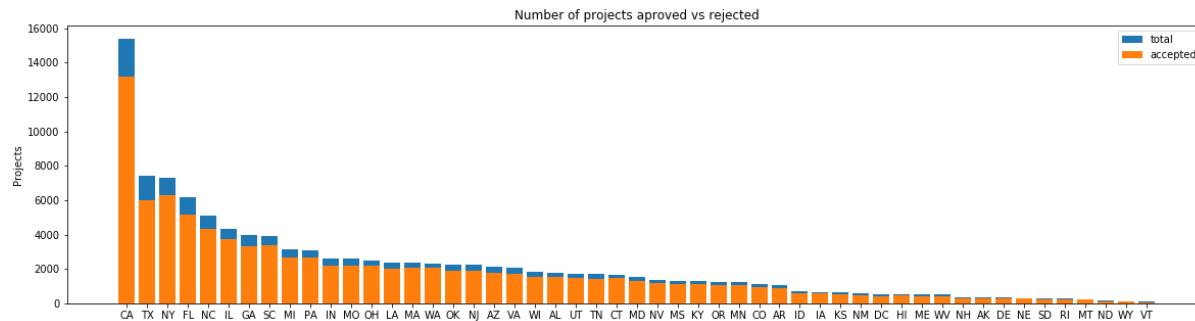
    if top:
        temp = temp[0:top]

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

```

In [16]:

```
univariate_barplots(project_data, 'school_state', 'project_is_approved',
, False)
```



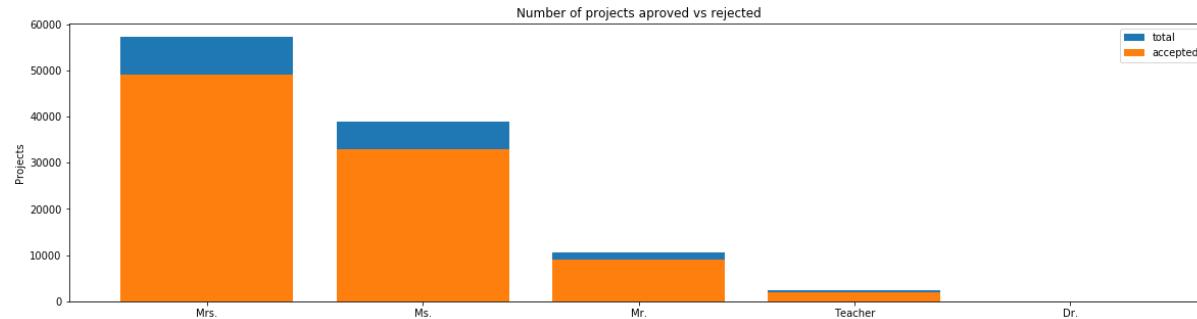
	school_state	project_is_approved	total	Avg
4	CA	13205	15388	0.858136
43	TX	6014	7396	0.813142
34	NY	6291	7318	0.859661
9	FL	5144	6185	0.831690
27	NC	4353	5091	0.855038
<hr/>				
39	RI	243	285	0.852632
26	MT	200	245	0.816327
28	ND	127	143	0.888112
50	WY	82	98	0.836735
46	VT	64	80	0.800000

SUMMARY :

1. Every state has greater than 80% success rate in approval.
2. There is a lot of variability in the number of projects that have been submitted across the States.
3. California(CA) has the highest number of project proposals when compared to the other states, Surprisingly, 85% of the projects gets approved on an average which is nearly 13205 out of 15388 project proposals.
4. Vermont(VT) has the lowest number of project proposals initiated (80) and almost 80% of the project proposal gets acceptance (64 out of 80). Well, in terms of rejection only 16 were rejected.

1.2.2 Univariate Analysis: teacher_prefix

```
In [17]: univariate_barplots(project_data, 'teacher_prefix', 'project_is_approved', top=False)
```



```
teacher_prefix  project_is_approved  total      Avg
2             Mrs.                  48997  57269  0.855559
3             Ms.                  32860  38955  0.843537
1             Mr.                  8960   10648  0.841473
4             Teacher              1877   2360   0.795339
0             Dr.                   9     13   0.692308
=====
```

```
teacher_prefix  project_is_approved  total      Avg
2             Mrs.                  48997  57269  0.855559
3             Ms.                  32860  38955  0.843537
1             Mr.                  8960   10648  0.841473
4             Teacher              1877   2360   0.795339
0             Dr.                   9     13   0.692308
```

SUMMARY :

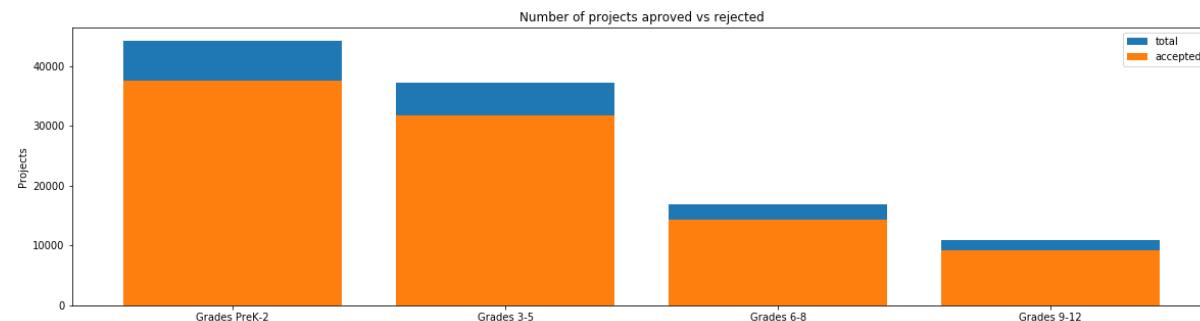
1. Female Teachers have the maximum number of projects proposed and accepted compared to the male teachers.
2. Teachers with prefixes Mrs. , which means Married Women as teachers have a higher number of projects Proposed as well as Accepted when compared to the younger

Unmarried Women Teachers.

3. Teachers with Dr. title have proposed hardly 13 projects and out of which 9 of them have been approved.

1.2.3 Univariate Analysis: project_grade_category

```
In [18]: univariate_barplots(project_data, 'project_grade_category', 'project_is_approved', top=False)
```



	project_grade_category	project_is_approved	total	Avg
3	Grades PreK-2	37536	44225	0.848751
0	Grades 3-5	31729	37137	0.854377
1	Grades 6-8	14258	16923	0.842522
2	Grades 9-12	9183	10963	0.837636

	project_grade_category	project_is_approved	total	Avg
3	Grades PreK-2	37536	44225	0.848751
0	Grades 3-5	31729	37137	0.854377
1	Grades 6-8	14258	16923	0.842522
2	Grades 9-12	9183	10963	0.837636

SUMMARY :

1. There are a lot of projects proposed for the students between Pre Kindergarten and 2nd Grade while for the rest it keeps decreasing.
2. The average Acceptance rate of the project is 84% irrespective of the Grade.
3. We also notice that Students between the 9th Grade and 12th Grade have the lowest number of projects proposed as well as accepted.

1.2.4 Univariate Analysis: project_subject_categories

In [19]:

```
#by following the below code we are doing simple Text Preprocessing
categories = list(project_data['project_subject_categories'].values)
# remove special characters from list of strings python: https://stackoverflow.com/a/47301924/4084039

# https://www.geeksforgeeks.org/removing-stop-words-nltk-python/
# https://stackoverflow.com/questions/23669024/how-to-strip-a-specific-word-from-a-string
# https://stackoverflow.com/questions/8270092/remove-all-whitespace-in-a-string-in-python
cat_list = []
for i in categories:
    temp = ""
    # consider we have text like this "Math & Science, Warmth, Care & Hunger"
    for j in i.split(','):# it will split it in three parts ["Math & Science", "Warmth", "Care & Hunger"]
        if 'The' in j.split():# this will split each of the category based on space "Math & Science"=> "Math", "&", "Science"
            j=j.replace('The','') # if we have the words "The" we are going to replace it with ''(i.e removing 'The')
            j = j.replace(' ','') # we are placing all the ' '(space) with ''(empty) ex:"Math & Science"=>"Math&Science"
            temp+=j.strip()+" "# abc ".strip() will return "abc", remove the trailing spaces
    temp = temp.replace('&','_') # we are replacing the & value int
```

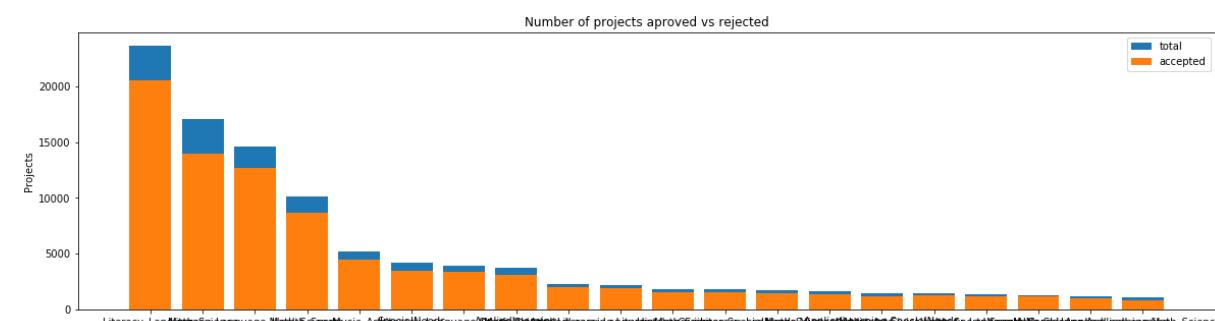
O cat_list.append(temp.strip())

```
In [20]: project_data['clean_categories'] = cat_list  
project_data.drop(['project_subject_categories'], axis=1, inplace=True)  
project_data.head(2)
```

Out[20]:

Unnamed: 0	id	teacher_id	teacher_prefix	school_state	project_s
0	160221 p253737	c90749f5d961ff158d4b4d1e7dc665fc	Mrs.	IN	2

```
In [21]: univariate_barplots(project_data, 'clean_categories', 'project_is_approved', top=20)
```



	clean_categories	project_is_approved	total	Avg
g 24 0	Literacy_Language	20520	23655	0.86747
32 9	Math_Science	13991	17072	0.81952

28	Literacy_Language	Math_Science	12725	14636	0.86943
2					
8		Health_Sports	8640	10177	0.84897
3					
40		Music_Arts	4429	5180	0.85501
9					
<hr/>					
Avg		clean_categories	project_is_approved	total	
19	History_Civics	Literacy_Language	1271	1421	0.894
441					
14	Health_Sports	SpecialNeeds	1215	1391	0.873
472					
50		Warmth_Care_Hunger	1212	1309	0.925
898					
33	Math_Science	AppliedLearning	1019	1220	0.835
246					
4	AppliedLearning	Math_Science	855	1052	0.812
738					

SUMMARY :

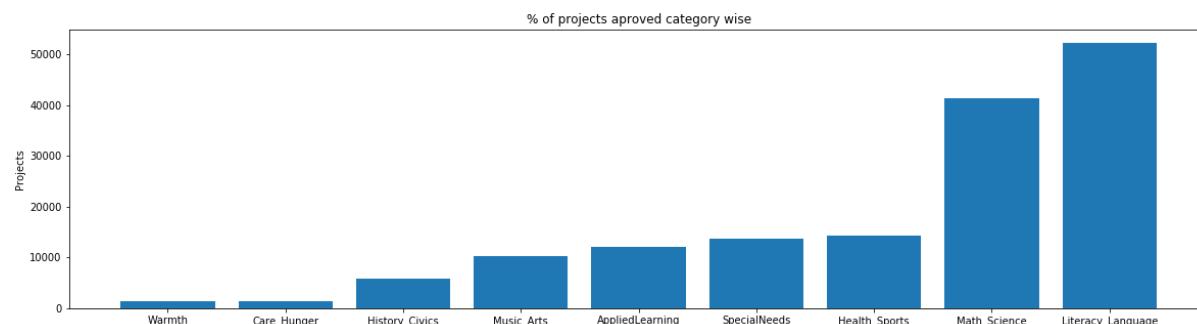
1. Projects belonging to the Literacy and Language categories have the highest number of projects proposed under. The maximum number of accepted projects also belong to this category, having an acceptance rate of nearly 87%.
2. Projects belonging to both Maths and Science have acceptance rate of nearly 82% while introducing the concept of Literacy and Language to this can increase its acceptance rate to nearly 87%
3. There is a lot of variability in the total number of projects proposed per Category of the project.
4. Projects belonging to both Maths and Science when combined with Applied Learning has the least number of projects proposed as well approved.
5. There is also Variability in Acceptance rate, projects under the category Warmth, Care and Hunger have an acceptance rate of 93.5%

```
In [22]: # 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 [23]: # 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 approved category wise')
plt.xticks(ind, list(sorted_cat_dict.keys()))
plt.show()
```



```
In [24]: for i, j in sorted_cat_dict.items():
    print("{:20} :{:10}".format(i,j))
```

Warmth	:	1388
Care_Hunger	:	1388
History_Civics	:	5914

