DonorsChoose

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

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

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

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

About the DonorsChoose Data Set

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

· ·	
Description Fourth application essay	Feature project_essay_4_
Datetime when project application was submitted. Example: 2016-04-28 12:43:56.245	<pre>project_submitted_datetime</pre>
A unique identifier for the teacher of the proposed project. Example: bdf8baa8fedef6bfeec7ae4ff1c15c56	teacher_id
Teacher's title. One of the following enumerated values:	
• nan • Dr.	
• Mr.	teacher_prefix
• Mrs.	
• Ms.	
• Teacher.	
Number of project applications previously submitted by the same teacher. Example: 2	teacher_number_of_previously_posted_projects

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

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

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

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

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

Label

Description

project_is_approved

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

Notes on the Essay Data

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

- __project_essay_1:__ "Introduce us to your classroom"
- __project_essay_2:__ "Tell us more about your students"
- __project_essay_3:__ "Describe how your students will use the materials you're requesting"
- __project_essay_4:__ "Close by sharing why your project will make a difference"

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

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

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

In [1]:

```
%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
# Tutorial about Python regular expressions: https://pymotw.com/2/re/
import string
from nltk.corpus import stopwords
from nltk.stem import PorterStemmer
from nltk.stem.wordnet import WordNetLemmatizer
from gensim.models import Word2Vec
from gensim.models import KeyedVectors
import pickle
from tqdm import tqdm
import os
from plotly import plotly
import plotly.offline as offline
import plotly.graph_objs as go
offline.init_notebook_mode()
from collections import Counter
1.1 Reading Data
In [2]:
project data = pd.read csv('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)
Number of data points in train data (109248, 17)
The attributes of data: ['Unnamed: 0' 'id' 'teacher_id' 'teacher_prefix' 'school state'
 'project_submitted_datetime' 'project_grade_category'
 'project subject categories' 'project subject subcategories'
 'project_title' 'project_essay_1' 'project_essay_2' 'project_essay_3'
 'project_essay_4' 'project_resource_summary'
 'teacher number of previously posted projects' 'project is approved']
In [4]:
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[4]:
```

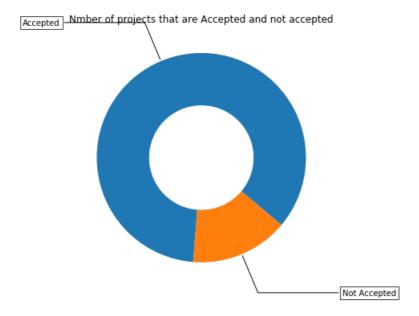
	id	description	quantity	price
0	p233245	LC652 - Lakeshore Double-Space Mobile Drying Rack	1	149.00
1	p069063	Bouncy Bands for Desks (Blue support pipes)	3	14.95

1.2 Data Analysis

```
In [5]:
```

```
# PROVIDE CITATIONS TO YOUR CODE IF YOU TAKE IT FROM ANOTHER WEBSITE.
# https://matplotlib.org/gallery/pie_and_polar_charts/pie_and_donut_labels.html#sphx-glr-gallery-p
ie-and-polar-charts-pie-and-donut-labels-py
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(arrowstyle="-"),
         bbox=bbox_props, zorder=0, va="center")
for i, p in enumerate(wedges):
   ang = (p.theta2 - p.theta1)/2. + p.theta1
    y = np.sin(np.deg2rad(ang))
    x = np.cos(np.deg2rad(ang))
   horizontalalignment = {-1: "right", 1: "left"}[int(np.sign(x))]
    connectionstyle = "angle, angleA=0, angleB={}".format(ang)
    kw["arrowprops"].update({"connectionstyle": connectionstyle})
    ax.annotate(recipe[i], xy=(x, y), xytext=(1.35*np.sign(x), 1.4*y),
                 horizontalalignment=horizontalalignment, **kw)
ax.set title ("Nmber of projects that are Accepted and not accepted")
plt.show()
```

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



1.2.1 Univariate Analysis: School State

```
temp = pd.DataFrame(project data.groupby("school state")
["project is approved"].apply(np.mean)).reset index()
print(temp)
\# if you have data which contain only 0 and 1, then the mean = percentage (think about it)
temp.columns = ['state code', 'num proposals'] #here we are changing the name of coloum
project is approved to num proposal
print("/"*76)
print(temp)
""# 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)'], [0.4, 'rgb(218,218,235)']
            [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')
```

```
school_state project_is_approved
0
            ΑK
                           0.840580
                           0.854711
1
            AΤ
2
            AR
                           0.831268
3
            ΑZ
                           0.838379
            CA
                           0.858136
4
5
            CO
                           0.841584
6
            СТ
                           0.868912
7
            DC
                           0.802326
8
            DE
                           0.897959
9
            FL
                           0.831690
10
            GA
                           0.840020
11
            ΗI
                           0.856016
12
            ΙA
                           0.852853
13
            ID
                           0.835498
14
            ΙL
                           0.852874
15
            IN
                           0.845038
16
            KS
                           0.839117
17
            ΚY
                           0.863497
18
            LA
                           0.831245
19
            MA
                           0.860193
20
            MD
                           0.838838
2.1
            ME
                           0.847525
22
            MI
                           0.845302
23
            MN
                           0.857616
24
            MO
                           0.854814
25
            MS
                           0.845049
26
            MT
                           0.816327
27
            NC
                           0.855038
                           0.888112
28
            ND
29
            NE
                           0.841424
30
            NH
                           0.873563
31
            NJ
                           0.843987
32
            NM
                           0.859964
33
            NV
                           0.853694
34
            NY
                           0.859661
```

```
36
             OK
                            0.834798
37
            OR
                           0.850242
38
            PΑ
                           0.854937
39
            RΙ
                           0.852632
40
            SC
                           0.860010
41
            SD
                           0.840000
42
            TN
                           0.850118
43
            ΤX
                           0.813142
44
            UT
                           0.836511
45
            VA
                           0.850367
46
            VT
                           0.800000
47
             WA
                            0.876178
48
            WΤ
                           0.845649
49
                           0.854871
50
             WY
                            0.836735
state code num proposals
0
          AK
                   0.840580
1
          ΔT.
                   0.854711
                   0.831268
          AR
                   0.838379
          ΑZ
3
4
          CA
                    0.858136
5
          CO
                   0.841584
          CT
                   0.868912
6
          DC
                   0.802326
8
          DE
                   0.897959
9
          FT.
                   0.831690
10
          GΑ
                   0.840020
11
          ΗI
                   0.856016
12
                   0.852853
          TΑ
13
          ID
                   0.835498
14
                   0.852874
          IL
15
          IN
                    0.845038
16
          KS
                   0.839117
17
          ΚY
                   0.863497
18
          LA
                   0.831245
19
          MA
                   0.860193
20
          MD
                   0.838838
21
          ME
                    0.847525
22
          MI
                   0.845302
23
          MN
                   0.857616
24
          MO
                   0.854814
2.5
                   0.845049
          MS
26
          MT
                    0.816327
27
          NC
                    0.855038
28
          ND
                   0.888112
29
          NE
                    0.841424
30
          NH
                   0.873563
31
          NJ
                    0.843987
32
          NM
                    0.859964
33
          NV
                   0.853694
34
          NY
                   0.859661
35
          ОН
                   0.875152
                   0.834798
36
          OK
37
          OR
                   0.850242
38
          PΑ
                    0.854937
39
          RI
                   0.852632
40
          SC
                   0.860010
41
          SD
                   0.840000
          TN
                   0.850118
42
43
          ΤX
                    0.813142
44
          IJТ
                   0.836511
45
                   0.850367
          VA
46
          VT
                    0.800000
47
          WA
                    0.876178
48
          WΙ
                    0.845649
49
          WV
                    0.854871
50
                    0.836735
          WY
```

Out[6]:

