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	Feature
A unique identifier for the proposed project. Example: p036502	project_id
Title of the project. Examples:	
Art Will Make You Happy!First Grade Fun	project_title
Grade level of students for which the project is targeted. One of the following enumerated values: Grades PreK-2 Grades 3-5 Grades 6-8 Grades 9-12	project_grade_category
One or more (comma-separated) subject categories for the project from the following enumerated list of values: Applied Learning Care & Hunger Health & Sports History & Civics Literacy & Language Math & Science Music & The Arts Special Needs Warmth Examples: Music & The Arts Literacy & Language, Math & Science	project_subject_categories

school_state	State where school is located (<u>Two-letter U.S. postal code</u>). Example: WY
project_subject_subcategories	One or more (comma-separated) subject subcategories for the project. Examples: Literacy Literature & Writing, Social Sciences
project_resource_summary	An explanation of the resources needed for the project. Example: • My students need hands on literacy materials to manage sensory needs!
project_essay_1	First application essay [*]
project_essay_2	Second application essay*
project_essay_3	Third application essay*
project_essay_4	Fourth application essay*
<pre>project_submitted_datetime</pre>	Datetime when project application was submitted. Example: 2016-04-28 12:43:56.245
teacher_id	A unique identifier for the teacher of the proposed project. Example: bdf8baa8fedef6bfeec7ae4ff1c15c56
	Teacher's title. One of the following enumerated values:
teacher_prefix	 nan Dr. Mr. Mrs. Ms. Teacher.
teacher_number_of_previously_posted_projects	Number of project applications previously submitted by the same teacher. Example: 2

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

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

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

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

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

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

Notes on the Essay Data

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

- __project_essay_1:__ "Introduce us to your classroom"
- project_essay_2:__ "Tell us more about your students"
- __project_essay_3:__ "Describe how your students will use the materials you're requesting"
- __project_essay_4:__ "Close by sharing why your project will make a difference"

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

__project_essay_1:__ "Describe your students: What makes

your students special? Specific details about their background, your neighborhood, and your school are all helpful."

 __project_essay_2:__ "About your project: How will these materials make a difference in your students' learning and improve their school lives?"

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

In [2]:

```
%matplotlib inline
import warnings
warnings.filterwarnings("ignore")
import sqlite3
import pandas as pd
import numpy as np
import nltk
import string
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.feature_extraction.text import TfidfTransformer
from sklearn.feature extraction.text import TfidfVectorizer
from sklearn.feature extraction.text import CountVectorizer
from sklearn.metrics import confusion matrix
from sklearn import metrics
from sklearn.metrics import roc_curve, auc
from nltk.stem.porter import PorterStemmer
import re
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 chart_studio.plotly import plot, iplot
import plotly.offline as offline
import plotly.graph_objs as go
offline.init_notebook_mode()
from collections import Counter
```

1.1 Reading Data

```
In [3]:
project_data = pd.read_csv('train_data.csv')
resource data = pd.read_csv('resources.csv')
                                                        In [4]:
print("Number of data points in train data", project_data.sha
pe)
print('-'*50)
print("The attributes of data :", project_data.columns.values
Number of data points in train data (109248, 1
The attributes of data : ['Unnamed: 0' 'id' 't
eacher_id' 'teacher_prefix' 'school_state'
 'project_submitted_datetime' 'project_grade_c
ategory'
 'project_subject_categories' 'project_subject
_subcategories'
 'project_title' 'project_essay_1' 'project_es
say_2' 'project_essay_3'
 'project_essay_4' 'project_resource_summary'
 'teacher_number_of_previously_posted_projects
' 'project_is_approved']
                                                        In [5]:
print("Number of data points in train data", resource_data.sh
ape)
```

```
print(resource_data.columns.values)
resource_data.head(2)
Number of data points in train data (1541272,
4)
['id' 'description' 'quantity' 'price']
                                                          Out[5]:
        id
                        description quantity
                                            price
0
             LC652 - Lakeshore Double-
   p233245
                                        1 149.00
              Space Mobile Drying Rack
               Bouncy Bands for Desks
   p069063
                                        3
                                            14.95
                  (Blue support pipes)
                                                          In [6]:
np.unique(project_data["project_grade_category"].values)
                                                          Out[6]:
array(['Grades 3-5', 'Grades 6-8', 'Grades 9-1
2', 'Grades PreK-2'], dtype=object)
                                                          In [7]:
# We need to get rid of The spaces between the text and the h
yphens because they're special characters.
#Rmoving multiple characters from a string in Python
#https://stackoverflow.com/questions/3411771/multiple-charact
er-replace-with-python
project_grade_category = []
for i in range(len(project_data)):
    a = project_data["project_grade_category"][i].replace(" "
, "_").replace("-", "_")
    project_grade_category.append(a)
```

```
In [8]:
```

```
project_data.drop(['project_grade_category'], axis = 1, inpla
ce = True)
project_data["project_grade_category"] = project_grade_catego
ry
print("After removing the special characters ,Column values:
    ",np.unique(project_data["project_grade_category"].values))
```

```
After removing the special characters ,Column values: ['Grades_3_5' 'Grades_6_8' 'Grades_9 _12' 'Grades_PreK_2']
```