```
Music_Arts      : 10293
AppliedLearning : 12135
SpecialNeeds    : 13642
Health_Sports   : 14223
Math_Science    : 41421
Literacy_Language: 52239
```

SUMMARY

1. The highest number of projects are registered under Literacy and Language with 52,239 projects, followed by Maths and Science having 41,421 projects.

1. There are only 1388 projects under the category of Warmth , Care or Hunger.

1.2.5 Univariate Analysis: project_subject_subcategories

```
In [25]: sub_catogories = list(project_data['project_subject_subcategories'].values)
# remove special characters from list of strings python: https://stackoverflow.com/a/47301924/4084039

# https://www.geeksforgeeks.org/removing-stop-words-nltk-python/
# https://stackoverflow.com/questions/23669024/how-to-strip-a-specific-word-from-a-string
# https://stackoverflow.com/questions/8270092/remove-all whitespace-in-a-string-in-python

sub_cat_list = []
for i in sub_catogories:
    temp = ""
    # consider we have text like this "Math & Science, Warmth, Care & Hunger"
    for j in i.split(','):
```

```

        science", "Warmth", "Care & Hunger"]
    if 'The' in j.split(): # this will split each of the category based on space "Math & Science"=> "Math","&","Science"
        j=j.replace('The','') # if we have the words "The" we are going to replace it with ''(i.e removing 'The')
        j = j.replace(' ','') # we are 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())

```

In [26]:

```

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

```

Out[26]:

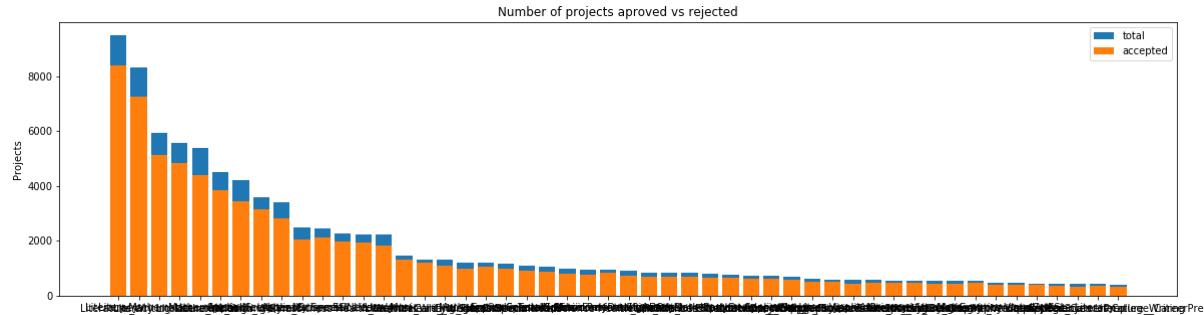
	Unnamed: 0	id	teacher_id	teacher_prefix	school_state	project_s
0	160221	p253737	c90749f5d961ff158d4b4d1e7dc665fc	Mrs.	IN	2
1	140945	p258326	897464ce9ddc600bcfd1151f324dd63a	Mr.	FL	2

In [27]:

```

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

```



	clean_subcategories	project_is_approved	total	A
vg 317	Literacy	8371	9486	0.8824
58				
319	Literacy Mathematics	7260	8325	0.8720
72				
331	Literature_Writing Mathematics	5140	5923	0.8678
03				
318	Literacy Literature_Writing	4823	5571	0.8657
33				
342	Mathematics	4385	5379	0.8152
07				

	clean_subcategories	project_is_approved	total	
Avg				
196	EnvironmentalScience Literacy	389	444	0.
876126				
127	ESL	349	421	0.
828979				
79	College_CareerPrep	343	421	0.
814727				
17	AppliedSciences Literature_Writing	361	420	0.
859524				
3	AppliedSciences College_CareerPrep	330	405	0.
814815				

SUMMARY :

1. The sub-Category Literacy has the highest number of projects approved with 8371 projects.
Also the acceptance rate is 88%.
2. The sub-Category Health and Wellness have the lowest number of projects proposed with 3,583 projects only.

```
In [28]: # count of all the words in corpus python: https://stackoverflow.com/a/22898595/4084039
```

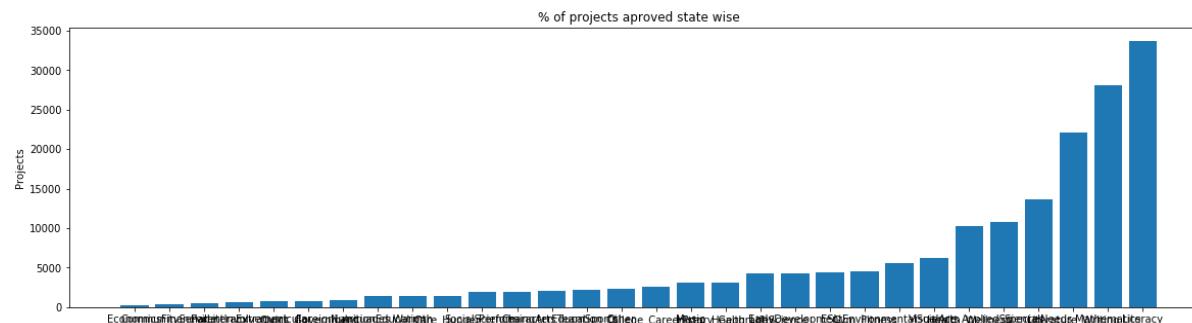
```
from collections import Counter
my_counter = Counter()
for word in project_data['clean_subcategories'].values:
    my_counter.update(word.split())
```

```
In [29]: # 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 approved state wise')
plt.xticks(ind, list(sorted_sub_cat_dict.keys()))
plt.show()
```



```
In [30]: for i, j in sorted_sub_cat_dict.items():
    print("{:20} :{:10}".format(i,j))
```

Economics	:	269
CommunityService	:	441
FinancialLiteracy	:	568
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	:	10234
AppliedSciences	:	10816
SpecialNeeds	:	13642
Literature_Writing	:	22179
Mathematics	:	28074
Literacy	:	33700

Summary:

- 1.The largest amount of projects are recorded with 33700 & 28074 projects in Literacy and Mathematics followed by 22179 projects in Literature_Writing.

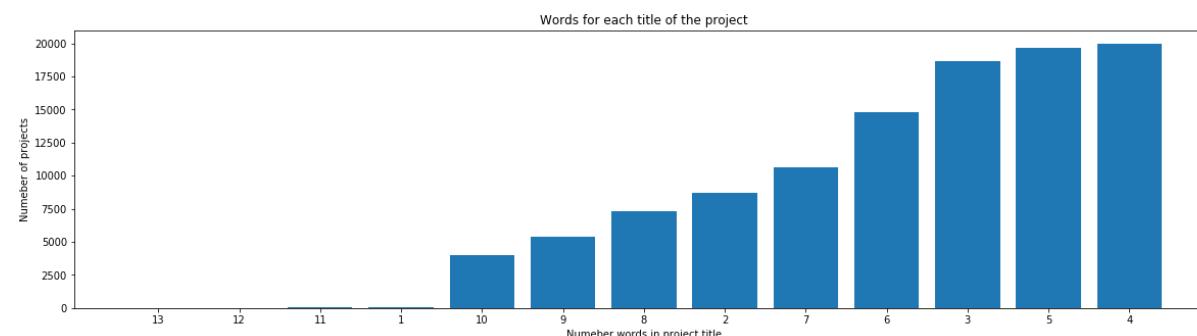
2.In the Economics(269)and CommunityService (441) recorded very low projects among all subjects.

1.2.6 Univariate Analysis: Text features (Title)

```
In [31]: #How to calculate number of words in a string in DataFrame: https://stackoverflow.com/a/37483537/4084039
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))
pl = plt.bar(ind, list(word_dict.values()))

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



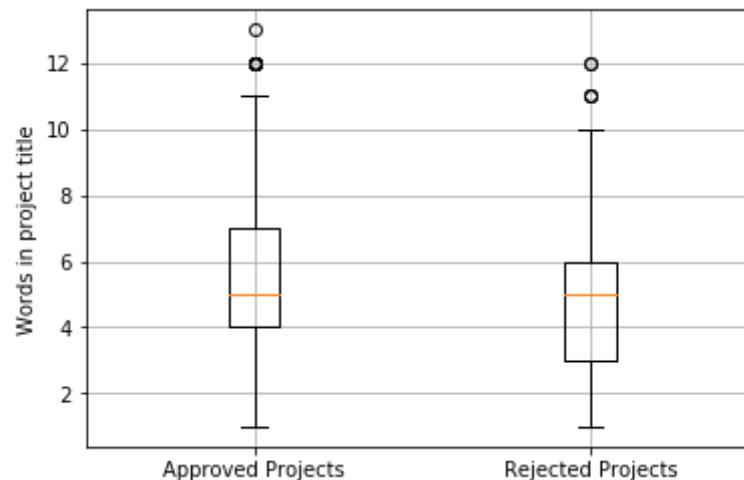
SUMMARY :

1. Most of the projects have 4 words in the title.
2. Roughly most of the projects have 3, 4 or 5 words in the title.
3. There are hardly any project titles containing more than 10 words and single word

```
In [32]: approved_title_word_count = project_data[project_data['project_is_approved']==1]['project_title'].str.split().apply(len)
approved_title_word_count = approved_title_word_count.values

rejected_title_word_count = project_data[project_data['project_is_approved']==0]['project_title'].str.split().apply(len)
rejected_title_word_count = rejected_title_word_count.values
```

```
In [33]: # 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()
```

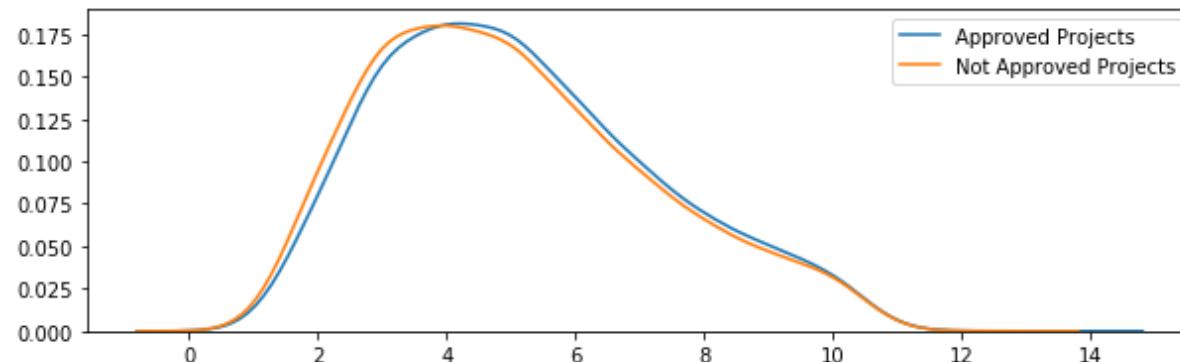


Summary:

1-From above box plot we can see that boxplot for approved projects is slightly higher than rejected projects although median of both plots coincides slightly.

2- The no of words in approved projects are slightly more than rejected projects

```
In [34]: 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()
```



Summary:

1-This is the PDF of no of words in Title.Blue line indicates Project approved curve
and orange line shows projects not approved curve.