35

ОН

0.875152

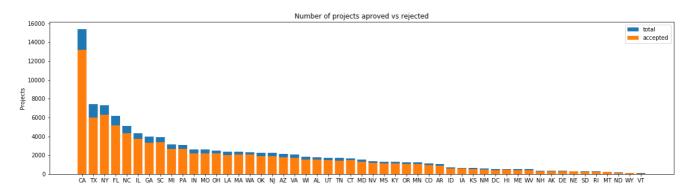
```
'USA-states\',\n text = temp[\'state_code\'],\n marker = dict(line = dict (color = \'
rgb(255,255,255)\',width = 2)),\n colorbar = dict(title = "% of pro")\n ) ]\n\nlayout = c
           title = \'Project Proposals % of Acceptance Rate by US States\',\n
            scope=\'usa\',\n projection=dict(type=\'albers usa\'),\n
akes = True, \n
                lakecolor = \rule (255, 255, 255) \rule , \ ) , \ ) \n = 
go.Figure(data=data, layout=layout)\noffline.iplot(fig, filename=\'us-map-heat-map\')\n'
                                                                                        •
In [7]:
# https://www.csi.cuny.edu/sites/default/files/pdf/administration/ops/2letterstabbrev.pdf
temp.sort values(by=['num proposals'], inplace=True)
print("States with lowest % approvals")
print(temp.head(5))
print('='*50)
print("States with highest % approvals")
print(temp.tail(5))
States with lowest % approvals
  state code num proposals
       VT
             0.800000
46
7
          DC
                  0.802326
4.3
          TX
                  0.813142
         MT
                 0.816327
26
         LA
                 0.831245
______
States with highest % approvals
state code num proposals
30 NH
                 0.873563
3.5
        OH
                 0.875152
                 0.876178
47
        WA
        ND
                 0.888112
28
         DE
                  0.897959
In [8]:
#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])
   print(ind)
   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 [9]:
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/19385591/4084039
    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()[
    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 [10]:

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

[0 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23 24 25 26 27 28 29 30 31 32 33 34 35 36 37 38 39 40 41 42 43 44 45 46 47 48 49 50]



	school_state	<pre>project_is_approved</pre>	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
==:				
==:	school state	project is approved	total	-==== Avg
39	school_state RI	project_is_approved 243	total 285	Avg 0.852632
39 26	_			3
	RI	243	285	0.852632
26	- RI MT	243	285 245	0.852632 0.816327
26 28	RI MT ND	243 200 127	285 245 143	0.852632 0.816327 0.888112

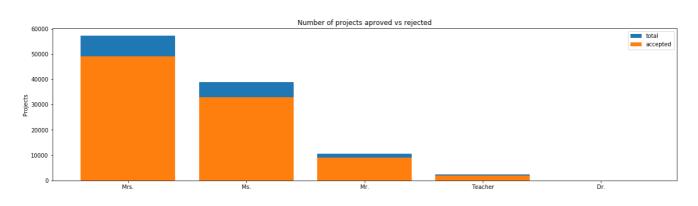
SUMMARY: Every state has greater than 80% success rate in approval

1.2.2 Univariate Analysis: teacher_prefix

In [11]:

```
univariate_barplots(project_data, 'teacher_prefix', 'project_is_approved' , top=False)
```

[0 1 2 3 4]



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

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:

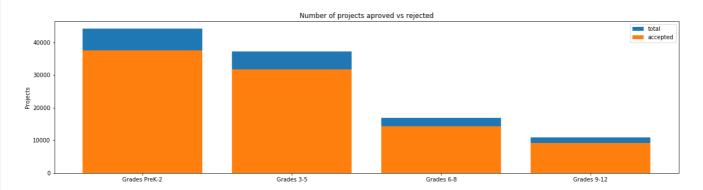
- 1. Teacher prefixes having 'Mrs.' having highest number of projects approved.
- 2. Teacher prefix having 'Dr.' having almost zero number of projects approved.
- 3. Teachers with prefixes 'Mrs.,Mr. and Ms.' have more than 80 percent project approval rates.

1.2.3 Univariate Analysis: project_grade_category

In [12]:

```
univariate_barplots(project_data, 'project_grade_category', 'project_is_approved', top=False)
```

[0 1 2 3]



Summary:

- 1. Project approval counts for Project Grade Category of Grades PreK-2 is highest.
- 2. Project approval counts for Project Grade Category of Grades 9-12 is lowest.

1.2.4 Univariate Analysis: project_subject_categories

In [13]:

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

# https://www.geeksforgeeks.org/removing-stop-words-nltk-python/
# https://stackoverflow.com/questions/23669024/how-to-strip-a-specific-word-from-a-string
# https://stackoverflow.com/questions/8270092/remove-all-whitespace-in-a-string-in-python
cat_list = []
for i in catogories:
    temp = ""
# consider we have text like this "Math & Science, Warmth, Care & Hunger"
```

In [14]:

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

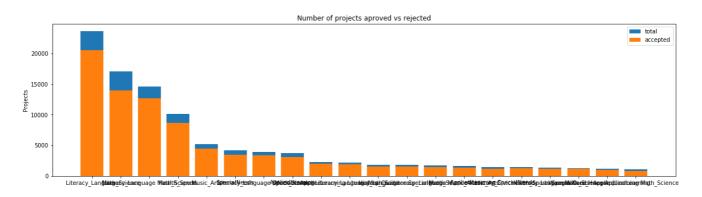
Out[14]:

_	Un	named: 0	id	teacher_id	teacher_prefix	school_state	project_submitted_datetime	project_grade_cate
	0	160221	p253737	c90749f5d961ff158d4b4d1e7dc665fc	Mrs.	IN	2016-12-05 13:43:57	Grades P
	1	140945	p258326	897464ce9ddc600bced1151f324dd63a	Mr.	FL	2016-10-25 09:22:10	Grade

In [15]:

4

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

	clean_categories	project_is_approved	total	Avg
24	Literacy_Language	20520	23655	0.867470
32	Math_Science	13991	17072	0.819529
28	Literacy_Language Math_Science	12725	14636	0.869432
8	Health Sports	8640	10177	0.848973
40	Music_Arts	4429	5180	0.855019
===				
	clean_categorie:	s project_is_approve	d total	Avg
19	History_Civics Literacy_Language	e 127	1 1421	0.894441
14	Health_Sports SpecialNeed:	121	5 1391	0.873472
50	Warmth Care_Hunge:	r 1213	2 1309	0.925898
33	Math_Science AppliedLearning	101:	9 1220	0.835246
4	AppliedLearning Math_Science	e 85!	5 1052	0.812738

Summary:

- 1. Project approval rates for all categories is more than 80 percent.
- 2. Highest success rate category is Literacy Language where is lowest is applied learning Math Science.

In [16]:

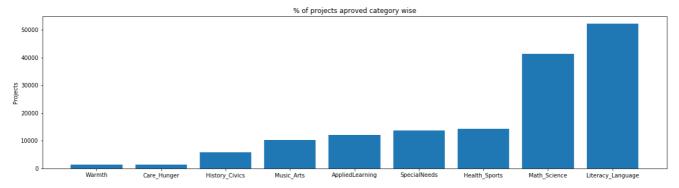
```
# count of all the words in corpus python: https://stackoverflow.com/a/22898595/4084039
from collections import Counter
my_counter = Counter()
for word in project_data['clean_categories'].values:
    my_counter.update(word.split())
```

In [17]:

```
# 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))
pl = plt.bar(ind, list(sorted_cat_dict.values()))

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



In [18]:

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

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

1.2.5 Univariate Analysis: project_subject_subcategories

In [19]:

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

# https://swww.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 i in i subject_strip_in_python in the second of the second
```

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

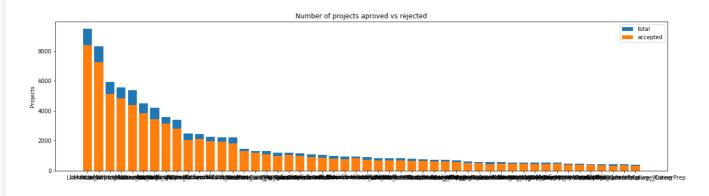
Out[20]:

	Unnamed: 0	id	teacher_id	teacher_prefix	school_state	project_submitted_datetime	project_grade_cate
C	160221	p253737	c90749f5d961ff158d4b4d1e7dc665fc	Mrs.	IN	2016-12-05 13:43:57	Grades P
1	140945	p258326	897464ce9ddc600bced1151f324dd63a	Mr.	FL	2016-10-25 09:22:10	Grade
4							<u>, </u>

In [21]:

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

[0 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23 24 25 26 27 28 29 30 31 32 33 34 35 36 37 38 39 40 41 42 43 44 45 46 47 48 49]



	clean_subcategories pro	ject_is_approved	tota	1	Avg
317	Literacy	8371	948	6 0.8	882458
319	Literacy Mathematics	7260	832	5 0.8	372072
331	Literature_Writing Mathematics	5140	592	3 0.8	367803
318	Literacy Literature_Writing	4823	557	1 0.8	365733
342	Mathematics	4385	537	9 0.8	315207
====					
	clean_subcategories	project_is_appro	oved	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: Success rate of each subcategories is more than 80 percent for project approval.