1.2 Data Analysis

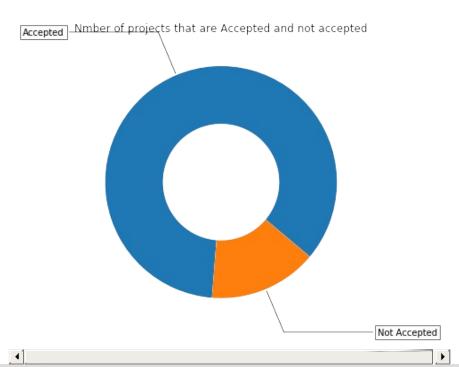
```
In [9]:
```

```
# https://matplotlib.org/gallery/pie_and_polar_charts/pie_and
_donut_labels.html#sphx-glr-gallery-pie-and-polar-charts-pie-
and-donut-labels-pv
y_value_counts = project_data['project_is_approved'].value_co
unts()
print("Number of projects than 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 than are not approved for funding "
, y_value_counts[0], ", (", (y_value_counts[0]/(y_value_count
s[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), star
tangle=-40)
bbox_props = dict(boxstyle="square,pad=0.3", fc="w", ec="k",
1w = 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.sig
n(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, **k
w)

ax.set_title("Nmber of projects that are Accepted and not accepted")
plt.show()
```

Number of projects thar are approved for funding 92706 , (84.8583040422 %)
Number of projects thar are not approved for funding 16542 , (15.1416959578 %)



From the pie chart, we can observe that out of 109k project proposals:

- 92706 proposals, nearly 85% of the total project proposals were approved for funding
- 16542 proposals , about 15% of the total project proposals were rejected

1.2.1 Univariate Analysis: School State

4

In [19]:

```
'''Reference == https://datascience.stackexchange.com/questio
ns/9616/how-to-create-us-state-heatmap/9620#9620'''
# Pandas dataframe groupby count, mean: https://stackoverflow
.com/a/19385591/4084039
temp = pd.DataFrame(project_data.groupby("school_state")["pro
ject_is_approved"].apply(np.mean)).reset_index()
# if you have data which contain only 0 and 1, then the mean
= percentage (think about it)
temp.columns = ['state_code', 'num_proposals']
scl = [[0.0, 'rgb(242,240,247)'],[0.2, 'rgb(218,218,235)'],[0
.4, 'rgb(188,189,220)'],\
            [0.6, 'rgb(158,154,200)'], [0.8, 'rgb(117,107,177)
'],[1.0, 'rgb(84,39,143)']]
data = [ dict(
        type='choropleth',
        colorscale = scl,
        autocolorscale = False,
        locations = temp['state_code'],
```

```
z = temp['num_proposals'].astype(float),
        locationmode = 'USA-states',
        text = temp['state_code'],
        marker = dict(line = dict (color = 'rgb(255, 255, 255)'
, width = 2)),
        colorbar = dict(title = "% of pro")
    ) ]
layout = dict(
        title = 'Project Proposals % of Acceptance Rate by US
 States',
        geo = dict(
            scope='usa',
            projection=dict( type='albers usa' ),
            showlakes = True,
            lakecolor = 'rgb(255, 255, 255)',
        ),
    )
fig = go.Figure(data=data, layout=layout)
offline.iplot(fig, filename='d3-cloropleth-map')
```

In [12]:

```
# https://www.csi.cuny.edu/sites/default/files/pdf/administra
tion/ops/2letterstabbrev.pdf
temp.sort_values(by=['num_proposals'], inplace=True)
print("States with lowest % approvals")
print(temp.head(5))
print('='*50)
print("States with highest % approvals")
print(temp.tail(5))
States with lowest % approvals
   state_code num_proposals
46
                   0.800000
          VT
7
                   0.802326
          DC
43
          TΧ
                   0.813142
26
          MΤ
                   0.816327
18
          LA
                   0.831245
_____
States with highest % approvals
   state_code
              num_proposals
30
                   0.873563
          NH
35
          ОН
                   0.875152
47
                   0.876178
          WA
28
                   0.888112
          ND
                   0.897959
8
          DE
```