2-From PDF graph we can observe that blue line is slightly ahead orange line which

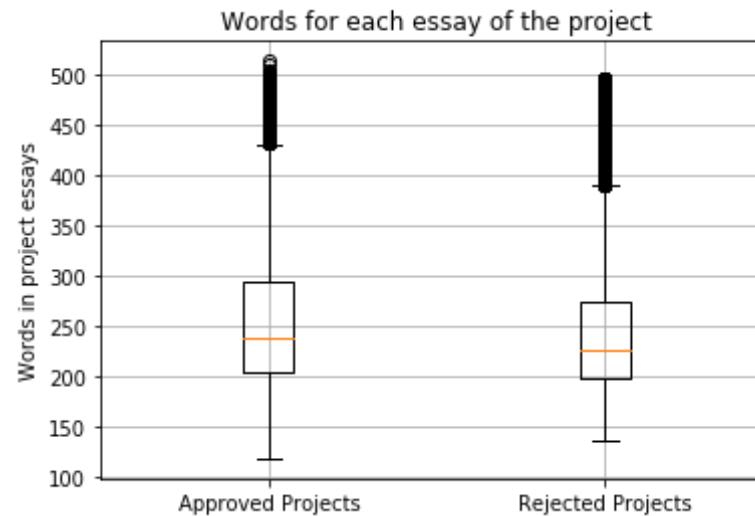
```
means no of words are more in approved projects than not approved
```

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

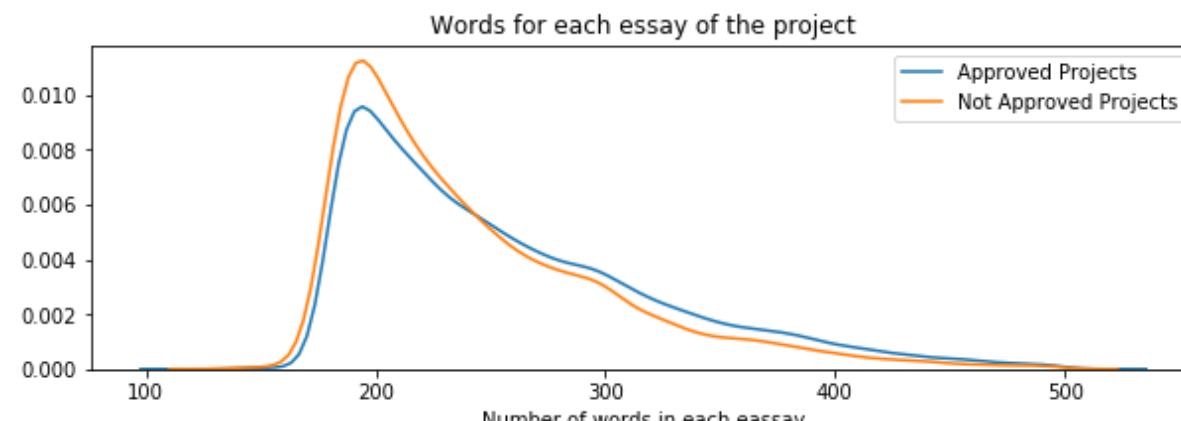
```
In [35]: # merge two column text dataframe:  
project_data["essay"] = project_data["project_essay_1"].map(str) +\\  
    project_data["project_essay_2"].map(str) + \\  
    project_data["project_essay_3"].map(str) + \\  
    project_data["project_essay_4"].map(str)
```

```
In [36]: approved_word_count = project_data[project_data['project_is_approved']=  
=1]['essay'].str.split().apply(len)  
approved_word_count = approved_word_count.values  
  
rejected_word_count = project_data[project_data['project_is_approved']=  
=0]['essay'].str.split().apply(len)  
rejected_word_count = rejected_word_count.values
```

```
In [38]: #we choose the following code to find the no of words in essays  
# 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()
```



```
In [39]: 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 essay')
plt.legend()
plt.show()
```



SUMMARY :

From both the plot it is clear that the number of Projects approved have a slightly more number of words in the Title when compared to the Rejected Projects. The Boxplots use the Percentiles while the above graph used Probability densities.

1.2.8 Univariate Analysis: Cost per project

```
In [40]: # we get the cost of the project using resource.csv file
resource_data.head(3)
```

Out[40]:

	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

```
In [41]: # https://stackoverflow.com/questions/22407798/how-to-reset-a-dataframe-s-indexes-for-all-groups-in-one-step
price_data = resource_data.groupby('id').agg({'price':'sum', 'quantity':'sum'}).reset_index()
price_data.head(2)
```

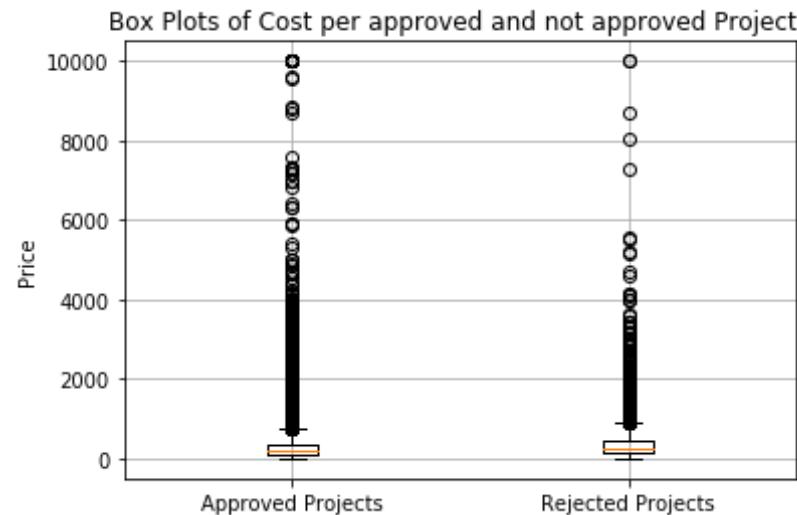
Out[41]:

	id	price	quantity
0	p000001	459.56	7
1	p000002	515.89	21

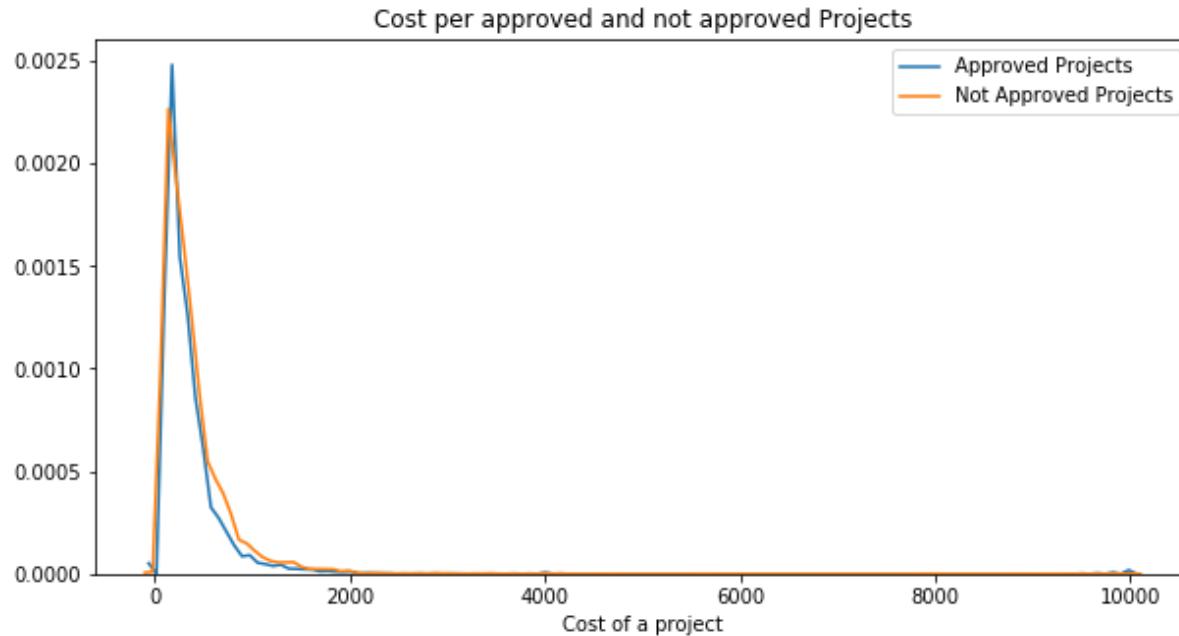
```
In [42]: # join two dataframes in python:
project_data = pd.merge(project_data, price_data, on='id', how='left')
```

```
In [43]: 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 [44]: # 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()
```



```
In [45]: plt.figure(figsize=(10,5))  
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()
```



Summary:

Not much can be understood from the box plot depicting the Cost involved per project. We can generalise from the PDF curves that mostly Projects that are very costly are usually not approved.

```
In [48]: # http://zetcode.com/python/prettytable/  
from prettytable import PrettyTable  
  
#If you get a ModuleNotFoundError error , install prettytable using: pi  
p3 install prettytable  
  
x = PrettyTable()  
x.field_names = ["Percentile", "Approved Projects", "Not Approved Proj  
cts"]
```

```

for i in range(0,101,5):
    x.add_row([i,np.round(np.percentile(approved_price,i), 3), np.round
(np.percentile(rejected_price,i), 3)])
print(x)

```

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

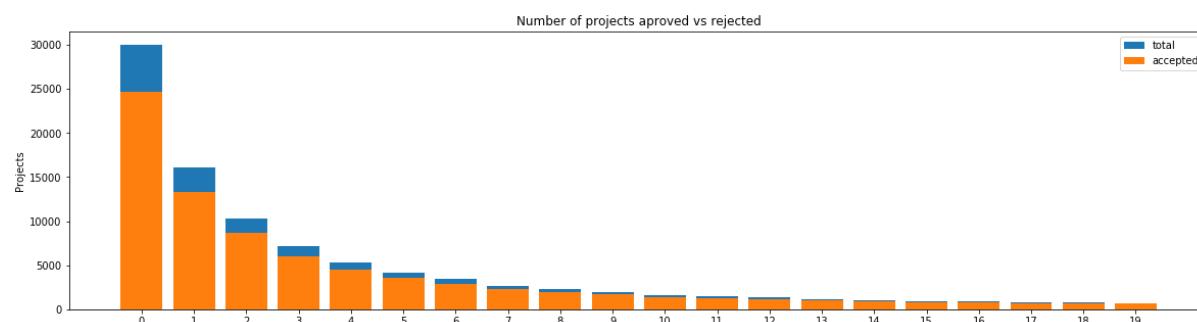
SUMMARY :

1. The approved projects tend to have lower cost when compared to the projects that have not been approved. This can be noticed by looking at the percentile values. The 50th percentile Cost value for an approved project is 199 dollars while for the cost for the not approved projects is 263 dollars.
2. The Maximum price for any project should be less than 10,000 dollars.

3. Typically, any approved Project costs less than the that of the Projects not approved across the spectrum of Percentiles.

1.2.9 Univariate Analysis: `teacher_number_of_previously_posted_projects`

```
In [49]: univariate_barplots(project_data, 'teacher_number_of_previously_posted_projects',  
                           'project_is_approved', top=20)
```



```
teacher_number_of_previously_posted_projects project_is_approved t  
total \\  
0 0014 0 24652 3  
1 6058 1 13329 1  
2 0350 2 8705 1  
3 7110 3 5997  
4 5266 4 4452
```

```
Avg  
0 0.821350  
1 0.830054  
2 0.841063
```

```

3  0.843460
4  0.845423
=====
              teacher_number_of_previously_posted_projects  project_is_approved
total  \
15                  15                      818
942
16                  16                      769
894
17                  17                      712
803
18                  18                      666
772
19                  19                      632
710

          Avg
15  0.868365
16  0.860179
17  0.886675
18  0.862694
19  0.890141

```

SUMMARY :

1. There is a lot of variability in the number of projects previously proposed by the teacher varying from 0 to more than 20.
2. We observe that it is not mandatory for a teacher to have proposed any project prior. Maximum number of teachers, nearly 82% of the approved projects have been submitted by teachers with no prior project proposals. New talent and efforts are well appreciated.
3. Very few teachers who have proposed more than 20 projects have got approval. But the rate of approval is Higher given the teacher has proposed atleast 19 different projects.

1.2.10 Univariate Analysis: project_resource_summary

```
In [50]: ## Let us separate the data and carry out our work only on the required  
Project Resource Summaries
```

```
summaries = []  
for a in project_data["project_resource_summary"] :summaries.append(a)  
summaries[0:10]
```

```
Out[50]: ['My students need opportunities to practice beginning reading skills i  
n English at home.',  
 'My students need a projector to help with viewing educational program  
s',  
 'My students need shin guards, athletic socks, Soccer Balls, goalie g  
loves, and training materials for the upcoming Soccer season.',  
 'My students need to engage in Reading and Math in a way that will ins  
pire them with these Mini iPads!',  
 'My students need hands on practice in mathematics. Having fun and per  
sonalized journals and charts will help them be more involved in our da  
ily Math routines.',  
 'My students need movement to be successful. Being that I have a varie  
ty of students that have all different types of needs, flexible seating  
would assist not only these students with special needs, but all studen  
ts.',  
 'My students need some dependable laptops for daily classroom use for  
reading and math.',  
 'My students need ipads to help them access a world of online resource  
s that will spark their interest in learning.',  
 "My students need three devices and three management licenses for smal  
l group's easy access to newly-implemented online programs--Go Noodle P  
lus, for increased in-class physical activity and Light Sail, an intera  
ctive reading program.",  
 'My students need great books to use during Independent Reading, Read  
Alouds, Partner Reading and Author Studies.]
```

```
In [ ]: The length of the obtained list of Project summaries should match the t  
otal number of project s  
ummaries in the project data
```

```
In [51]: len(summaries)
```

```
Out[51]: 109248
```

```
In [52]: ##### Identifying the numbers from the project summaries and storing the values as a key value pair in a dictionary to avoid the position of the value within huge value of summary data
```

```
numeric_summary_values = []
for x in tqdm(range(len(summaries))):
    for s in summaries[x].split():
        if s.isdigit():
            numeric_summary_values[x] = int()
```

```
100%|██████████| 109248/109248 [00:01<00:00, 7369 0.33it/s]
```

```
In [53]: numeric_summary_values[14]
```

```
Out[53]: 0
```

```
In [54]: # now we have only key value pairs for summaries containing numeric value, so
```

```
numeric_digits = {}
for c in range(len(summaries)):
    if c in numeric_summary_values.keys():
        numeric_digits[c]= numeric_summary_values[c]
    else:
        numeric_digits[c]=0
```

```
In [55]: len(numeric_digits)
```

```
Out[55]: 109248
```

```
In [56]: ## Converting the key value pairs to 1 or 0 based on presence of Numeric Values.
```

```
digit_in_summary = []
```

```
for a in numeric_digits.values():
    if a > 0 :
        digit_in_summary.append(1)
    else :
        digit_in_summary.append(0)
```

```
In [57]: digit_in_summary[0:20]
```

```
Out[57]: [0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0]
```

```
In [58]: project_data['digit_in_summary']= digit_in_summary
```

```
In [59]: project_data.head(20)
```

```
Out[59]:
```