. رککی بند

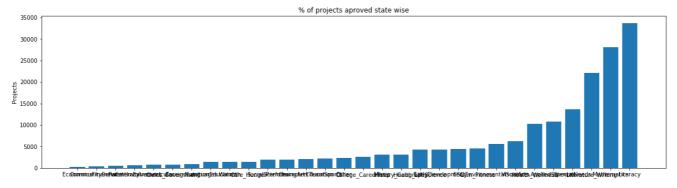
```
# count of all the words in corpus python: https://stackoverflow.com/a/22898595/4084039
from collections import Counter
my_counter = Counter()
for word in project_data['clean_subcategories'].values:
    my_counter.update(word.split())
```

In [23]:

```
# 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))
pl = plt.bar(ind, list(sorted_sub_cat_dict.values()))

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



In [24]:

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

Economics 269 CommunityService 441 FinancialLiteracy : 568 ParentInvolvement : Extracurricular : 677 Extracurricular 810 815 Civics Government : ForeignLanguages : 890 NutritionEducation : 1355 Care_Hunger 1388 1388 Care_Hunger
SocialSciences : 1920 1961 2065 CharacterEducation : TeamSports : 2192 Other 2372 2568 College_CareerPrep : Music 3145 : History_Geography 3171 4235 Health_LifeScience : EarlyDevelopment : 4254 4367 ESL 4509 Gym Fitness 5591 EnvironmentalScience: VisualArts 62.78 : Health_Wellness :
AppliedSciences :
SpecialNeeds : 10234 10816 13642 Literature_Writing : 22179 Mathematics 28074 33700 Literacy

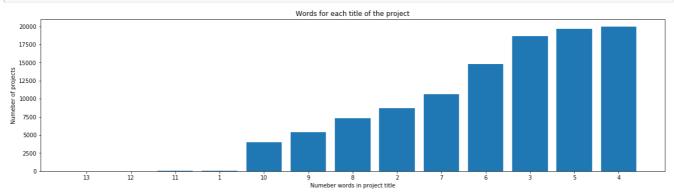
1.2.6 Univariate Analysis: Text features (Title)

In [25]:

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



Summary:

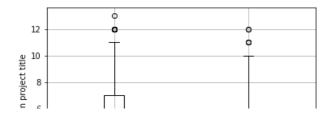
- 1. From the above plot, we can conclude that maximum number of projects is having words 4 to 5.
- 2. Very few project title is of length between 11 to 13.

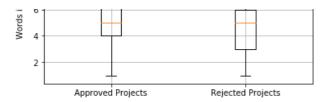
In []:

```
approved_title_word_count = project_data[project_data['project_is_approved']==1]['project_title'].
str.split().apply(len)
#print(approved_title_word_count)
approved_title_word_count = approved_title_word_count.values
print('='*75)
#print(approved_title_word_count)
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 [27]:

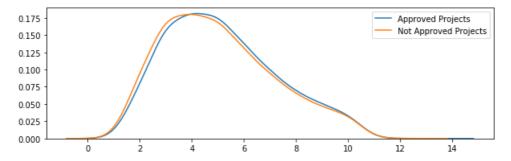
```
# https://glowingpython.blogspot.com/2012/09/boxplot-with-matplotlib.html
plt.boxplot([approved_title_word_count, rejected_title_word_count])
plt.xticks([1,2],('Approved Projects','Rejected Projects'))
plt.ylabel('Words in project title')
plt.grid()
plt.show()
```





In [28]:

```
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. From the above plot, we can conclude that Approved projects tend to have more number of words in its Project title than Rejected Projects.

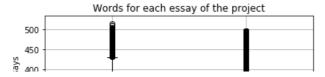
1.2.7 Univariate Analysis: Text features (Project Essay's)

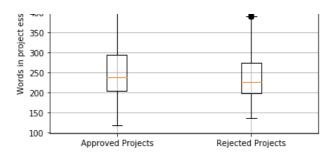
In [29]:

In [30]:

In [31]:

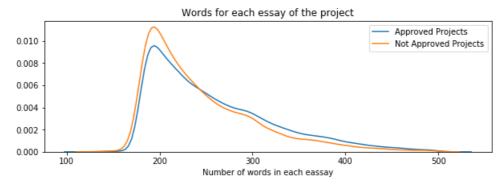
```
# https://glowingpython.blogspot.com/2012/09/boxplot-with-matplotlib.html
plt.boxplot([approved_word_count, rejected_word_count])
plt.title('Words for each essay of the project')
plt.xticks([1,2], ('Approved Projects', 'Rejected Projects'))
plt.ylabel('Words in project essays')
plt.grid()
plt.show()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





In [32]:

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



Summary:

1. From the above plot, we can conclude that Approved projects tend to have more number of words in its Project essay than Rejected Projects.

1.2.8 Univariate Analysis: Cost per project

In [33]:

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

Out[33]:

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

In [34]:

```
# https://stackoverflow.com/questions/22407798/how-to-reset-a-dataframes-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[34]:

	id	price	quantity
0	p000001	459.56	7
1	იიიიიი	515 80	21

```
id price quantity
```

In [35]:

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

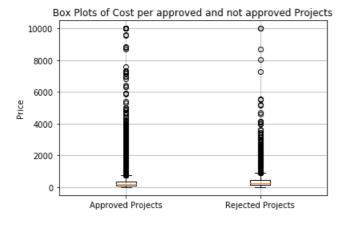
In [36]:

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

[299. 232.9 67.98 ... 239.96 73.05 109.9 ]
[154.6 516.85 219.46 ... 747. 300.18 737.95]
```

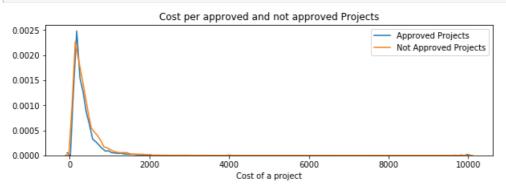
In [37]:

```
# 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 [38]:

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



```
In [39]:
```

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

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

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

for i in range(0,101,5):
    x.add_row([i,np.round(np.percentile(approved_price,i), 3), np.round(np.percentile(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. Before 100th percentile, cost of a project in Not Approved projects is more than Approved Projects.
- 2. At 100th percentile, cost of a project in Approved projects is equal to not Approved Projects.

1.2.9 Univariate Analysis: teacher_number_of_previously_posted_projects

In [40]:

```
from sklearn.preprocessing import StandardScaler

proj_data=np.array(project_data['teacher_number_of_previously_posted_projects']).reshape(-1,1)
scalerObj = StandardScaler()
xyz=scalerObj.fit_transform(proj_data)

project_data['teacher_no_of_previously_posted_projects']=pd.DataFrame(xyz)
approved_projects_previouslyposted_count = project_data[project_data['project_is_approved']==1]['teacher_no_of_previously_posted_projects']
#approved_title_word_count = approved_title_word_count.value

rejected_projects_previouslyposted_count = project_data[project_data['project_is_approved']==0]['teacher_no_of_previously_posted_projects']