Summary:

• Among all states, it's evident that Vermont has the lowest acceptance rate of 80%

In [13]:

```
#stacked bar plots matplotlib: https://matplotlib.org/gallery
/lines_bars_and_markers/bar_stacked.html

def stack_plot(data, xtick, col2='project_is_approved', col3=
'total'):
    ind = np.arange(data.shape[0])

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

plt.ylabel('Projects')
    plt.title('Number of projects aproved vs rejected')
    plt.xticks(ind, list(data[xtick].values))

plt.legend((p1[0], p2[0]), ('total', 'accepted'))
    plt.show()
```

In [14]:

```
def univariate_barplots(data, col1, col2='project_is_approved
', top=False):
    # Count number of zeros in dataframe python: https://stac
koverflow.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.co
m/a/19385591/4084039
    temp['total'] = pd.DataFrame(project_data.groupby(col1)[col1])
```

```
ol2].agg({'total':'count'})).reset_index()['total']
    temp['Avg'] = pd.DataFrame(project_data.groupby(col1)[col
2].agg({'Avg':'mean'})).reset_index()['Avg']

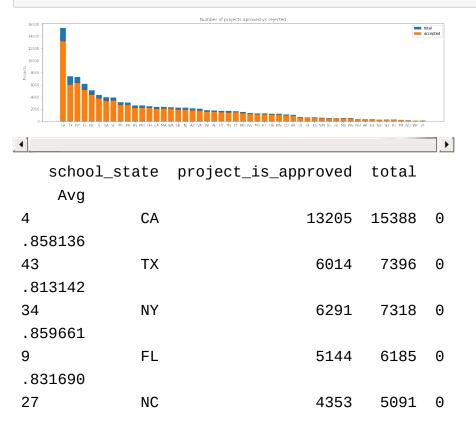
    temp.sort_values(by=['total'],inplace=True, ascending=Fal
se)

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 [15]:

```
univariate_barplots(project_data, 'school_state', 'project_is
_approved', False)
```



.855038									
====									
school_st	ate	<pre>project_is_approved</pre>	total						
Avg									
39	RI	243	285	0					
.852632									
26	MT	200	245	0					
.816327									
28	ND	127	143	0					
.888112									
50	WY	82	98	0					
.836735									
46	VT	64	80	0					
.800000									

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

However, the following observations can be made:

- California has the highest number of project proposals compared to other states
- Delaware has the highest acceptance rate of 89%
- Vermont has the least acceptance rate of 80%

In [18]:

#NaN values in techer prefix will create a problem while encoding, so we replace NaN values with the mode of that particula

```
r column
mode_of_teacher_prefix = project_data['teacher_prefix'].value
_counts().index[0]
project_data['teacher_prefix'] = project_data['teacher_prefix
'].fillna(mode_of_teacher_prefix)
project_data['teacher_prefix']
                                                       Out[18]:
0
             Mrs.
1
              Mr.
2
              Ms.
3
             Mrs.
4
             Mrs.
5
             Mrs.
6
             Mrs.
```

7

8

9

10

11

12

13

14

15

16

17

18

19

20

21

22

23

24

25

Ms.

Mrs.

Mrs.

Ms.

Mrs.

Mrs.

Ms.

Ms.

Ms.

Mrs.

Mrs.

Mrs.

Ms.

Mr.

Mrs.

Mrs.

Ms.

Mrs.

Ms.