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

1	140945	p258326	897464ce9ddc600bcfd1151f324dd63a	Mr.	FL	
---	--------	---------	----------------------------------	-----	----	--

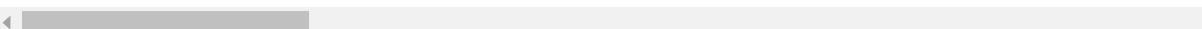
2	21895	p182444	3465aaf82da834c0582ebd0ef8040ca0	Ms.	AZ	
---	-------	---------	----------------------------------	-----	----	--

Unnamed: 0		id		teacher_id	teacher_prefix	school_state	project
3	45	p246581	f3cb9bffbba169bef1a77b243e620b60		Mrs.	KY	
4	172407	p104768	be1f7507a41f8479dc06f047086a39ec		Mrs.	TX	
5	141660	p154343	a50a390e8327a95b77b9e495b58b9a6e		Mrs.	FL	
6	21147	p099819	9b40170bfa65e399981717ee8731efc3		Mrs.	CT	
7	94142	p092424	5bfd3d12fae3d2fe88684bbac570c9d2		Ms.	GA	
8	112489	p045029	487448f5226005d08d36bdd75f095b31		Mrs.	SC	

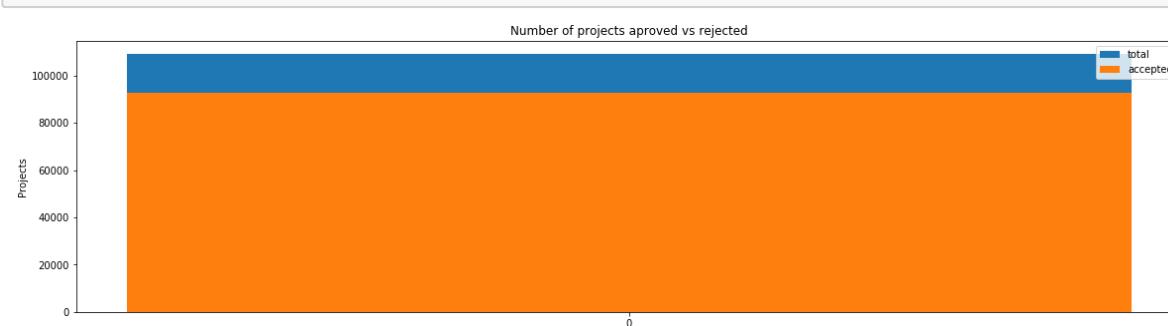
Unnamed: 0		id		teacher_id	teacher_prefix	school_state	project
9	158561	p001713	140eeac1885c820ad5592a409a3a8994		Ms.	NC	
10	43184	p040307	363788b51d40d978fe276bcb1f8a2b35		Mrs.	CA	
11	127083	p251806	4ba7c721133ef651ca54a03551746708		Ms.	CA	
12	19090	p051126	5e52c92b7e3c472aad247a239d345543		Mrs.	NY	
13	15126	p003874	178f6ae765cd4e0fb143a77c47fd65e2		Mrs.	OK	
14	62232	p233127	424819801de22a60bba7d0f4354d0258		Ms.	MA	
15	67303	p132832	bb6d6d054824fa01576ab38dfa2be160		Ms.	TX	

	Unnamed: 0	id		teacher_id	teacher_prefix	school_state	project
16	127215	p174627	4ad7e280fddff889e1355cc9f29c3b89		Mrs.	FL	
17	157771	p152491	e39abda057354c979c5b075cffbe5f88		Ms.	NV	
18	122186	p196421	fcd9b003fc1891383f340a89da02a1a6		Mrs.	GA	
19	146331	p058343	8e07a98deb1bc74c75b97521e05b1691		Ms.	OH	

20 rows × 21 columns



```
In [60]: univariate_barplots(project_data,'digit_in_summary','project_is_approved',top=2)
```



```
      digit_in_summary  project_is_approved  total      Avg
0                      0                  92706  109248  0.848583
=====
      digit_in_summary  project_is_approved  total      Avg
0                      0                  92706  109248  0.848583
```

1.3 Text preprocessing

1.3.1 Essay Text

In [61]: `project_data.head(2)`

Out[61]:

	Unnamed: 0	id	teacher_id	teacher_prefix	school_state	project_s
0	160221	p253737	c90749f5d961ff158d4b4d1e7dc665fc	Mrs.	IN	2
1	140945	p258326	897464ce9ddc600bcfd1151f324dd63a	Mr.	FL	2

2 rows × 21 columns

◀ ▶

In [62]: `# printing some random essays.`
`print(project_data['essay'].values[0])`
`print("="*50)`

```
print(project_data['essay'].values[150])
print("=="*50)
print(project_data['essay'].values[1000])
print("=="*50)
print(project_data['essay'].values[20000])
print("=="*50)
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 have 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 and experiences to us that open our eyes to 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. Many times our parents are learning to read and speak English along side of their children. Sometimes this creates barriers for parents to be able to help their child learn phonetics, letter recognition, and other reading skills.\r\n\r\nBy providing these dvd's and players, students 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 proficiency status, will be offered to be a part of this program. These educational 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 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 dvd player to use for the year. The plan is to use these videos and educational dvd's for the years to come for other EL students.\r\nnnannan

The 51 fifth grade students that will cycle through my classroom this year 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 students, 97.3% are minority students. \r\nThe school has a vibrant community that loves to get together and celebrate. Around Halloween there is a whole school parade to show off the beautiful costumes that students wear. 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 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 regular, stationary, 4-legged 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 times. 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 school.

When 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 Stools 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 there are not enough of them.

We ask a lot of students to sit for 7 hours a day. The Hokki stools will be a compromise 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 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 schools for a child who can't sit still.

How do you remember your days of school? Was it in a sterile environment 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 each day.

My class is made up of 28 wonderfully unique boys and girls of mixed races in Arkansas.

They attend a Title I school, which means 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, 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 cr

eate the mood in our classroom setting to be one of a themed nautical environment. Creating a classroom environment is very important in the success 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, 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 cards to their team groups.\r\n\r\nYour generous donations will help me to help make our classroom a fun, inviting, learning environment from day 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 to make our new school year a very successful one. Thank you!nannan

=====

My kindergarten students have varied disabilities ranging from speech and language delays, cognitive delays, gross/fine motor delays, to autism. 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. They want to be able to move as they learn or so they say. Wobble chairs are the answer and I love them because they develop their core, which enhances gross motor and 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

=====

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 makeup is 97.6% African-American, making up the largest 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 children and we focus not only on academics but one smart, effective, efficient, and disciplined students with good character. In our classroom we can utilize 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 [63]: # https://stackoverflow.com/a/47091490/4084039
import re

def decontracted(phrase):
    # specific
    phrase = re.sub(r"won't", "will not", phrase)
    phrase = re.sub(r"can't", "can not", phrase)

    # general
    phrase = re.sub(r"\n't", " not", phrase)
    phrase = re.sub(r"\re", " are", phrase)
    phrase = re.sub(r"\s", " is", phrase)
    phrase = re.sub(r"\d", " would", phrase)
    phrase = re.sub(r"\ll", " will", phrase)
    phrase = re.sub(r"\t", " not", phrase)
    phrase = re.sub(r"\ve", " have", phrase)
    phrase = re.sub(r"\m", " am", phrase)
    return phrase
```

```
In [64]: sent = decontracted(project_data['essay'].values[20000])
print(sent)
print("*"*50)
```

My kindergarten students have varied disabilities ranging from speech and language delays, cognitive delays, gross/fine motor delays, to autism. 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 them because they develop their core, which enhances gross motor and 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 [65]: # \r \n \t remove from string python: http://texthandler.com/info/remove-line-breaks-python/
sent = sent.replace('\r', ' ')
sent = sent.replace('\t', ' ')
sent = sent.replace('\n', ' ')
print(sent)
```

My kindergarten students have varied disabilities ranging from speech and language delays, cognitive delays, gross/fine motor delays, to autism. They are eager beavers and always strive to work their hardest working past 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 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 them because they develop their core, which enhances gross motor and fine motor skills. 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 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 [66]: #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 language delays cognitive delays gross fine motor delays to autism They are eager beavers and always strive to work their hardest working past 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 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 They want to be able to move as they learn or so they say Wobble chairs are the answer and I love them 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 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 [67]: # https://gist.github.com/sebleier/554280
# we are removing the words from the stop words list: 'no', 'nor', 'not'
stopwords =['i', 'me', 'my', 'myself', 'we', 'our', 'ours', 'ourselves',
           'you', "you're", "you've", \
           "you'll", "you'd", 'your', 'yours', 'yourself', 'yourselves',
           'he', 'him', 'his', 'himself', \
           'she', "she's", 'her', 'hers', 'herself', 'it', "it's", 'its',
           'itself', 'they', 'them', 'their', \
           'theirs', 'themselves', 'what', 'which', 'who', 'whom', 'this',
           'that', "that'll", 'these', 'those', \
           'am', 'is', 'are', 'was', 'were', 'be', 'been', 'being', 'h
```

```
ave', 'has', 'had', 'having', 'do', 'does', \
        'did', 'doing', 'a', 'an', 'the', 'and', 'but', 'if', 'or', 
    'because', 'as', 'until', 'while', 'of', \
        'at', 'by', 'for', 'with', 'about', 'against', 'between', 
    'into', 'through', 'during', 'before', '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 [68]: # Combining all the above statements
from tqdm import tqdm
preprocessed_essays = []
# tqdm is for printing the status bar
for sentence in tqdm(project_data['essay'].values):
    sent = decontracted(sentence)
    sent = sent.replace('\\r', ' ')
    sent = sent.replace('\\n', ' ')
    sent = sent.replace('\\t', ' ')
    sent = re.sub('[^A-Za-z0-9]+', ' ', sent)
    # https://gist.github.com/sebleier/554280
    sent = ' '.join(e for e in sent.split() if e not in stopwords)
    preprocessed_essays.append(sent.lower().strip())
```

100%|██████████| 109248/109248 [01:52<00:00, 97
0.03it/s]

```
In [69]: # after preprocessing
```

```
preprocessed_essays[15000]
```

Out[69]: 'you bring horse water not make drink however expose students world technology not able stop exploring how cool would children exposed simple technology ipad classroom my students reside small town hearts shape world they limited resources prepare real world it statistically proven students not prepared real world they might able text play game however skills needed capable surviving work force today involve technology integration among multiple curricula science technology reading multimedia it integral part students education tools capable striving advanced learning environment assist providing appropriate instruction effectively my students opportunity exposure endless resources simply not available classroom computer it said app everything i mind blown amount educational apps available educators students apps allow student developmental disability effectively communicate classroom teacher device need charged 24 7 app serve additional means instruction literacy centers 100 full engagement app allow students interactively enact founding country directional journey europe america list goes the purchase ipad also assist development classroom support newest initiative county school district titled byod bring your own device nannan'

1.3.2 Project title Text

In [70]: # printing some random essays.
print(project_data['project_title'].values[0])
print("=*50)
print(project_data['project_title'].values[150])
print("=*50)
print(project_data['project_title'].values[1000])
print("=*50)
print(project_data['project_title'].values[20000])
print("=*50)
print(project_data['project_title'].values[99999])
print("=*50)

Educational Support for English Learners at Home
=====

More Movement with Hokki Stools

```
=====
Sailing Into a Super 4th Grade Year
=====
We Need To Move It While We Input It!
=====
Inspiring Minds by Enhancing the Educational Experience
=====
```

```
In [71]: preprocessed_titles = []
for titles in tqdm(project_data["project_title"]):
    title = decontracted(titles)
    title = title.replace('\r', ' ')
    title = title.replace('\n', ' ')
    title = title.replace('\n', ' ')
    title = re.sub('[^A-Za-z0-9]+', ' ', title)
    title = ' '.join(f for f in title.split() if f not in stopwords)
    preprocessed_titles.append(title.lower().strip())
```

```
100%|██████████| 109248/109248 [00:06<00:00, 1805
2.00it/s]
```

```
In [72]: print(preprocessed_titles[0])
print("=*50)
print(preprocessed_titles[50])
print("=*50)
print(preprocessed_titles[500])
print("=*50)
print(preprocessed_titles[5000])
print("=*50)
print(preprocessed_titles[10000])
print("=*50)
```

```
educational support english learners home
=====
be active be energized
=====
classroom chromebooks college bound seniors
=====
bouncing our wiggles worries away
```

```
=====
family book clubs
=====
```