C:\Users\Roshan\Anaconda3\lib\site-packages\sklearn\utils\validation.py:590:
DataConversionWarning:

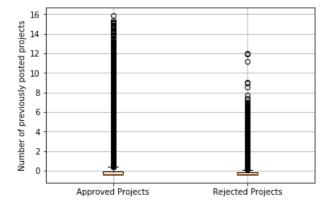
Data with input dtype int64 was converted to float64 by StandardScaler.
```

```
C:\Users\Roshan\Anaconda3\lib\site-packages\sklearn\utils\validation.py:590:
DataConversionWarning:
```

Data with input dtype int64 was converted to float64 by StandardScaler.

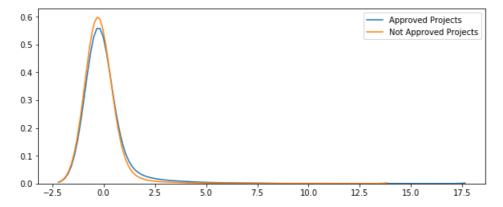
In [41]:

```
plt.boxplot([approved_projects_previouslyposted_count, rejected_projects_previouslyposted_count])
#print(project_data['teacher_number_of_previously_posted_projects'])
plt.xticks([1,2], ('Approved Projects', 'Rejected Projects'))
plt.ylabel('Number of previously posted projects')
plt.grid()
plt.show()
```



In [42]:

```
plt.figure(figsize=(10,4))
sns.kdeplot(approved_projects_previouslyposted_count,label="Approved Projects", bw=0.6)
sns.kdeplot(rejected_projects_previouslyposted_count,label="Not Approved Projects", bw=0.6)
plt.legend()
plt.show()
```



In [43]:

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

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

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

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

```
| Percentile | Approved Projects | Not Approved Projects |
```

				•		
	0		-0.402		-0.402	1
	5		-0.402		-0.402	1
	10		-0.402		-0.402	1
	15		-0.402		-0.402	1
-	20	1	-0.402		-0.402	1
	25		-0.402		-0.402	1
	30		-0.366		-0.402	1
	35		-0.366		-0.366	
	40		-0.366		-0.366	1
	45		-0.33		-0.366	
	50		-0.33		-0.33	1
	55		-0.294		-0.33	1
	60		-0.258		-0.294	
	65		-0.222		-0.294	
	70		-0.15		-0.258	1
	75		-0.078		-0.186	1
	80		0.066		-0.114	
	85		0.282		-0.006	
	90		0.679		0.21	
	95		1.651		0.715	
	100		15.835		12.019	
+-		+		+		+

Summary:

- 1. At 25th percentile, the number of Approved and Not Approved projects for the projects posted previously by teacher is almost same.
- 2. At 45th percentile, the number of projects posted previously by teacher got Approved are slighly greater than Not Approved.
- 3. At 75th percentile, the number of projects posted previously by teacher got Approved are greater than Not Approved.
- 4. At 100th percentile, the number of Approved Projects are much greater than Not Approved projects for the projects posted previously by teacher.

1.2.10 Univariate Analysis: project_resource_summary

Please do this on your own based on the data analysis that was done in the above cells

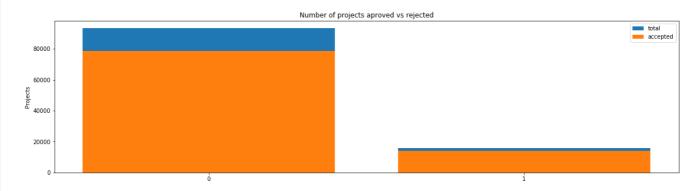
Check if the presence of the numerical digits in the project_resource_summary effects the acceptance of the project or not. If you observe that presence of the numerical digits is helpful in the classification, please include it for further process or you can ignore it.

In [44]:

```
abc=project data['project resource summary']
print(type(abc.tolist()))
#print(abc)
bool rs=[]
def num there(s):
   #return (1 if(i.isdigit()) else 0 for i in s)
    return any(i.isdigit() for i in s)
for i in abc:
   bool truFal=num there(i)
   bool_rs.append(bool_truFal)
#print(bool rs)
project data['project resource summary']=pd.DataFrame(bool rs)
project data['project resource summary'] = project data['project resource summary'].map({True: 1, F
alse: 0}) #1 which is having digits and 0 which doesnot have any digits.
#print(project data['project resource summary'])
<class 'list'>
```

```
In [45]:
```

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



```
project_resource_summary project_is_approved total
                                                Avg
                                78616 93492 0.840885
14090 15756 0.894263
0
               0
1
                    1
_____
  project_resource_summary project_is_approved total
                                                Avq
                         78616 93492 0.840885
              0
0
1
                                14090 15756 0.894263
                    1
```

Summary:

- 1. Here, 1 denotes projects resource summary which is having numerical digits and 0 which doesnot consists of any digits.
- 2. From the above plot, we can conclude that Project approval count for resource summary which doesnot have any numerical digits is **greater** than Project approval count for resource summary which have numerical digits.

1.3 Text preprocessing

1.3.1 Essay Text

```
In [46]:
```

```
project_data.head(2)
```

Out[46]:

	Unnamed: 0	id	teacher_id	teacher_prefix	school_state	project_submitted_datetime	project_grade_cate
0	160221	p253737	c90749f5d961ff158d4b4d1e7dc665fc	Mrs.	IN	2016-12-05 13:43:57	Grades P
1	140945	p258326	897464ce9ddc600bced1151f324dd63a	Mr.	FL	2016-10-25 09:22:10	Grade

2 rows × 21 columns

1

In [47]:

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

My students are English learners that are working on English as their second or third languages. W e are a melting pot of refugees, immigrants, and native-born Americans bringing the gift of langua ge to our school. \r\n\r\n We have over 24 languages represented in our English Learner program wi th 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 o f your world.\"-Ludwig Wittgenstein Our English learner's have a strong support system at home th at begs for more resources. Many times our parents are learning to read and speak English along s ide 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 hom e is able to assist. All families with students within the Level 1 proficiency status, will be a offered to be a part of this program. These educational videos will be specially chosen by the En glish 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 ed ucational dvd's for the years to come for other EL students.\r\nnannan

The 51 fifth grade students that will cycle through my classroom this year all love learning, at 1 east most of the time. At our school, 97.3% of the students receive free or reduced price lunch. O f 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 bea utiful costumes that students wear. On Cinco de Mayo we put on a big festival with crafts made by the students, dances, and games. At the end of the year the school hosts a carnival to celebrate t he hard work put in during the school year, with a dunk tank being the most popular activity.My st udents 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 hav e 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 us ed by the students who need the highest amount of movement in their life in order to stay focused on school.\r\n\r\nWhenever asked what the classroom is missing, my students always say more Hokki Stools. They can't get their fill of the 5 stools we already have. When the students are sitting i n 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 ta ken. 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. $\n \$ 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 th e 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.nannan

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.\r\n \r\nMy class is made up of 28 wonderfully unique boys and girls of mixed races in Arkansas.\r\nThey attend a Title I school, which means there is a high enough percentage of free a nd reduced-price lunch to qualify. Our school is an \"open classroom\" concept, which is very uniq ue as there are no walls separating the classrooms. These 9 and 10 year-old students are very eage r learners; they are like sponges, absorbing all the information and experiences and keep on wanti ng more.With these resources such as the comfy red throw pillows and the whimsical nautical hangin g decor and the blue fish nets, I will be able to help create the mood in our classroom setting to be one of a themed 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 pic tures 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 t o make our new school year a very successful one. Thank you!nannan

My kindergarten students have varied disabilities ranging from speech and language delays, cogniti ve delays, gross/fine motor delays, to autism. They are eager beavers and always strive to work th eir hardest working past their limitations. \n for my students. I teach in a Title I school where most of the students receive free or reduced pr ice 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 gr oove and move as you were in a meeting? This is how my kids feel all the time. The want to be able to move as they learn or so they say. Wobble chairs are the answer and I love then because they dev elop their core, which enhances gross motor and in Turn fine motor skills. \r\nThey also want to l

earn through games, my kids don't want to sit and do worksheets. They want to learn to count by ju mping and playing. Physical engagement is the key to our success. The number toss and color and sh ape 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 grea t teacher inspires. -William A. Ward\r\n\r\nMy school has 803 students which is makeup is 97.6% Af rican-American, making up the largest segment of the student body. A typical school in Dallas is m ade up of 23.2% African-American students. Most of the students are on free or reduced lunch. We a ren'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 smar t, effective, efficient, and disciplined students with good character. In our classroom we can util ize the Bluetooth for swift transitions during class. I use a speaker which doesn't amplify the so und 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 all ow 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 [48]:

```
# 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 [49]:

```
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 then because they develop their core, which enhances gross motor and in Turn fine motor skills. \r\nThey also want to learn through games, my kids do not want to sit and do worksheets. They want to learn to count by jumping and playing. Physical engagement is the key to our success. The number toss and color and shape mats can make that happen. My students will forget they are doing work and just have the fun a 6 year old deserves.nannan

In [50]:

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

My kindergarten students have varied disabilities ranging from speech and language delays, cogniti

ve derays, gross, the motor derays, to addrsm. They are eager beavers and always strive to work the eir 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 groov e and move as you were in a meeting? This is how my kids feel all the time. The want to be able to move as they learn or so they say. Wobble chairs are the answer and I love then because they develop their core, which enhances gross motor and in Turn fine motor skills. 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 [51]:

```
#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 The want to be able to move as the ey learn or so they say Wobble chairs are the answer and I love then because they develop their compared to the 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 nan name.

In [52]:

```
# 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', 'have', 'has', 'had', 'having',
'do', 'does', \
                           'did', 'doing', 'a', 'an', 'the', 'and', 'but', 'if', 'or', 'because', 'as', 'until', '
while', 'of', \
                           'at', 'by', 'for', 'with', 'about', 'against', 'between', 'into', 'through', 'during',
'before', 'after',\
                           'above', 'below', 'to', 'from', 'up', 'down', 'in', 'out', 'on', 'off', 'over', 'under'
, 'again', 'further',\
                           'then', 'once', 'here', 'there', 'when', 'why', 'how', 'all', 'any', 'both', '\epsilon
ach', 'few', 'more',\
                           'most', 'other', 'some', 'such', 'only', 'own', 'same', 'so', 'than', 'too', 'very', \
                           's', 't', 'can', 'will', 'just', 'don', "don't", 'should', "should've", 'now', 'd', 'll'
, 'm', 'o', 're', \
                           've', 'y', 'ain', 'aren', "aren't", 'couldn', "couldn't", 'didn', "didn't", 'doesn', "doesn', "doesn',
esn't", 'hadn',\
                           "hadn't", 'hasn', "hasn't", 'haven', "haven't", 'isn', "isn't", 'ma', 'mightn',
"mightn't", 'mustn',\
                          "mustn't", 'needn', "needn't", 'shan', "shan't", 'shouldn', "shouldn't", 'wasn',
"wasn't", 'weren', "weren't", \
                           'won', "won't", 'wouldn', "wouldn't"]
                                                                                                                                                                                                                        I
```

In [531:

```
# Combining all the above statemennts
from tqdm import tqdm
preprocessed_essays = []
# tqdm is for printing the status bar
for sentance in tqdm(project_data['essay'].values):
    sent = decontracted(sentance)
```

```
sent = sent.replace('\\r', '')
sent = sent.replace('\\n', '')
sent = sent.replace('\\n', '')
sent = re.sub('[^A-Za-z0-9]+', '', sent)
# https://gist.github.com/sebleier/554280
sent = ''.join(e for e in sent.split() if e not in stopwords)
preprocessed_essays.append(sent.lower().strip())

100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 1
```

In [54]:

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

Out[54]:

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

1.3.2 Project title Text

In [55]:

```
# similarly you can preprocess the titles also
# Combining all the above statemennts
from tqdm import tqdm
preprocessed titles = []
# tqdm is for printing the status bar
for sentance in tqdm(project data['project title'].values):
   sent = decontracted(sentance)
   sent = sent.replace('\\r', ' ')
   sent = sent.replace('\\"', ' ')
   sent = sent.replace('\\n', '')
   sent = re.sub('[^A-Za-z0-9]+', '', sent)
   # https://gist.github.com/sebleier/554280
   sent = ' '.join(e for e in sent.split() if e not in stopwords)
   preprocessed_titles.append(sent.lower().strip())
                                                                          109248/109248
[00:06<00:00, 16800.12it/s]
```

In [56]:

```
preprocessed_titles[25000]
```

Out [56]:

'moving to learn creative expression'

1. 4 Preparing data for models

```
In [57]:
```

```
project_data.columns
```

```
Out[57]:
```

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',
       'teacher no of previously posted projects'],
      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

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

In [58]:
# we use count vectorizer to convert the values into one hot encoded features
#clean categories
from sklearn.feature extraction.text import CountVectorizer
vectorizer = CountVectorizer(vocabulary=list(sorted cat dict.keys()), lowercase=False, binary=True
vectorizer.fit(project data['clean categories'].values)
print(vectorizer.get feature names())
categories_one_hot = vectorizer.transform(project_data['clean_categories'].values)
print("Shape of matrix after one hot encodig ",categories_one_hot.shape)
['Warmth', 'Care Hunger', 'History Civics', 'Music Arts', 'AppliedLearning', 'SpecialNeeds',
'Health Sports', 'Math Science', 'Literacy Language']
Shape of matrix after one hot encodig (109248, 9)
In [59]:
# we use count vectorizer to convert the values into one hot encoded features
#clean subcategories
vectorizer = CountVectorizer(vocabulary=list(sorted sub cat dict.keys()), lowercase=False, binary=
vectorizer.fit(project data['clean subcategories'].values)
print(vectorizer.get feature names())
sub categories one hot = vectorizer.transform(project data['clean subcategories'].values)
print("Shape of matrix after one hot encodig ", sub categories one hot.shape)
['Economics', 'CommunityService', 'FinancialLiteracy', 'ParentInvolvement', 'Extracurricular',
'Civics Government', 'ForeignLanguages', 'NutritionEducation', 'Warmth', 'Care Hunger',
```

'SocialSciences', 'PerformingArts', 'CharacterEducation', 'TeamSports', 'Other',

'SpecialNeeds', 'Literature_Writing', 'Mathematics', 'Literacy']

Shape of matrix after one hot encodig (109248, 30)

'College_CareerPrep', 'Music', 'History_Geography', 'Health_LifeScience', 'EarlyDevelopment', 'ESL', 'Gym Fitness', 'EnvironmentalScience', 'VisualArts', 'Health Wellness', 'AppliedSciences',