```
Ms.
26
27
           Teacher
28
              Mrs.
29
              Mrs.
109218
              Mrs.
109219
           Teacher
              Mrs.
109220
109221
          Teacher
109222
               Ms.
109223
               Ms.
109224
               Ms.
109225
              Mrs.
109226
               Ms.
109227
              Mrs.
109228
              Mrs.
109229
              Mrs.
109230
               Ms.
109231
              Mrs.
109232
              Mrs.
109233
               Ms.
109234
               Ms.
109235
              Mrs.
109236
              Mrs.
109237
              Mrs.
109238
              Mrs.
109239
              Mrs.
109240
              Mrs.
109241
              Mrs.
109242
              Mrs.
109243
               Mr.
109244
               Ms.
109245
              Mrs.
109246
              Mrs.
109247
               Ms.
Name: teacher_prefix, Length: 109248, dtype: o
bject
```

1.2.2 Univariate Analysis: teacher_prefix

In [20]:

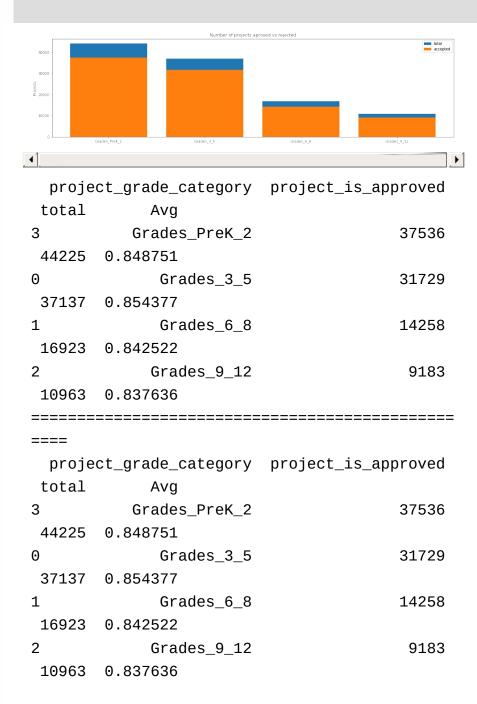
```
univariate_barplots(project_data, 'teacher_prefix', 'project_
is_approved' , top=False)
                                         total accepted
  teacher_prefix project_is_approved
                                      total
    Avg
2
                               49000
                                      57272
           Mrs.
0.855566
3
            Ms.
                               32860
                                      38955
0.843537
                                8960
                                      10648
            Mr.
0.841473
        Teacher
                                1877
                                       2360
0.795339
0
            Dr.
                                   9
                                         13
0.692308
_____
  teacher_prefix project_is_approved
                                     total
    Avg
2
                               49000
                                      57272
           Mrs.
0.855566
            Ms.
                               32860
                                      38955
0.843537
1
            Mr.
                                8960
                                      10648
0.841473
```

```
4 Teacher 1877 2360 0.795339 0 Dr. 9 13 0.692308
```

Summary:

- Female teachers have more number of projects proposed and also have a higher acceptance rate compared to the male teachers
- Among the female teachers, the ones with prefixes (Mrs.) that is married female teachers have more number of project proposals
- Teachers with Dr. prefix have the least number of project proposals and also the least acceptance rate compared to project proposals coming from teachers with other prefixes
- 1.2.3 Univariate Analysis: project_grade_category

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



1.2.4 Univariate Analysis: project_subject_categories

In [22]:

catogories = list(project_data['project_subject_categories'].

```
values)
# remove special characters from list of strings python: http
s://stackoverflow.com/a/47301924/4084039
# https://www.geeksforgeeks.org/removing-stop-words-nltk-pyth
on/
# https://stackoverflow.com/questions/23669024/how-to-strip-a
-specific-word-from-a-string
# https://stackoverflow.com/questions/8270092/remove-all-whit
espace-in-a-string-in-python
cat_list = []
for i in catogories:
    temp = ""
    # consider we have text like this "Math & Science, Warmth
, Care & Hunger"
    for j in i.split(','): # it will split it in three parts
["Math & Science", "Warmth", "Care & Hunger"]
        if 'The' in j.split(): # this will split each of the
catogory based on space "Math & Science"=> "Math", "&", "Scien
ce"
            j=j.replace('The','') # if we have the words "The
" we are going to replace it with ''(i.e removing 'The')
        j = j.replace(' ','') # we are placeing all the ' '(s
pace) with ''(empty) ex:"Math & Science"=>"Math&Science"
        temp+=j.strip()+" " #" abc ".strip() will return "abc
", remove the trailing spaces
        temp = temp.replace('&','_') # we are replacing the &
value into
    cat_list.append(temp.strip())
                                                      In [23]:
```

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

Out[23]:

```
Unnamed:
                   id
                                          teacher_id teacher_prefix school_
 0
      160221 p253737
                       c90749f5d961ff158d4b4d1e7dc665fc
                                                            Mrs.
 1
       140945 p258326 897464ce9ddc600bced1151f324dd63a
                                                             Mr.
4
                                                ▶
                                                         In [24]:
 univariate_barplots(project_data, 'clean_categories', 'projec
 t_is_approved', top=20)
                     Number of projects aproved vs rejected
                                             total accepted
, 4 ].
                                                |
                   clean_categories
                                      project_is
_approved total
                        Avg
24
                  Literacy_Language
     20520
                   0.867470
            23655
32
                       Math Science
     13991 17072 0.819529
28 Literacy_Language Math_Science
     12725 14636 0.869432
8
                      Health_Sports
      8640 10177 0.848973
40
                         Music_Arts
      4429
             5180 0.855019
_____
====
```