1. 4 Preparing data for models

```
In [73]: project_data.columns
```

```
Out[73]: Index(['Unnamed: 0', 'id', 'teacher_id', 'teacher_prefix', 'school_state',
       'project_submitted_datetime', 'project_grade_category', 'project_title',
       'project_essay_1', 'project_essay_2', 'project_essay_3',
       'project_essay_4', 'project_resource_summary',
       'teacher_number_of_previously_posted_projects', 'project_is_approved',
       'clean_categories', 'clean_subcategories', 'essay', 'price', 'quantity',
       'digit_in_summary'],
      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
- quantity : numerical
- teacher_number_of_previously_posted_projects : numerical
- price : numerical

1.4.1 Vectorizing Categorical data

```
In [ ]: #we will follow this link for vectorizing  
#https://www.appliedaicourse.com/course/applied-ai-course-online/lesson  
s/handling-categorical-and-numerical-features/  
  
# we use count vectorizer to convert the values into one hot encoded fe  
atures  
from sklearn.feature_extraction.text import CountVectorizer  
vectorizer = CountVectorizer(vocabulary=list(sorted_cat_dict.keys()), l  
owercase=False, binary=True)  
vectorizer.fit(project_data['clean_categories'].values)  
print(vectorizer.get_feature_names())  
  
categories_one_hot = vectorizer.transform(project_data['clean_categorie  
s'].values)  
print("Shape of matrix after one hot encoding ",categories_one_hot.shape)
```

```
In [75]: # we use count vectorizer to convert the values into one hot encoded fe  
atures  
vectorizer = CountVectorizer(vocabulary=list(sorted_sub_cat_dict.keys  
()), lowercase=False, binary=True)  
vectorizer.fit(project_data['clean_subcategories'].values)  
print(vectorizer.get_feature_names())  
  
sub_categories_one_hot = vectorizer.transform(project_data['clean_subca  
tegories'].values)  
print("Shape of matrix after one hot encoding ",sub_categories_one_hot.s  
hape)
```

```
['Economics', 'CommunityService', 'FinancialLiteracy', 'ParentInvolveme  
nt', 'Extracurricular', 'Civics_Government', 'ForeignLanguages', 'Nutri  
tionEducation', 'Warmth', 'Care_Hunger', 'SocialSciences', 'PerformingA  
rts', 'CharacterEducation', 'TeamSports', 'Other', 'College_CareerPre  
p', 'Music', 'History_Geography', 'Health_LifeScience', 'EarlyDevelopme
```

```
nt', 'ESL', 'Gym_Fitness', 'EnvironmentalScience', 'VisualArts', 'Healt  
h_Wellness', 'AppliedSciences', 'SpecialNeeds', 'Literature_Writing',  
'Mathematics', 'Literacy']  
Shape of matrix after one hot encoding (109248, 30)
```

school states

```
In [76]: my_counter = Counter()  
for state in project_data['school_state'].values:  
    my_counter.update(state.split())
```

```
In [77]: school_state_cat_dict = dict(my_counter)  
sorted_school_state_cat_dict = dict(sorted(school_state_cat_dict.items()  
(), key=lambda kv: kv[1]))
```

```
In [78]: ## we use count vectorizer to convert the values into one hot encoded f  
eatures  
vectorizer = CountVectorizer(vocabulary=list(sorted_school_state_cat_di  
ct.keys()), lowercase=False  
, binary=True)  
vectorizer.fit(project_data['school_state'].values)  
print(vectorizer.get_feature_names())  
school_state_categories_one_hot = vectorizer.transform(project_data['sc  
hool_state'].values)  
print("Shape of matrix after one hot encoding ",school_state_categories  
_one_hot.shape)
```

```
['VT', 'WY', 'ND', 'MT', 'RI', 'SD', 'NE', 'DE', 'AK', 'NH', 'WV', 'M  
E', 'HI', 'DC', 'NM', 'KS', 'IA', 'ID', 'AR', 'CO', 'MN', 'OR', 'KY',  
'MS', 'NV', 'MD', 'CT', 'TN', 'UT', 'AL', 'WI', 'VA', 'AZ', 'NJ', 'OK',  
'WA', 'MA', 'LA', 'OH', 'MO', 'IN', 'PA', 'MI', 'SC', 'GA', 'IL', 'NC',  
'FL', 'NY', 'TX', 'CA']  
Shape of matrix after one hot encoding (109248, 51)
```

One hot encode Teacher Prefix

```
In [79]: my_counter = Counter()
for teacher_prefix in project_data['teacher_prefix'].values:
    teacher_prefix = str(teacher_prefix)
    my_counter.update(teacher_prefix.split())

In [80]: teacher_prefix_cat_dict = dict(my_counter)
sorted_teacher_prefix_cat_dict = dict(sorted(teacher_prefix_cat_dict.items(), key=lambda kv: kv[1]))
)
```

we use count vectorizer to convert the values into one hot encoded features unlike

the previous catagories this catagory returns a value error :np.nan invalid document ref-
<https://stackoverflow.com/questions/39303912/tfidfvectorizer-in-scikit-learn-valueerror-np-nan-is-an-invalid-document/39308809#39308809>

```
In [81]: vectorizer = CountVectorizer(vocabulary=list(sorted_teacher_prefix_cat_dict.keys()), lowercase=False, binary=True)

vectorizer.fit(project_data['teacher_prefix'].values.astype("U"))
print(vectorizer.get_feature_names())
teacher_prefix_categories_one_hot = vectorizer.transform(project_data['teacher_prefix'].values.astype("U"))
print("Shape of matrix after one hot encoding ",teacher_prefix_categories_one_hot.shape)

['nan', 'Dr.', 'Teacher', 'Mr.', 'Ms.', 'Mrs.']
Shape of matrix after one hot encoding  (109248, 6)
```

One Hot Encode - Project Grade Category

```
In [82]: my_counter = Counter()
for project_grade in project_data['project_grade_category'].values:
```

```
my_counter.update(project_grade.split())
```

```
In [83]: project_grade_cat_dict = dict(my_counter)
sorted_project_grade_cat_dict = dict(sorted(project_grade_cat_dict.items(), key=lambda kv: kv[1]))
```

```
In [84]: ## we use count vectorizer to convert the values into one hot encoded features
vectorizer = CountVectorizer(vocabulary=list(sorted_project_grade_cat_dict.keys()), lowercase=False, binary=True)
vectorizer.fit(project_data['project_grade_category'].values)
print(vectorizer.get_feature_names())
project_grade_categories_one_hot = vectorizer.transform(project_data['project_grade_category'].values)
print("Shape of matrix after one hot encoding ",project_grade_categories_one_hot.shape)

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

1.4.2 Vectorizing Text data

1.4.2 Vectorizing Text data

```
In [85]: # We are considering only the words which appeared in at least 10 documents(rows or projects).
vectorizer = CountVectorizer(min_df=10)
text_bow = vectorizer.fit_transform(preprocessed_essays)
print("Shape of matrix after one hot encoding ",text_bow.shape)

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

```
In [86]: print("There are{} unique words among the {} number of Project essays,considering atleast 10 different projects has the same word".format(text_bow.shape[1],text_bow.shape[0]))
```

There are 16623 unique words among the 109248 number of Project essays, considering atleast 10 different projects has the same word

1.4.2.2 Bag of Words on project_title

```
In [87]: vectorizer = CountVectorizer(min_df=10)
title_bow = vectorizer.fit_transform(preprocessed_titles)
print("Shape of matrix after one hot encoding ", title_bow.shape)
```

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

```
In [88]: print("There are {} unique words among the {} number of Project titles, considering atleast 10 different projects has the same word".format(text_bow.shape[1],text_bow.shape[0]))
```

There are 16623 unique words among the 109248 number of Project titles, considering atleast 10 different projects has the same word

1.4.2.3 TFIDF vectorizer

```
In [89]: 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 encoding ", text_tfidf.shape)
```

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

1.4.2.4 TFIDF Vectorizer on project_title

```
In [90]: from sklearn.feature_extraction.text import TfidfVectorizer
vectorizer = TfidfVectorizer(min_df=10)
title_tfidf = vectorizer.fit_transform(preprocessed_titles)
print("Shape of matrix after one hot encoding ", title_tfidf.shape)
```

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

1.4.2.5 Using Pretrained Models: Avg W2V

In [91]:

```
'''  
# Reading glove vectors in python: https://stackoverflow.com/a/3823034  
# 9/4084039  
def loadGloveModel(gloveFile):  
    print ("Loading Glove Model")  
    f = open(gloveFile, 'r', encoding="utf8")  
    model = {}  
    for line in tqdm(f):  
        splitLine = line.split()  
        word = splitLine[0]  
        embedding = np.array([float(val) for val in splitLine[1:]])  
        model[word] = embedding  
    print ("Done.",len(model)," words loaded!")  
    return model  
model = loadGloveModel('glove.42B.300d.txt')  
  
# ======  
Output:  
  
Loading Glove Model  
1917495it [06:32, 4879.69it/s]  
Done. 1917495 words loaded!  
  
# ======  
  
words = []  
for i in preproc_texts:  
    words.extend(i.split(' '))  
  
for i in preproc_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-pickle-to-save-and-load-variables-in-python/

import pickle
with open('glove_vectors', 'wb') as f:
    pickle.dump(words_courpus, f)

...

```

Out[91]:

```

'\n# Reading glove vectors in python: 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        splitLine = line.split()\n        word = splitLine[0]\n        embedding = np.array([float(val) for val in splitLine[1:]])\n        model[word] = embedding\n    print ("Done.",len(model)," words loaded!")\n    return model\nmodel = loadGloveModel(\n    'glove.42B.300d.txt')\n\n# =====\n\nOutput:\n\nLoading Glove Model\n1917495it [06:32, 4879.69it/s]\nDone. 1917495 words loaded!\n# =====\n\nwords = []\nfor i in preproc_texts:\n    words.extend(i.split(' '))\nfor i in preproc_titles:\n    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))\nninter_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), "%)")

```

```
ds)/len(words)*100,3), "%")\n\nwords_courpus = {}\nwords_glove = set(mo  
del.keys())\nfor i in words:\n    if i in words_glove:\n        words_c  
ourpus[i] = model[i]\nprint("word 2 vec length", len(words_courpus))\n  
\n# stronging variables into pickle files python: http://www.jessicayung.  
com/how-to-use-pickle-to-save-and-load-variables-in-python/\nimport pickle\nwith open('glove_vectors', 'wb') as f:\n    pickle.dump  
(words_courpus, f)\n\n
```

```
In [93]: # 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 [95]: # average Word2Vec  
# compute average word2vec for each review.  
avg_w2v_vectors_essays = [] # the avg-w2v for each sentence/review is  
# stored in this list  
for sentence in tqdm(preprocessed_essays): # for each review/sentence  
    vector = np.zeros(300) # as word vectors are of zero length  
    cnt_words = 0; # num of words with a valid vector in the sentence/re  
view  
    for word in sentence.split(): # for each word in a review/sentence  
        if word in glove_words:  
            vector += model[word]  
            cnt_words += 1  
    if cnt_words != 0:  
        vector /= cnt_words  
    avg_w2v_vectors_essays.append(vector)  
  
print(len(avg_w2v_vectors_essays))  
print(len(avg_w2v_vectors_essays[0]))
```

100%|██████████| 109248/109248 [01:06<00:00, 164
0.52it/s]

109248
300

1.4.2.6 Using Pretrained Models: AVG W2V on project_title

```
In [96]: avg_w2v_vectors_title = [] # the avg-w2v for each sentence/review is stored in this list
for sentence in tqdm(preprocessed_titles): # 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_title.append(vector)

print(len(avg_w2v_vectors_title))
print(len(avg_w2v_vectors_title[0]))
```

100%|██████████| 109248/109248 [00:07<00:00, 1467 4.09it/s]

109248
300

1.4.2.7 Using Pretrained Models: TFIDF weighted W2V

```
In [97]: # 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 [98]: # average Word2Vec
# compute average word2vec for each review.
tfidf_w2v_vectors_essay = [] # the avg-w2v for each sentence/review is
# stored in this list
for sentence in tqdm(preprocessed_essays): # for each review/sentence
    vector = np.zeros(300) # as word vectors are of zero length
    tf_idf_weight = 0; # num of words with a valid vector in the sentence/review
    for word in sentence.split(): # for each word in a review/sentence
        if (word in glove_words) and (word in tfidf_words):
            vec = model[word] # getting the vector for each word
            # here we are multiplying idf value(dictionary[word]) and the tf value((sentence.count(word)/len(sentence.split())))
            tf_idf = dictionary[word]*(sentence.count(word)/len(sentence.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_essay.append(vector)

print(len(tfidf_w2v_vectors_essay))
print(len(tfidf_w2v_vectors_essay[0]))
```

```
100%|██████████| 109248/109248 [09:26<00:00, 20
8.24it/s]
```

```
109248
300
```

1.4.2.9 Using Pretrained Models: TFIDF weighted W2V on `project_title`