```
In [60]:
# 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['school state'].values:
   my counter.update(word.split())
# dict sort by value python: https://stackoverflow.com/a/613218/4084039
state dict = dict(my counter)
sorted_state_dict = dict(sorted(state_dict.items(), key=lambda kv: kv[1]))
In [61]:
# Please do the similar feature encoding with state, teacher_prefix and project_grade_category als
from sklearn.feature_extraction.text import CountVectorizer
statevectorizer = CountVectorizer(vocabulary=list(sorted_state_dict.keys()), lowercase=False, bina
rv=True)
statevectorizer.fit(project data['school state'].values)
print(statevectorizer.get_feature_names())
state one hot = statevectorizer.transform(project data['school state'].values)
print("Shape of matrix after one hot encodig ", state one hot.shape)
['VT', 'WY', 'ND', 'MT', 'RI', 'SD', 'NE', 'DE', 'AK', 'NH', 'WV', 'ME', 'HI', 'DC', 'NM', 'KS', 'I
A', 'ID', 'AR', 'CO', 'MN', 'OR', 'KY', 'MS', 'NV', 'MD', 'CT', 'TN', 'UT', 'AL', 'WI', 'VA', 'AZ',
    , 'OK', 'WA', 'MA', 'LA', 'OH', 'MO', 'IN', 'PA', 'MI', 'SC', 'GA', 'IL', 'NC', 'FL', 'NY', 'T%
', 'CA'1
Shape of matrix after one hot encodig (109248, 51)
In [62]:
# 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['project_grade_category'].values:
    my counter.update(word.split('\n'))
# dict sort by value python: https://stackoverflow.com/a/613218/4084039
grade dict = dict(my counter)
sorted grade dict = dict(sorted(grade dict.items(), key=lambda kv: kv[1]))
In [63]:
# Please do the similar feature encoding with state, teacher prefix and project grade category als
from sklearn.feature_extraction.text import CountVectorizer
statevectorizer = CountVectorizer(vocabulary=list(sorted grade dict.keys()), lowercase=False, bina
statevectorizer.fit(project data['project grade category'].values)
print(statevectorizer.get feature names())
grade one hot = statevectorizer.transform(project data['project grade category'].values)
print("Shape of matrix after one hot encodig ",grade one hot.shape)
['Grades 9-12', 'Grades 6-8', 'Grades 3-5', 'Grades PreK-2']
Shape of matrix after one hot encodig (109248, 4)
In [107]:
# 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['teacher prefix'].values:
```

my counter.update(str(word).split('\n'))

```
print("=="*50)

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

-

```
In [65]:
unique_Data=project_data['teacher_prefix'].unique()
print(unique_Data)

#replacing nan with empty string
project_data.teacher_prefix=project_data.teacher_prefix.fillna('')
uniqueData=project_data['teacher_prefix'].unique()
print(uniqueData)

vectorizer = CountVectorizer(vocabulary=list(uniqueData).remove(""), lowercase=False, binary=True)
vectorizer.fit(project_data['teacher_prefix'].values)
#vectorizer.fit(project_data['teacher_prefix'].values.astype('U'))
print(vectorizer.get_feature_names())

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

['Mrs.' 'Mr.' 'Ms.' 'Teacher' nan 'Dr.']
['Mrs.' 'Mr.' 'Ms.' 'Teacher' '' 'Dr.']
['Dr', 'Mr', 'Mrs', 'Ms', 'Teacher']
```

1.4.2 Vectorizing Text data

1.4.2.1 Bag of words

```
In [66]:
```

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

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

Shape of matrix after one hot encoding (109248, 5)

1.4.2.2 Bag of Words on `project_title`

```
In [67]:
```

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

In [68]:

```
# Similarly you can vectorize for title also
vectorizer_titles = CountVectorizer(min_df=10)
text_bow_titles = vectorizer.fit_transform(preprocessed_titles)
print("Shape of matrix after one hot encodig ",text_bow_titles.shape)
```

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

4.400 TEIDE . . .

1.4.2.3 I FIDF vectorizer

In [69]:

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

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

1.4.2.4 TFIDF Vectorizer on 'project title'

In [70]:

```
# Similarly you can vectorize for title also
from sklearn.feature_extraction.text import TfidfVectorizer
vectorizer = TfidfVectorizer(min_df=10)
text_tfidf_titles = vectorizer.fit_transform(preprocessed_titles)
print("Shape of matrix after one hot encodig ",text_tfidf.shape)
```

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

1.4.2.5 Using Pretrained Models: Avg W2V

In [71]:

```
# Reading glove vectors in python: https://stackoverflow.com/a/38230349/4084039
def loadGloveModel(gloveFile):
   print ("Loading Glove Model")
   f = open(gloveFile,'r', encoding="utf8")
   model = {}
   for line in tqdm(f):
       splitLine = line.split()
       word = splitLine[0]
       embedding = np.array([float(val) for val in splitLine[1:]])
       model[word] = embedding
    print ("Done.",len(model)," words loaded!")
   return model
model = loadGloveModel('glove.42B.300d.txt')
'''Output:
Loading Glove Model
1917495it [06:32, 4879.69it/s]
Done. 1917495 words loaded!
# -----
words = []
#for i in preproced texts:
# words.extend(i.split(' '))
for i in preprocessed essays:
   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-sa
ve-and-load-variables-in-python/
import pickle
with open('glove_vectors', 'wb') as f:
    pickle.dump(words_courpus, f)
```

Loading Glove Model

```
Done. 1917495 words loaded!

all the words in the coupus 16540843
the unique words in the coupus 56381
The number of words that are present in both glove vectors and our coupus 49637 ( 88.039 %)
word 2 vec length 49637

In [72]:

# stronging variables into pickle files python: http://www.jessicayung.com/how-to-use-pickle-to-sa
ve-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 [73]:

```
# average Word2Vec
# compute average word2vec for each review.
avg_w2v_vectors = []; # the avg-w2v for each sentence/review is stored in this list
\textbf{for} \ \texttt{sentence} \ \textbf{in} \ \texttt{tqdm} \ (\texttt{preprocessed\_essays}) : \ \textit{\#} \ \textit{for each review/sentence}
    vector = np.zeros(300) # as word vectors are of zero length
    cnt words =0; # num of words with a valid vector in the sentence/review
    for word in sentence.split(): # for each word in a review/sentence
        if word in glove words:
             vector += model[word]
             cnt words += 1
    if cnt_words != 0:
        vector /= cnt words
    avg w2v vectors.append(vector)
print(len(avg_w2v_vectors))
print(len(avg w2v vectors[0]))
                                                                               109248/109248
100%|
[01:37<00:00, 1120.24it/s]
```

109248

1.4.2.6 Using Pretrained Models: AVG W2V on `project_title`

In [74]:

```
# Similarly you can vectorize for title also

# Reading glove vectors in python: https://stackoverflow.com/a/38230349/4084039

def loadGloveModel(gloveFile):
    print ("Loading Glove Model")
    f = open(gloveFile,'r', encoding="utf8")
    model = {}
    for line in tqdm(f):
        splitLine = line.split()
        word = splitLine[0]
        embedding = np.array([float(val) for val in splitLine[1:]])
        model[word] = embedding
    print ("Done.",len(model)," words loaded!")
    return model
```

```
model = loadGloveModel('glove.42B.300d.txt')
# ================
'''Output:
Loading Glove Model
1917495it [06:32, 4879.69it/s]
Done. 1917495 words loaded!
words = []
#for i in preproced texts:
   words.extend(i.split(' '))
for i in preprocessed 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", <math>\setminus
      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-sa
ve-and-load-variables-in-python/
import pickle
with open('glove vectors', 'wb') as f:
    pickle.dump(words_courpus, f)
Loading Glove Model
1917495it [11:35, 2758.30it/s]
Done. 1917495 words loaded!
all the words in the coupus 473570
the unique words in the coupus 16903
The number of words that are present in both glove vectors and our coupus 15917 ( 94.167 %)
word 2 vec length 15917
In [75]:
```

```
the unique words in the coupus 16903
The number of words that are present in both glove vectors and our coupus 15917 ( 94.167 %)
word 2 vec length 15917

In [75]:

# stronging variables into pickle files python: http://www.jessicayung.com/how-to-use-pickle-to-sa
ve-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 [76]:

```
# average Word2Vec
# compute average word2vec for each review.
avg_w2v_vectors_titles = []; # 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_titles.append(vector)

print(len(avg_w2v_vectors_titles))
print(len(avg_w2v_vectors_titles[0]))

100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 1
```

1.4.2.7 Using Pretrained Models: TFIDF weighted W2V

In [77]:

```
# 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 [78]:

```
# average Word2Vec
# compute average word2vec for each review.
tfidf w2v vectors = []; # the avg-w2v for each sentence/review is stored in this list
for sentence in tqdm(preprocessed essays): # for each review/sentence
   vector = np.zeros(300) # as word vectors are of zero length
   tf idf weight =0; # num of words with a valid vector in the sentence/review
   for word in sentence.split(): # for each word in a review/sentence
       if (word in glove_words) and (word in tfidf_words):
           vec = model[word] # getting the vector for each word
            # here we are multiplying idf value(dictionary[word]) and the tf
value((sentence.count(word)/len(sentence.split())))
           tf idf = dictionary[word]*(sentence.count(word)/len(sentence.split())) # getting the tf
idf value for each word
           vector += (vec * tf idf) # calculating tfidf weighted w2v
            tf_idf_weight += tf_idf
   if tf idf weight != 0:
       vector /= tf idf weight
   tfidf_w2v_vectors.append(vector)
print(len(tfidf w2v vectors))
print(len(tfidf w2v vectors[0]))
                                                                             | 109248/109248
[10:08<00:00, 179.53it/s]
```

109248 300

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

In [79]:

```
# Similarly you can vectorize for title also
```

In [80]:

```
# S = ["abc def pqr", "def def def abc", "pqr pqr def"]
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 [81]:
# average Word2Vec
# compute average word2vec for each review.
tfidf w2v vectors titles = []; # 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())) # getting the tf
idf 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 titles.append(vector)
print(len(tfidf w2v vectors titles))
print(len(tfidf_w2v_vectors_titles[0]))
[00:10<00:00, 9948.25it/s]
109248
```

1.4.3 Vectorizing Numerical features

check this one: https://www.youtube.com/watch?v=0HOqOcln3Z4&t=530s

In [82]:

300

```
# standardization sklearn: https://scikit-
learn.org/stable/modules/generated/sklearn.preprocessing. Standard Scaler.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=[725.05 213.03 329. ... 399.
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(price scalar.var [0])}")
# Now standardize the data with above maen and variance.
price standardized = price scalar.transform(project data['price'].values.reshape(-1, 1))
Mean : 298.1193425966608, Standard deviation : 367.49634838483496
In [93]:
price standardized
price standardized.shape
Out[93]:
(109248, 1)
```

For quantity feature

```
In [94]:
```

```
qtv scalar = StandardScaler()
```

```
qty scalar.fit(resource data['quantity'].values.reshape(-1,1)) # finding the mean and standard
deviation of this data
print(f"Mean : {qty_scalar.mean_[0]}, Standard deviation : {np.sqrt(qty_scalar.var_[0])}")
# Now standardize the data with above maen and variance.
qty standardized = qty scalar.transform(resource data['quantity'].values.reshape(-1, 1))
C:\Users\Roshan\Anaconda3\lib\site-packages\sklearn\utils\validation.py:590:
DataConversionWarning:
Data with input dtype int64 was converted to float64 by StandardScaler.
Mean: 2.86050937148018, Standard deviation: 7.570342469523774
C:\Users\Roshan\Anaconda3\lib\site-packages\sklearn\utils\validation.py:590:
DataConversionWarning:
Data with input dtype int64 was converted to float64 by StandardScaler.
In [95]:
qty std=qty standardized[0:109248]
qty_std
Out [95]:
array([[-0.24576291],
       [ 0.01842593],
       [-0.24576291],
       [-0.24576291],
       [ 0.01842593],
       [ 0.01842593]])
In [96]:
qty_std.shape
Out[96]:
(109248, 1)
```

For summary feature

```
In [97]:
```

```
projNumber_scalar = StandardScaler()
projNumber_scalar.fit(project_data['teacher_number_of_previously_posted_projects'].values.reshape(
-1,1)) # finding the mean and standard deviation of this data
print(f"Mean : {projNumber_scalar.mean_[0]}, Standard deviation :
{np.sqrt(projNumber_scalar.var_[0])}")

# Now standardize the data with above maen and variance.
projNumber_standardized =
projNumber_scalar.transform(project_data['teacher_number_of_previously_posted_projects'].values.re
shape(-1, 1))

C:\Users\Roshan\Anaconda3\lib\site-packages\sklearn\utils\validation.py:590:
DataConversionWarning:

Data with input dtype int64 was converted to float64 by StandardScaler.

Mean : 11.153165275336848, Standard deviation : 27.77702641477403
```

```
Data with input dtype int64 was converted to float64 by StandardScaler.
In [98]:
projNumber standardized
Out[98]:
array([[-0.40152481],
       [-0.14951799],
       [-0.36552384],
       [-0.29352189],
       [-0.40152481],
       [-0.40152481]])
1.4.4 Merging all the above features
 · we need to merge all the numerical vectors i.e catogorical, text, numerical vectors
In [99]:
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 [100]:
# merge two sparse matrices: https://stackoverflow.com/a/19710648/4084039
from scipy.sparse import hstack
from scipy.sparse import coo matrix
# with the same hstack function we are concatinating a sparse matrix and a dense matirx :)
X first = hstack((categories one hot, sub categories one hot, state one hot,
grade one hot, teacher prefix one hot, text bow titles, price standardized, qty std, proj Number standard
ized))
X_first.shape
x_bow=X_first.tocsr()[0:5000,:]
X_sec = hstack((categories_one_hot, sub_categories_one_hot, state_one_hot,
grade_one_hot,teacher_prefix_one_hot,text_tfidf_titles,price_standardized,qty_std,projNumber_standa
rdized))
x tfidf=X sec.tocsr()[0:5000,:]
X third = hstack((categories one hot, sub categories one hot, state one hot,
grade one hot, teacher prefix one hot, avg w2v vectors titles, price standardized, qty std, projNumber &
tandardized))
x avgw2v=X third.tocsr()[0:5000,:]
X fourth = hstack((categories one hot, sub categories one hot, state one hot,
grade one hot, teacher prefix one hot, tfidf w2v vectors titles, price standardized, qty std, proj Number
standardized))
x tfidfw2v=X fourth.tocsr()[0:5000,:]
4
In [101]:
# merge two sparse matrices: https://stackoverflow.com/a/19710648/4084039
from scipy.sparse import hstack
# with the same hstack function we are concatinating a sparse matrix and a dense matirx :)
X = hstack((categories one hot, sub categories one hot, text bow, price standardized,))
```

C. \USELS \NUSHall \MHACUHUAS \LID \SILE PACKAYES \SKIEALH \UCLIS \VAILUACIOH.PY.JJU.

```
X.shape
Out[101]:
(109248, 16663)
```

Assignment 2: Apply TSNE

If you are using any code snippet from the internet, you have to provide the reference/citations, as we did in the above cells. Otherwise, it will be treated as plagiarism without citations.

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

In [102]:

```
'''# 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 into dense matrix
for tsne = np.hstack((X embedding, y.reshape(-1,1)))
for\_tsne\_df = pd.DataFrame (data=for\_tsne, columns=['Dimension\_x','Dimension\_y','Score'])
colors = {0:'red', 1:'blue', 2:'green'}
plt.scatter(for\ tsne\ df['Dimension\ x'],\ for\ tsne\ df['Dimension\ y'],
c=for_tsne_df['Score'].apply(lambda x: colors[x]))
plt.show()'''
```

Out[102]:

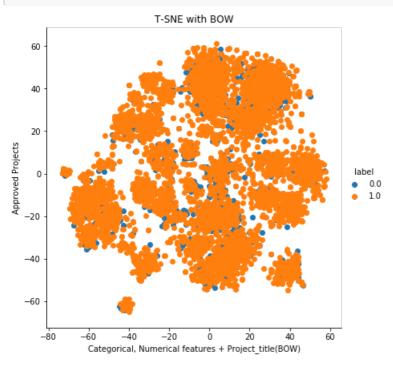
```
"# this is the example code for TSNE\nimport numpy as np\nfrom sklearn.manifold import TSNE\nfrom sklearn import datasets\nimport pandas as pd\nimport matplotlib.pyplot as plt\n\n# iris = datasets.load_iris()\nx = iris['data']\ny = iris['target']\n\ntsne = TSNE(n_components=2, perplexity=30, learning_rate=200)\n\nX_embedding = tsne.fit_transform(x)\n# 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 into dense matrix\n\nfor_tsne = np.hstack((X_embedding, y.reshape(-1,1)))\nfor_tsne_df = pd.DataFrame(data=for_tsne, columns=
```