```
clean_categories project_
is_approved total
                        Avg
19
    History_Civics Literacy_Language
       1271
              1421 0.894441
14
          Health_Sports SpecialNeeds
       1215
              1391 0.873472
50
                  Warmth Care_Hunger
       1212
              1309 0.925898
33
        Math_Science AppliedLearning
       1019
              1220 0.835246
4
        AppliedLearning Math_Science
              1052 0.812738
        855
```

In [25]:

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

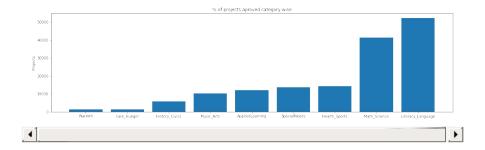
In [26]:

```
# dict sort by value python: https://stackoverflow.com/a/6132
18/4084039
cat_dict = dict(my_counter)
sorted_cat_dict = dict(sorted(cat_dict.items(), key=lambda kv
: kv[1]))

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

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

plt.show()



In [27]:

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

Warmth 1388 Care_Hunger 1388 History_Civics 5914 Music_Arts 10293 AppliedLearning 12135 SpecialNeeds 13642 Health_Sports 14223 Math_Science 41421 Literacy_Language 52239

SUMMARY:

- Projects belonging to the Literacy and Language categories have the highest number of projects proposals
- Warmth,care and Hunger have the least number of project proposals

1.2.5 Univariate Analysis: project_subject_subcategories

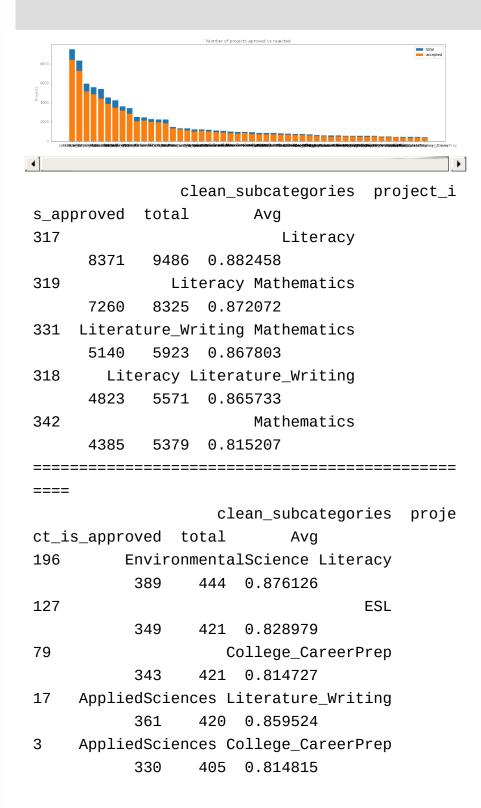
In [28]:

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

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

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

```
for j in i.split(','): # it will split it in three parts
["Math & Science", "Warmth", "Care & Hunger"]
         if 'The' in j.split(): # this will split each of the
 catogory based on space "Math & Science"=> "Math", "&", "Scien
ce"
             j=j.replace('The','') # if we have the words "The
 " we are going to replace it with ''(i.e removing 'The')
         j = j.replace(' ','') # we are placeing all the ' '(s
pace) with ''(empty) ex:"Math & Science"=>"Math&Science"
         temp +=j.strip()+" "#" abc ".strip() will return "abc
 ", remove the trailing spaces
         temp = temp.replace('&','_')
    sub_cat_list.append(temp.strip())
                                                        In [29]:
 project_data['clean_subcategories'] = sub_cat_list
 project_data.drop(['project_subject_subcategories'], axis=1,
inplace=True)
 project_data.head(2)
                                                        Out[29]:
    Unnamed:
                                         teacher_id teacher_prefix school_
                  id
 0
      160221 p253737
                      c90749f5d961ff158d4b4d1e7dc665fc
                                                           Mrs.
 1
      140945 p258326 897464ce9ddc600bced1151f324dd63a
                                                            Mr.
4
                                                •
                                                        In [30]:
univariate_barplots(project_data, 'clean_subcategories', 'pro
ject_is_approved', top=50)
```



SUMMARY:

- From the bar plot, it can be observed that the number of project proposals and acceptance rate from each category is not uniform and there is a lot variability based on category
- Projects belonging to Literacy and Language categories have the highest acceptance rate of about 87-88% compared to other categories