```
In [99]: # Similarly we can vectorize for title also
tfidf_model = TfidfVectorizer()
tfidf_model.fit(preprocessed_titles)
# 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 [100]: # average Word2Vec
# compute average word2vec for each review.
tfidf_w2v_vectors_title = [] # the avg-w2v for each sentence/review is
    stored in this list
for sentence in tqdm(preprocessed_titles): # for each review/sentence
    vector = np.zeros(300) # as word vectors are of zero length
    tf_idf_weight = 0; # num of words with a valid vector in the sentence/review
    for word in sentence.split(): # for each word in a review/sentence
        if (word in glove_words) and (word in tfidf_words):
            vec = model[word] # getting the vector for each word
            # here we are multiplying idf value(dictionary[word]) and the tf value((sentence.count(word)/len(sentence.split())))
            tf_idf = dictionary[word]*(sentence.count(word)/len(sentence.split()))
            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_title.append(vector)

print(len(tfidf_w2v_vectors_title))
print(len(tfidf_w2v_vectors_title[0]))
```

```
100%|██████████| 109248/109248 [00:10<00:00, 1076
1.39it/s]
```

```
109248
300
```

1.4.3 Vectorizing Numerical features

```
In [101]: # after observing this: 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].
# 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 mean and variance.
price_standardized = price_scalar.transform(project_data[['price']].value
s.reshape(-1, 1))
```

Mean : 298.1193425966608, Standard deviation : 367.49634838483496

In [102]: price_standardized

Out[102]: array([[-0.3905327],
[0.00239637],
[0.59519138],
...,
[-0.15825829],
[-0.61243967],
[-0.51216657]])

Summary :

1. We observe that on an average Each project costs nearly 298 Dollars. With a Standard Deviation of 368 dollars. So , mostly majority of the projects are less than 1000 Dollars.

Vectorizing - Quantity (Numerical Data)

```
In [103]: import warnings
warnings.filterwarnings("ignore")
from sklearn.preprocessing import StandardScaler
quantity_scalar = StandardScaler()
quantity_scalar.fit(project_data['quantity'].values.reshape(-1,1)) # finding the mean and standard deviation of this data
print(f"Mean : {quantity_scalar.mean_[0]}, Standard deviation : {np.sqrt(quantity_scalar.var_[0])}")

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

Mean : 16.965610354422964, Standard deviation : 26.182821919093175

```
In [104]: quantity_standardized
```

```
Out[104]: array([[ 0.23047132],
 [-0.60977424],
 [ 0.19227834],
 ...,
 [-0.4951953 ],
 [-0.03687954],
 [-0.45700232]])
```

Summary :

1. The projects on an average require atleast 17 Different of similar items. We observe that the Price paid is generally for the purchase of these Items. Donors can choose on projects to donate based on the Items provided to aid the Students of any Grade.

Vectorizing - Number of Projects Proposed Previously by the Teacher

(Numerical Data)

```
In [105]: prev_projects_scalar = StandardScaler()
## Finding the mean and standard deviation of this data
prev_projects_scalar.fit(project_data['teacher_number_of_previously_posted_projects'].values.reshape(-1,1))
print("Mean : {}".format(prev_projects_scalar.mean_[0]))
print("Standard deviation : {}".format(np.sqrt(prev_projects_scalar.var_[0])))
# Now standardize the data with above mean and variance.
prev_projects_standardized = prev_projects_scalar.transform(project_data['teacher_number_of_previously_posted_projects'].values.reshape(-1, 1))
```

```
Mean : 11.153165275336848
Standard deviation : 27.77702641477403
```

```
In [106]: prev_projects_standardized
```

```
Out[106]: array([[-0.40152481],
                 [-0.14951799],
                 [-0.36552384],
                 ...,
                 [-0.29352189],
                 [-0.40152481],
                 [-0.40152481]])
```

Summary:

We observe that Teachers usually on a median propose atleast 11 completely different projects.

The Teachers area unit so actively seeking facilitate to help for the betterment of the students in their neighborhood.

1.4.4 Merging all the above features

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

```
In [107]: 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 [108]: # merge two sparse matrices: https://stackoverflow.com/a/19710648/4084039
from scipy.sparse import hstack
# with the same hstack function we are concatenating a sparse matrix and a dense matrix :
X = hstack((categories_one_hot, sub_categories_one_hot, text_bow, price_standardized))
X.shape
```

```
Out[108]: (109248, 16663)
```

now we are going to use one of the most powerful technique for dimensionality reduction i.e. t-SNE which is very intuitively geometrically elegant solution for dimensionality reduction for visualisation of data basically tSNE is Neighbourhood Preserving Embedding

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

```
In [109]: # https://github.com/pavlin-policar/fastTSNE
# http://colah.github.io/posts/2014-10-Visualizing-MNIST
```

```

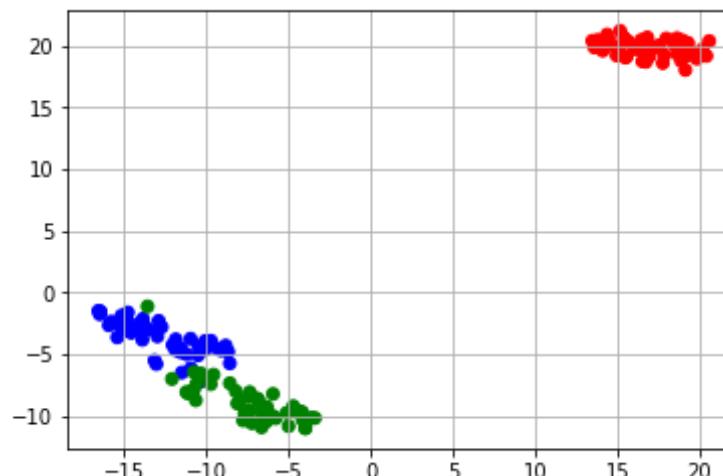
# this is the example code for TSNE
import numpy as np
from sklearn.manifold import TSNE
from sklearn import datasets
import pandas as pd
import matplotlib.pyplot as plt

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()) , .toarray() will convert the sparse matrix int
o dense matrix

for_tsne = np.hstack((X_embedding, y.reshape(-1,1)))
for_tsne_df = pd.DataFrame(data=for_tsne, columns=['Dimension_x','Dimen
sion_y','Score'])
colors = {0:'red', 1:'blue', 2:'green'}
plt.scatter(for_tsne_df['Dimension_x'], for_tsne_df['Dimension_y'], c=f
or_tsne_df['Score'].apply(lambda x: colors[x]))
plt.grid()
plt.show()

```



```
In [110]: print("The Shape of Data matrices for Categorical Data are :")
print("\n")
print("The Shape of Data Matrix for different Categories of projects is
    :{}".format(categories_one_hot.shape))
print("The Shape of Data Matrix for different Sub-categories of project
s is : {}".format(sub_categories_one_hot.shape))
print("The Shape of Data Matrix with respect to Projects from a particu
lar State in the United States is : {}".format(school_state_categories_
one_hot.shape))
print("The Shape of the Data Matrix of the different projects with resp
ect to the Grades of the students is : {} ".format(project_grade_catego
ries_one_hot.shape))
print("The Shape of the Data Matrix with respect to title of the Teache
r proposing the Teacher is: {}".format(teacher_prefix_categories_one_ho
t.shape))
print("\n")
print("*"*100)
print("\n")
print("The Shape of Data matrices for Numerical Data are :")
print("\n")
print("The Shape of the Data Matrix for price of the projects is : {}".f
ormat(price_standardized.shape))
print("The Shape of the Data Matrix for Quantity of the items for the p
rojects is : {}".format(quantity_standardized.shape))
print("The Shape of the Data Matrix for the Number of Projects Proposed
    Previously by the Teacher is : {}".format(prev_projects_standardized.s
hape))
print("\n")
print("*"*100)
print("\n")
print("TITLE BOW : {}".format(title_bow.shape))
print("\n")
print("TITLE TFIDF : {}".format(title_tfidf.shape))
print("\n")
print("TITLE AVG W2V : ({}, {})".format(len(avg_w2v_vectors_title), len(
avg_w2v_vectors_title[0])))
print("\n")
```

```
print("TITLE TFIDF W2V : ({}, {})".format(len(tfidf_w2v_vectors_title),  
len(tfidf_w2v_vectors_title[0])))
```

The Shape of Data matrices for Categorical Data are :

The Shape of Data Matrix for different Categories of projects is :(1092
48, 9)

The Shape of Data Matrix for different Sub-categories of projects is :
(109248, 30)

The Shape of Data Matrix with respect to Projects from a particular Sta
te in the United States is : (109248, 51)

The Shape of the Data Matrix of the different projects with respect to
the Grades of the students is : (109248, 5)

The Shape of the Data Matrix with respect to title of the Teacher propo
sing the Teacher is: (109248, 6)

=====

The Shape of Data matrices for Numerical Data are :

The Shape of the Data Matrix for price of the projects is : (109248, 1)

The Shape of the Data Matrix for Quantity of the items for the projects
is : (109248, 1)

The Shape of the Data Matrix for the Number of Projects Proposed Previo
usly by the Teacher is : (109248, 1)

=====

TITLE BOW : (109248, 3329)

TITLE TFIDF : (109248, 3329)

```
TITLE AVG W2V : (109248, 300)
```

```
TITLE TFIDF W2V : (109248, 300)
```

```
In [111]: X = hstack((categories_one_hot, sub_categories_one_hot, school_state_categories_one_hot,
project_grade_categories_one_hot, teacher_prefix_categories_one_hot, price_standardized,
quantity_standardized, prev_projects_standardized, title_bow))
X.shape
```

```
Out[111]: (109248, 3433)
```

```
In [112]: from sklearn.manifold import TSNE
X = X.tocsr()
X_new = X[0:5000,:]
```

```
In [113]: X_new = X_new.toarray()
model = TSNE(n_components = 2, perplexity = 100.0, random_state = 0)
tsne_data_b = model.fit_transform(X_new)
```

```
In [114]: labels = project_data["project_is_approved"]
labels_new = labels[0: 5000]
len(labels_new)
```

```
Out[114]: 5000
```

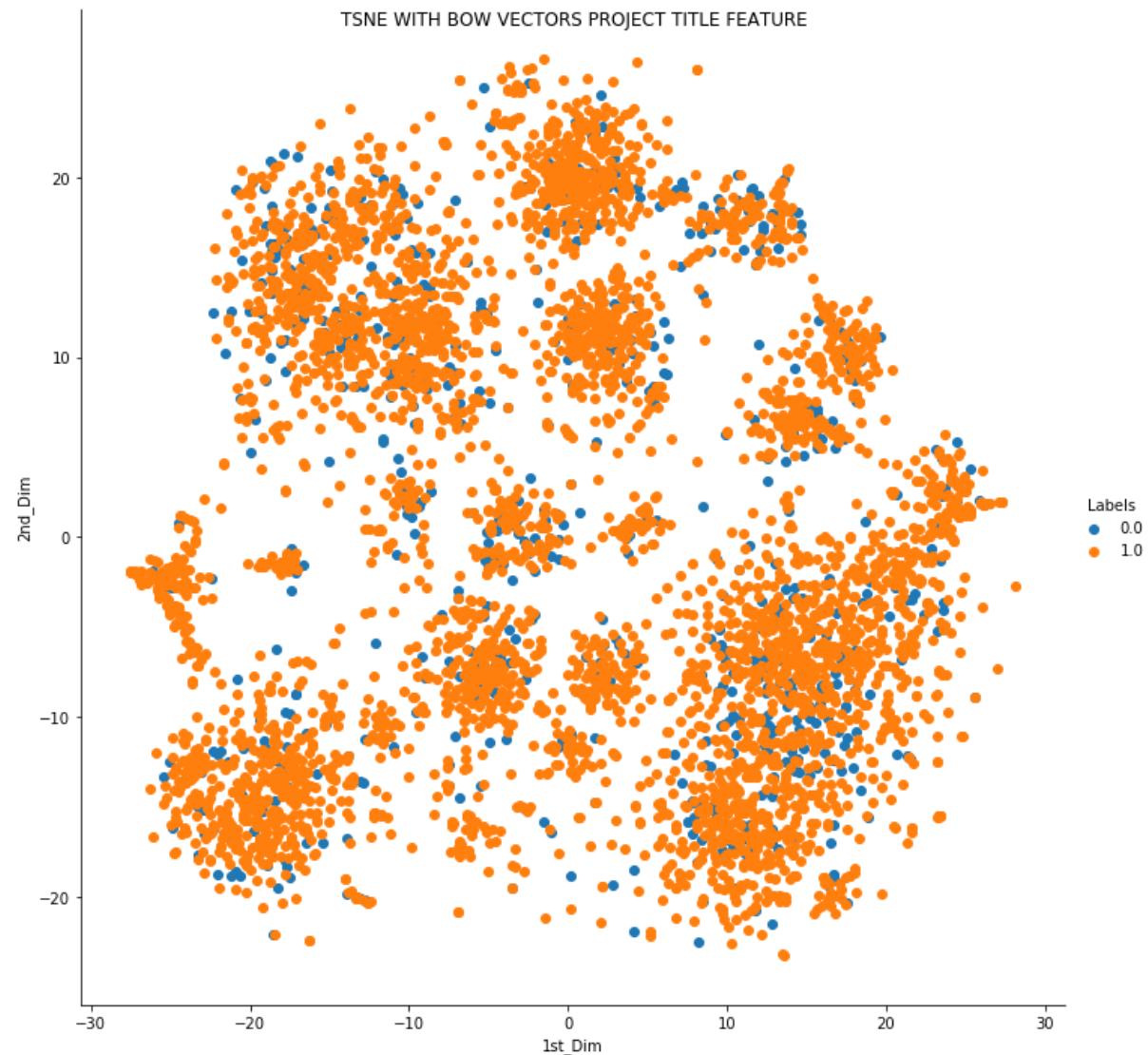
```
In [115]: tsne_data_b = np.vstack((tsne_data_b.T, labels_new)).T
tsne_df_b = pd.DataFrame(tsne_data_b, columns = ("1st_Dim","2nd_Dim","Labels"))
```

```
In [116]: tsne_df_b.shape
```

Out[116]: (5000, 3)

2.1 TSNE with BOW encoding of project_title feature

```
In [117]: sns.FacetGrid(tsne_df_b, hue = "Labels", size = 10).map(plt.scatter, "1st_Dim", "2nd_Dim")\
    .add_legend().\
    fig.suptitle("TSNE WITH BOW VECTORS PROJECT TITLE FEATURE")
    plt.show()
```



Summary :

1. We can see that from above plot most of overlapping in the datapoints.
2. The points are well scattered, unable to draw any proper conclusion.