```
['Dimension_x','Dimension_y','Score'])\ncolors = {0:'red', 1:'blue',
2:'green'}\nplt.scatter(for_tsne_df['Dimension_x'], for_tsne_df['Dimension_y'],
c=for_tsne_df['Score'].apply(lambda x: colors[x]))\nplt.show()"
In []:
```

2.1 TSNE with `BOW` encoding of `project_title` feature

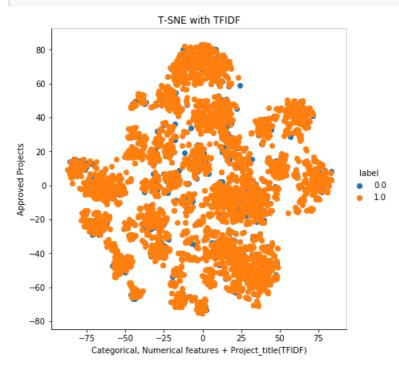
```
In [103]:
```

```
# please write all of the code with proper documentation and proper titles for each subsection
# when you plot any graph make sure you use
   # a. Title, that describes your plot, this will be very helpful to the reader
    # b. Legends if needed
    # c. X-axis label
    # d. Y-axis label
import numpy as np
from sklearn.manifold import TSNE
import seaborn as sns
from sklearn import datasets
import pandas as pd
import matplotlib.pyplot as plt
y = project_data['project_is_approved']
y bow=y[0:5000]
model=TSNE(n components=2,random state=0, perplexity=30, n iter=2000)
#model=TSNE(n_components=2,random_state=0)
tsne data=model.fit transform(x bow.toarray())
tsne_data=np.vstack((tsne_data.T,y_bow)).T
tsne_df=pd.DataFrame(data=tsne_data,columns=("Categorical, Numerical features +
Project_title(BOW)","Approved Projects","label"))
sns.FacetGrid(tsne df,hue="label",height=6).map(plt.scatter,"Categorical, Numerical features +
Project title(BOW)","Approved Projects").add legend();
plt.title('T-SNE with BOW')
plt.show()
```



2.2 TSNE with `TFIDF` encoding of `project_title` feature

```
# please write all the code with proper documentation, and proper titles for each subsection
# when you plot any graph make sure you use
    # a. Title, that describes your plot, this will be very helpful to the reader
    # b. Legends if needed
    # c. X-axis label
    # d. Y-axis label
import numpy as np
from sklearn.manifold import TSNE
import seaborn as sns
from sklearn import datasets
import pandas as pd
import matplotlib.pyplot as plt
y = project_data['project_is_approved']
y bow=y[0:5000]
#model=TSNE(n_components=2,random_state=0, perplexity=50, n_iter=5000)
model=TSNE(n components=2, random state=0)
tsne data=model.fit transform(x tfidf.toarray())
tsne data=np.vstack((tsne data.T,y bow)).T
tsne_df=pd.DataFrame(data=tsne_data,columns=("Categorical, Numerical features +
Project title(TFIDF)", "Approved Projects", "label"))
sns.FacetGrid(tsne df,hue="label",height=6).map(plt.scatter,"Categorical, Numerical features +
Project title(TFIDF)","Approved Projects").add legend();
plt.title('T-SNE with TFIDF')
plt.show()
```



2.3 TSNE with `AVG W2V` encoding of `project_title` feature

```
In [105]:
```

```
# please write all the code with proper documentation, and proper titles for each subsection
# when you plot any graph make sure you use
# a. Title, that describes your plot, this will be very helpful to the reader
# b. Legends if needed
# c. X-axis label
# d. Y-axis label
import numpy as np
from sklearn.manifold import TSNE
import seaborn as sns
from sklearn import datasets
```

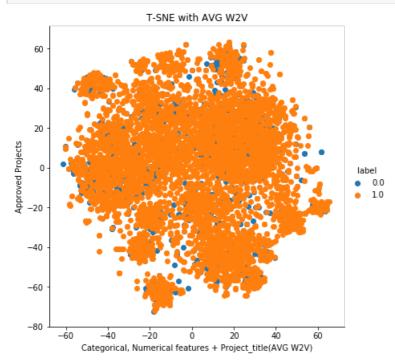
```
import pandas as pd
import matplotlib.pyplot as plt

y = project_data['project_is_approved']
y_bow=y[0:5000]
#model=TSNE(n_components=2,random_state=0, perplexity=50, n_iter=5000)
model=TSNE(n_components=2,random_state=0)
tsne_data=model.fit_transform(x_avgw2v.toarray())

tsne_data=mp.vstack((tsne_data.T,y_bow)).T

tsne_data=np.vstack((tsne_data.T,y_bow)).T

tsne_df=pd.DataFrame(data=tsne_data,columns=("Categorical, Numerical features + Project_title(AVG W2V)", "Approved Projects", "label"))
sns.FacetGrid(tsne_df,hue="label",height=6).map(plt.scatter, "Categorical, Numerical features + Project_title(AVG W2V)", "Approved Projects").add_legend();
plt.title('T-SNE with AVG W2V')
plt.show()
```

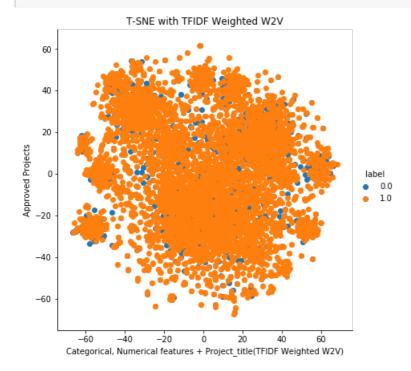


2.4 TSNE with `TFIDF Weighted W2V` encoding of `project_title` feature

```
In [106]:
```

```
# please write all the code with proper documentation, and proper titles for each subsection
# when you plot any graph make sure you use
   # a. Title, that describes your plot, this will be very helpful to the reader
    # b. Legends if needed
    # c. X-axis label
    # d. Y-axis label
import numpy as np
from sklearn.manifold import TSNE
import seaborn as sns
from sklearn import datasets
import pandas as pd
import matplotlib.pyplot as plt
y = project_data['project_is_approved']
y bow=y[0:5000]
#model=TSNE(n components=2,random state=0, perplexity=50, n iter=5000)
model=TSNE(n_components=2,random_state=0)
tsne_data=model.fit_transform(x_tfidfw2v.toarray())
tsne_data=np.vstack((tsne_data.T,y_bow)).T
tsne df=pd.DataFrame(data=tsne data,columns=("Categorical, Numerical features +
```

```
Project_title(TFIDF Weighted W2V)","Approved Projects","label"))
sns.FacetGrid(tsne_df,hue="label",height=6).map(plt.scatter,"Categorical, Numerical features +
Project_title(TFIDF Weighted W2V)","Approved Projects").add_legend();
plt.title('T-SNE with TFIDF Weighted W2V')
plt.show()
```



2.5 Summary

- 1. Number of approved projects is greater than number of non-approved projects.
- 2. Every schoo state state is having approval rates of 80 percent.
- 3. Teacher prefixes having 'Mrs.' having highest number of projects approved.
- 4. Teacher prefix having 'Dr.' having almost zero number of projects approved.
- 5. Teachers with prefixes 'Mrs.,Mr. and Ms.' have more than 80 percent project approval rates.
- 6. Project approval counts for Project Grade Category of Grades PreK-2 is highest.
- 7. Project approval counts for Project Grade Category of Grades 9-12 is lowest.
- 8. Project approval rates for all categories is more than 80 percent.
- 9. Highest success rate category is Literacy Language where is lowest is applied learning Math_Science.
- 10. Success rate of each subcategories is more than 80 percent for project approval.
- 11. Approved projects tends to have more number of words in its Project Title than non approved words. And the same applies for Project Essays as well.
- 12. Project approval count for resource summary which doesnot have any numerical digits is greater than Project approval count for resource summary which have numerical digits.