In [31]:

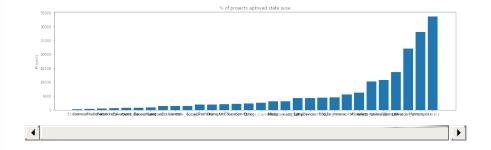
```
# count of all the words in corpus python: https://stackoverf
low.com/a/22898595/4084039
from collections import Counter
my_counter = Counter()
for word in project_data['clean_subcategories'].values:
    my_counter.update(word.split())
```

In [32]:

```
# dict sort by value python: https://stackoverflow.com/a/6132
18/4084039
sub_cat_dict = dict(my_counter)
sorted_sub_cat_dict = dict(sorted(sub_cat_dict.items(), key=1
ambda kv: kv[1]))

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

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



In [33]:

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

Economics 269 CommunityService 441 FinancialLiteracy 568 ParentInvolvement 677 Extracurricular 810 Civics_Government 815 ForeignLanguages 890 NutritionEducation 1355 Warmth 1388 Care_Hunger 1388 SocialSciences 1920 PerformingArts 1961 CharacterEducation 2065 TeamSports 2192 0ther 2372 College_CareerPrep 2568 Music 3145 History_Geography 3171 Health_LifeScience 4235 EarlyDevelopment 4254 **ESL** 4367 Gym_Fitness 4509 EnvironmentalScience: 5591 VisualArts 6278 Health_Wellness 10234 AppliedSciences 10816

SpecialNeeds : 13642 Literature_Writing : 22179 Mathematics : 28074 Literacy : 33700

SUMMARY:

- Literacy has the highest number of project proposals
- Economics has the lowest number of projects proposals with 269 projects only

1.2.6 Univariate Analysis: Text features (Title)

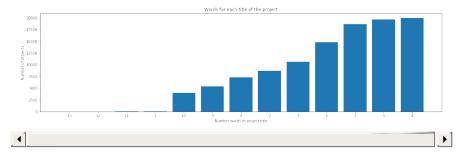
In [35]:

```
#How to calculate number of words in a string in DataFrame: h
ttps://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))
p1 = plt.bar(ind, list(word_dict.values()))

plt.ylabel('Number of projects')
plt.xlabel('Number words in project title')
```

```
plt.title('Words for each title of the project')
plt.xticks(ind, list(word_dict.keys()))
plt.show()
```



SUMMARY:

- The word count in the title for Majority of the projects is between 3-5 words
- Very few project titles have more than 10 words

```
In [36]:
```

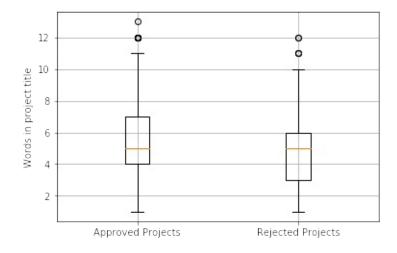
```
approved_title_word_count = project_data[project_data['project_is_approved']==1]['project_title'].str.split().apply(len)
approved_title_word_count = approved_title_word_count.values

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

In [37]:

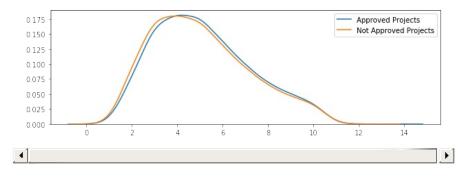
```
# https://glowingpython.blogspot.com/2012/09/boxplot-with-mat
plotlib.html
plt.boxplot([approved_title_word_count, rejected_title_word_c
ount])
plt.xticks([1,2],('Approved Projects','Rejected Projects'))
```

```
plt.ylabel('Words in project title')
plt.grid()
plt.show()
```



In [38]:

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



SUMMARY:

- From the kde plot and box plots, it's evident that there is some loose correlation between the word count in the project title and the approval status of the project
- Projects with more number of words in the title have a slightly better acceptance rate than projects with less number of words