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

```
In [118]: X = hstack((categories_one_hot, sub_categories_one_hot, school_state_categories_one_hot,
project_grade_categories_one_hot, teacher_prefix_categories_one_hot, price_standardized,
quantity_standardized, prev_projects_standardized, title_tfidf))
X.shape
```

```
Out[118]: (109248, 3433)
```

```
In [119]: X = X.tocsr()
X_new = X[0:5000,:]
```

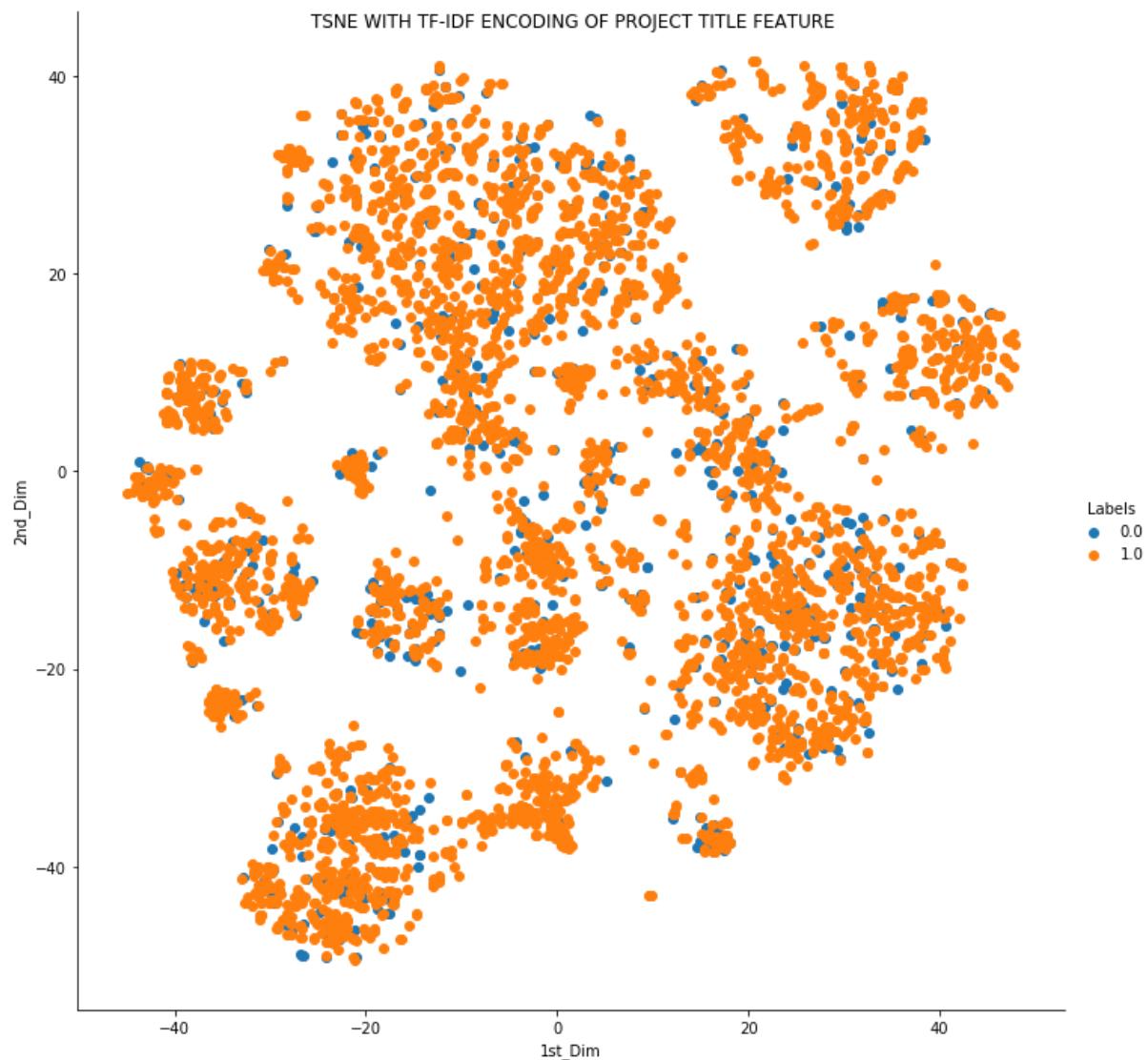
```
In [120]: X_new = X_new.toarray()
model = TSNE(n_components = 2, perplexity = 100.0, random_state = 0)
tsne_data_tfidf = model.fit_transform(X_new)
```

```
In [121]: tsne_data_tfidf = np.vstack((tsne_data_tfidf.T, labels_new)).T
tsne_df_tfidf = pd.DataFrame(tsne_data_tfidf, columns = ("1st_Dim","2nd_Dim","Labels"))
```

```
In [122]: tsne_df_tfidf.shape
```

```
Out[122]: (5000, 3)
```

```
In [123]: sns.FacetGrid(tsne_df_tfidf, hue = "Labels", size = 10).map(plt.scatter,
,"1st_Dim", "2nd_Dim").\
add_legend().fig.suptitle("TSNE WITH TF-IDF ENCODING OF PROJECT TITLE FEATURE")
plt.show()
```



Summary :

The Blue and the Orange points do not form any clusters or accumulation of any type, Hence drawing conclusions seems to quite impossible with the current state of the T-SNE data using TF - IDF Encoding.No conclusion can be find out

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

```
In [125]: X = hstack((categories_one_hot, sub_categories_one_hot, school_state_categories_one_hot,  
project_grade_categories_one_hot,\  
teacher_prefix_categories_one_hot, price_standardized, quantity_standar  
dized, prev_projects_standardized,avg_w2v_vectors_title))  
X.shape
```

```
Out[125]: (109248, 404)
```

```
In [126]: X = X.tocsr()  
X_new = X[0:5000,:]
```

```
In [127]: X_new = X_new.toarray()  
model = TSNE(n_components = 2, perplexity = 100.0, random_state = 0)  
tsne_data_avg_w2v = model.fit_transform(X_new)
```

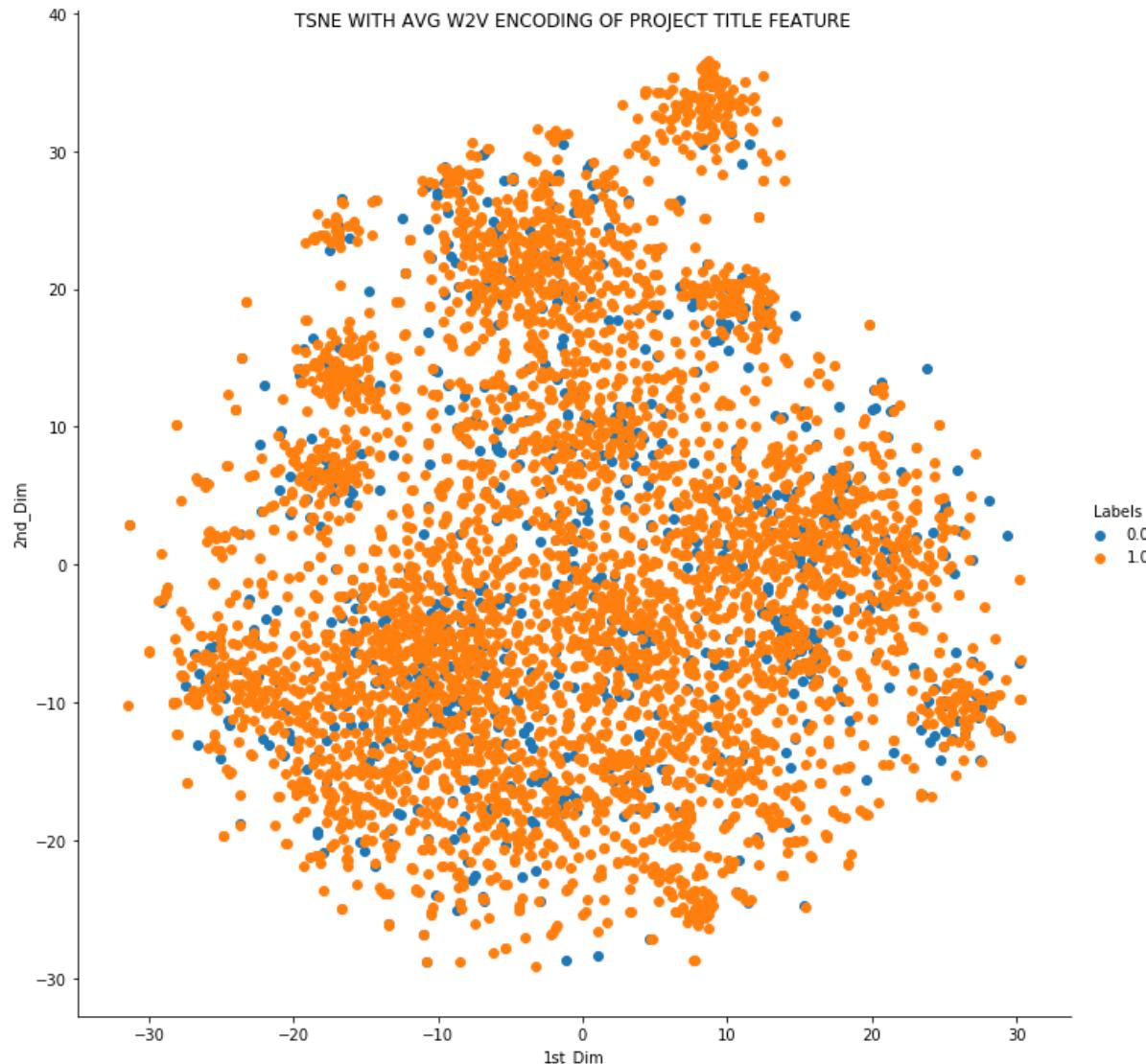
```
In [128]: tsne_data_avg_w2v = np.vstack((tsne_data_avg_w2v.T, labels_new)).T  
tsne_df_avg_w2v = pd.DataFrame(tsne_data_avg_w2v, columns = ("1st_Dim",  
"2nd_Dim", "Labels"))
```

```
In [129]: tsne_df_avg_w2v.shape
```

```
Out[129]: (5000, 3)
```

```
In [130]: sns.FacetGrid(tsne_df_avg_w2v, hue = "Labels", size = 10).map(plt.scatter, "1st_Dim", "2nd_Dim").\\  
add_legend().fig.suptitle("TSNE WITH AVG W2V ENCODING OF PROJECT TITLE")
```

```
FEATURE ")
plt.show()
```



Summary :

We cant observe any clusters for whether the Project is accepted or not accepted. Hence we are not able to achieve the desired result using Avg- Word2vec

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

```
In [131]: X = hstack((categories_one_hot, sub_categories_one_hot, school_state_categories_one_hot, project_grade_categories_one_hot, teacher_prefix_categories_one_hot, price_standardized, quantity_standardized, prev_projects_standardized, tfidf_w2v_vectors_title))
X.shape
```

Out[131]: (109248, 404)

```
In [132]: X = X.tocsr()
X_new = X[0:5000,:]
```

```
In [133]: X_new = X_new.toarray()
model = TSNE(n_components = 2, perplexity = 100.0, random_state = 0)
tsne_data_tfidf_w2v = model.fit_transform(X_new)
```