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

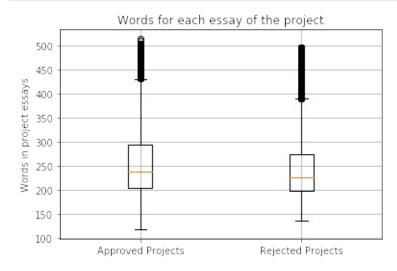
```
In [39]:
```

In [40]:

```
approved_word_count = project_data[project_data['project_is_a
pproved']==1]['essay'].str.split().apply(len)
approved_word_count = approved_word_count.values

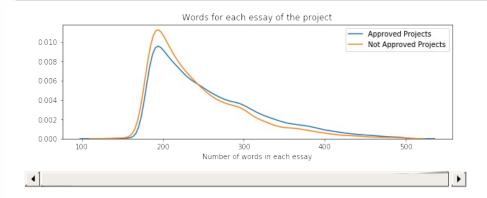
rejected_word_count = project_data[project_data['project_is_a
pproved']==0]['essay'].str.split().apply(len)
rejected_word_count = rejected_word_count.values
```

```
# https://glowingpython.blogspot.com/2012/09/boxplot-with-mat
plotlib.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 [42]:

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



SUMMARY:

From the kde plot or the pdf,it's observable that the approved projects have slighly more number of words in their project essays as compared to rejected projects

1.2.8 Univariate Analysis: Cost per project

In [43]:

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

Out[43]:

_		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 [44]:

```
# 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[44]:

	id	price	quantity
0	p000001	459.56	7
1	p000002	515.89	21

In [45]:

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

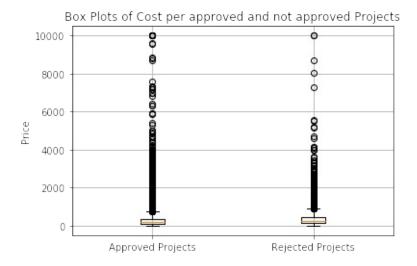
In [46]:

```
approved_price = project_data[project_data['project_is_approv
ed']==1]['price'].values

rejected_price = project_data[project_data['project_is_approv
ed']==0]['price'].values
```

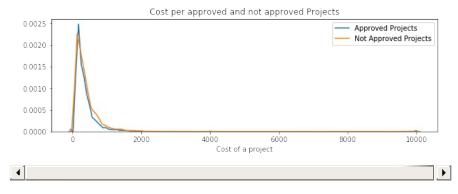
In [47]:

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



In [48]:

```
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()
```



SUMMARY:

- The pdf's of approved projects and rejected projects based on the feature price seem to almost overlap
- However,in general if the price of the project proposal is more or if it's expensive,then it is likely to get rejected

```
4
                                                    b
                                              In [49]:
# 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 Appr
oved 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 Approve
d Projects |
+----+---+-----
                  0.66
                                      1.
     0 |
97
          5
                 13.59
                                      41
. 9
                  33.88
     10
          73.
67
```

I	15	I	58.0	I	99.
109					
	20		77.38	1	118
.56		l			
	25	I	99.95	I	140.
892		l			
	30	I	116.68	I	162
. 23					
	35		137.232	l	184.
014					
	40	l	157.0	l	208.
632		1			
	45		178.265	l	235.
106		1		_	
	50		198.99	l	263.
145		l .		_	
	55	.	223.99	I	292
.61		l .			
	60	.	255.63	l	325.
144	0.5	1	005 440		000
	65	. 1	285.412	I	362
.39	70	1 ,	224 225		200
1	70	. 1	321.225	I	399
.99	7.5	1	200 075		4.40
045	75	. 1	366.075	I	449.
945	90	1	411 67	1	E10
। 282	80	1	411.67	ı	519.
	85	1	479.0	1	618.
ı 276	03	1	479.0	ı	010.
	90	1	593.11	1	739.
356	30	1	333.11	ı	755.
1	95	· I	801.598	ı	992.
1 486	33	ı	001.000	ı	3321
	100	ı I	9999.0	ı	999
9.0	_55	'		1	
+		' +		+	

----+

SUMMARY:

- The approved projects tend to have lower cost when compared to the projects that have not been approved. This is evident from the percentile values. The 50th percentile Cost value for an approved project is 199 dollars whereas the cost for rejected projects is 263 dollars.
- In general, if the price of any project exceeds more than 10,000 dollars, then it will be rejected.