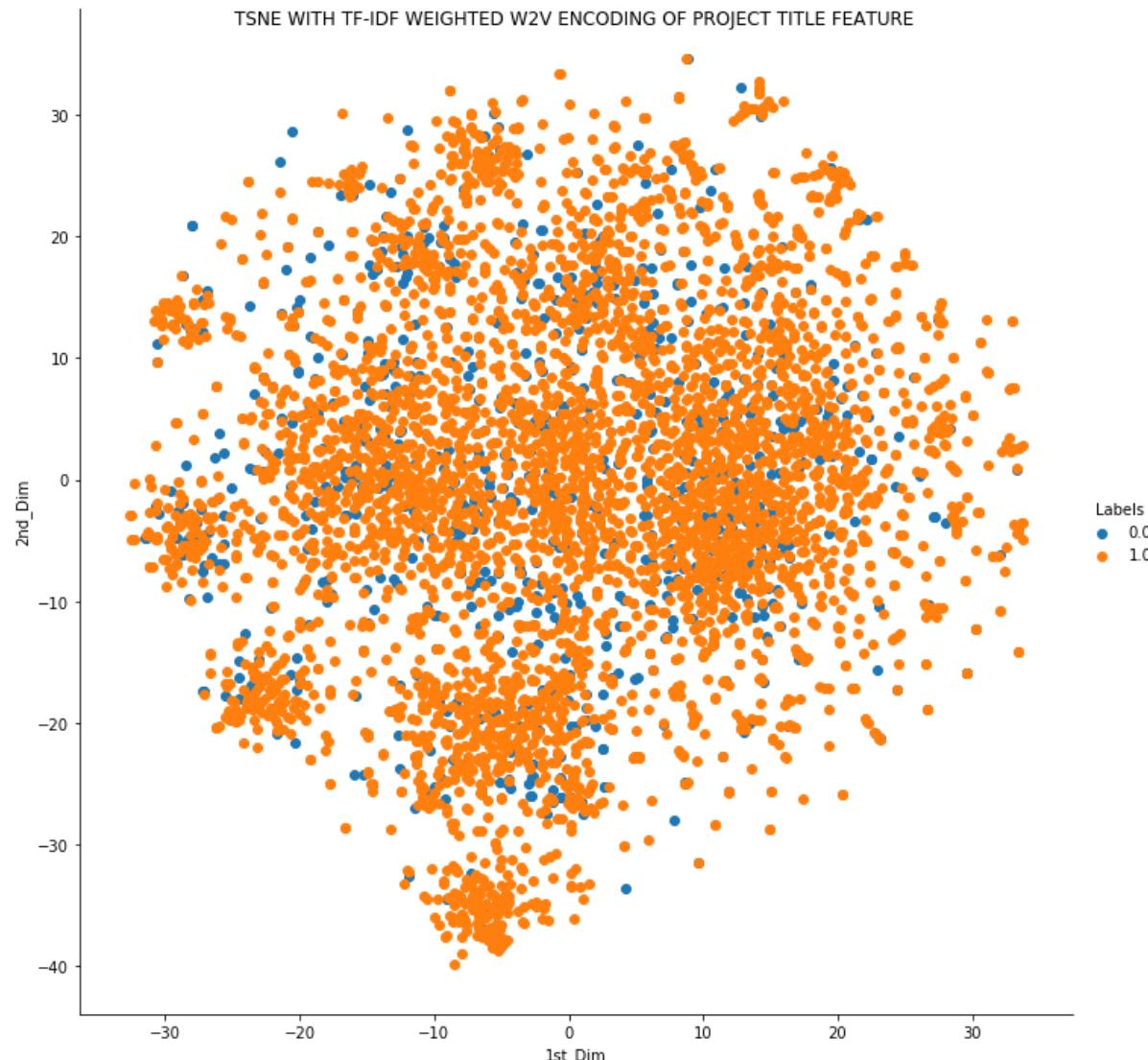
```
In [134]: tsne_data_tfidf_w2v = np.vstack((tsne_data_tfidf_w2v.T, labels_new)).T
tsne_df_tfidf_w2v = pd.DataFrame(tsne_data_tfidf_w2v, columns = ("1st_Dim", "2nd_Dim", "Labels"))
```

```
In [135]: tsne_df_tfidf_w2v.shape
```

Out[135]: (5000, 3)

```
In [137]: sns.FacetGrid(tsne_df_tfidf_w2v, hue = "Labels", size = 10).map(plt.scatter, "1st_Dim", "2nd_Dim").\nadd_legend().fig.suptitle("TSNE WITH TF-IDF WEIGHTED W2V ENCODING OF PR
```

```
OBJECT TITLE FEATURE ")
plt.show()
```



Summary :

This visualisation of TSNE with TF-IDF Weighted Word2Vec does not seem to yield the expected result of clustering similar data points. Hence we would have to try any other method

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

```
In [140]: X = hstack((categories_one_hot, sub_categories_one_hot, school_state_categories_one_hot,
project_grade_categories_one_hot, teacher_prefix_categories_one_hot, price_standardized,
quantity_standardized, prev_projects_standardized, title_bow, title_tfidf,
avg_w2v_vectors_title,
tfidf_w2v_vectors_title))
X.shape
```

```
Out[140]: (109248, 7362)
```

```
In [141]: X = X.tocsr()
X_new = X[0:5000,:]
```

```
In [142]: X_new = X_new.toarray()
model = TSNE(n_components = 2, perplexity = 100.0, random_state = 0)
tsne_data_complete = model.fit_transform(X_new)
```

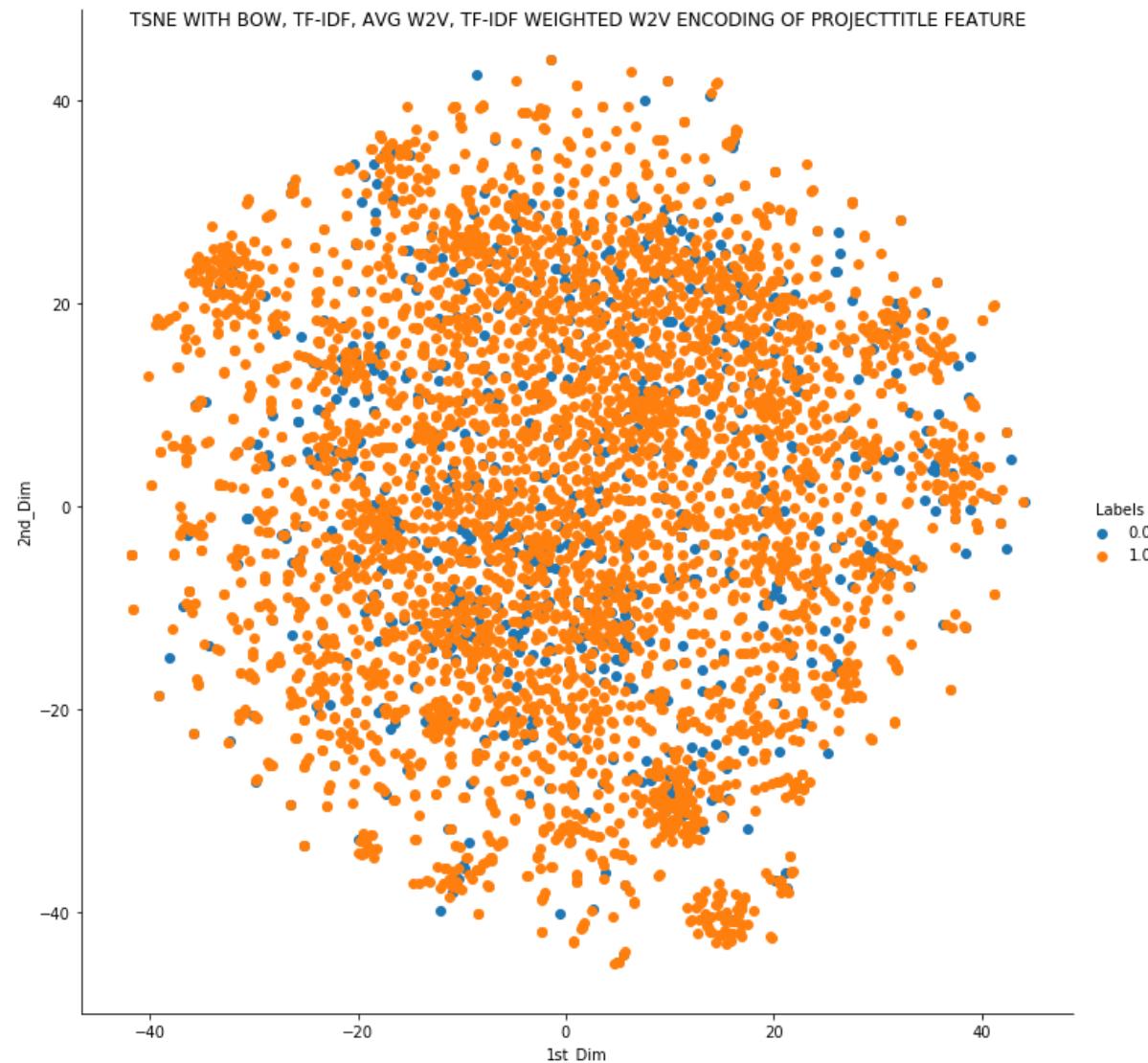
```
In [143]: tsne_data_complete = np.vstack((tsne_data_complete.T, labels_new)).T
tsne_df_complete = pd.DataFrame(tsne_data_complete, columns = ("1st_Dim",
"2nd_Dim","Labels"))
```

```
In [144]: tsne_df_complete.shape
```

```
Out[144]: (5000, 3)
```

```
In [147]: sns.FacetGrid(tsne_df_complete, hue = "Labels", size = 10).map(plt.scatter,
"1st_Dim", "2nd_Dim").\
```

```
add_subplot().set_title("TSNE WITH BOW, TF-IDF, AVG W2V, TF-IDF WEIGHTED W2V ENCODING OF PROJECTTITLE FEATURE")
plt.show()
```



Summary :

This visualisation of TSNE with Bag of Words, TF-IDF, Avg Word2Vec, TF-IDF Weighted Word2Vec does not seem to yield the expected result of clustering similar data points. Hence we would have to try any other method.

2.6 CONCLUSION

From above Observations we can Conclude that

1-Averageky there were 85% of projects from total projects are approved (got funding) while remaining 15% are non funding

2-Delaware (DE) state from the United States has the highest percent of projects accepted within the whole country having almost 90% acceptance rate, followed by North Dakota (ND) and Washington(WA) nearly 89% and 88% respectively each.

3-Vermont(VT) has the lowest Approval rate with exactly 80% followed by District of Columbia (DC) and Texas(TX) with nearly 80% and 81% respectively.

4-California has supermassive no of project proposals nearly 85% whereas Vermount has low 80%

5-Female Teachers have the maximum number of projects proposed and accepted compared to the male teachers.

6-There are alot of projects proposed for the students between Pre Kindergarten and 2nd Grade while for the rest it keeps decreasing.

7-The average Acceptance rate of the project is 84% irrespective of the Grade

8-Projects belonging to the Literacy and Language categories have the highest number of projects proposed under. The maximum number of accepted projects also belong to this category, having an acceptance rate of nearly 87%.

9-There is a lot of variability in the total number of projects proposed per Category of the project.

10-Projects belonging to both Maths and Science have acceptance rate of nearly 82% while introducing the concept of Literacy and Language to this can increase its acceptance rate to nearly 87%

11-There is also Variability in Acceptance rate, projects under the category Warmth, Care and Hunger have an acceptance rate of 93.5%.

12-The highest number of projects are registered under Literacy and Language with 52,239 projects,followed by Maths and Science having 41,421 projects whereas2.There are only 1388 projects under the category of Warmth , Care or Hunger

13-The sub-Category Literacy has the highest number of projects approved with 8371 projects. Also the acceptance rate is 88% whereas Health and Wellness have the lowest number of projects proposed with 3,583 projects only.

14-Roughly most of the projects have 3, 4 or 5 words in the title.There are hardly any project titles containing more than 10 words

15-The number of words in the Project Essays of Approved Projects are slightly more than the number of words in the Project Essays of the Rejected Projects.

16-The Maximum price for any project should be less than 10,000 dollars.The approved projects tend to have lower cost when compared to the projects that have not been approved.

17-The authorized projects tend to have reduced costs compared to non-approved projects. Looking at the percentile values,this can be noticed. For an approved project, the 50th percentile cost value is 198.99 dollars, while the cost for the unapproved projects is 263.145 dollars.

18-The project summaries containing numeric values have a very high acceptance rate of 90%. Well,proper numbered requirements suggest clarity in the proposals and hence A lot of people tend to donate for a better cause, that is to help children.

19-We notice that an average cost almost 298 dollars each project, with a standard deviation of 368 dollars. So, most of the projects are less than 750 dollars.

20-We observe that on an average Each project costs nearly 298 Dollars. The Price paid is generally for the purchase of the Items. The projects on an average require atleast 17 Different of similar items.

21-Visualisation of tSNE with Bag of Words, TF-IDF, Avg Word2Vec, TF-IDF Weighted Word2Vec does not seem to yield the expected result of clustering similar data points.Hence we have to think other method of Data Analysis.