1.2.9 Univariate Analysis: teacher_number_of_previously_posted_projects

```
In [50]:
univariate_barplots(project_data, 'teacher_number_of_previous
ly_posted_projects', 'project_is_approved', top=30)

**Transfer of project approved vs rejected**

teacher_number_of_previously_posted_project
s project_is_approved total \
0
24652 30014
```

```
1
1
               13329
                     16058
2
2
                8705
                     10350
3
3
                5997
                      7110
4
4
                4452
                      5266
       Avg
0 0.821350
1 0.830054
2 0.841063
3 0.843460
4 0.845423
_____
====
   teacher_number_of_previously_posted_projec
   project_is_approved total \
ts
24
24
                  405
                        449
26
26
                  378
                        445
27
27
                  352
                        394
29
29
                  336
                        370
28
28
                  313
                        352
        Avg
24 0.902004
26 0.849438
27 0.893401
29 0.908108
   0.889205
28
```

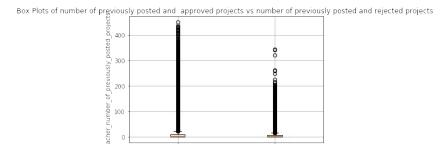
```
In [51]:
```

```
approved_previously_posted_projects = project_data[project_da
ta['project_is_approved']==1]['teacher_number_of_previously_p
osted_projects'].values

rejected_previously_posted_projects = project_data[project_da
ta['project_is_approved']==0]['teacher_number_of_previously_p
osted_projects'].values
```

In [52]:

```
plt.boxplot([approved_previously_posted_projects, rejected_pr
eviously_posted_projects])
plt.title('Box Plots of number of previously posted and appr
oved projects vs number of previously posted and rejected pro
jects ')
plt.xticks([1,2],('Approved Projects','Rejected Projects'))
plt.ylabel('teacher_number_of_previously_posted_projects')
plt.grid()
plt.show()
```



√]

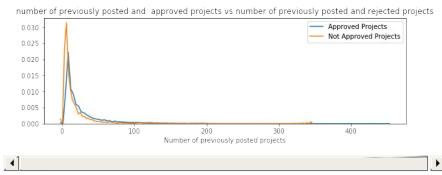
Approved Projects

In [53]:

```
plt.figure(figsize=(10,3))
sns.distplot(approved_previously_posted_projects, hist=False,
  label="Approved Projects")
sns.distplot(rejected_previously_posted_projects, hist=False,
  label="Not Approved Projects")
plt.title('number of previously posted and approved projects
```

Rejected Projects

```
vs number of previously posted and rejected projects')
plt.xlabel('Number of previously posted projects')
plt.legend()
plt.show()
```



SUMMARY:

- We can observe that the approval status is not affected much by how many projects were submitted in the past. From the barplots, it is evident that majority of the approved project proposals have no previous submissions.
- But among the accepted project proposals with more number of previous submissions, the rate of approval seems to be Higher if the teacher has proposed atleast 19 different projects.

1.2.10 Univariate Analysis: project_resource_summary

In [54]:

```
#Some summaries contain alphanumeric characters but we don't
want them to be added in the new column.
#Hence, we only consider numeric digits present in each summar
y and store it in a dictionary
summaries = project_data['project_resource_summary']
digits_in_each_summary = {}
for x in tqdm(range(len(project_data))):
            for word in summaries[x].split():
                        if word.isdigit() :
                                    digits_in_each_summary[x] = int(word) #count
ing number of digits in each summary
100%| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1
06391.59it/s]
                                                                                                                                                                  In [55]:
#saving it into a dictionary. If a digit is not present in a
summary, it's value is 0
numeric_digits = {}
for i in tqdm(range(len(summaries))) :
            if i in digits in each summary.keys() :
                        numeric_digits[i] = digits_in_each_summary[i]
            else:
                        numeric_digits[i] = 0
100%| 100%| 100248/109248 [00:00<00:00, 2
356743.71it/s]
                                                                                                                                                                  In [56]:
#replacing actual values of digit count with 1 because we are
  only interested to know if a digit is present or not
for i in tqdm(range(len(numeric_digits))):
```

```
if numeric_digits.get(i)>0:
    numeric_digits.update({i:1})

100%| 100%| 1009248/109248 [00:00<00:00, 2
237704.98it/s]</pre>
```

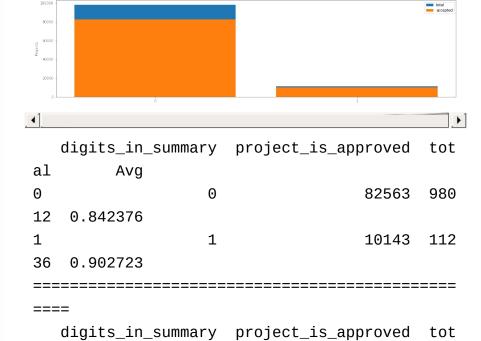
In [57]:

```
#how to convert a dictionary to list: https://www.quora.com/H
ow-do-I-convert-a-dictionary-to-a-list-in-Python
digits_in_summary=[i for i in numeric_digits.values()]
project_data['digits_in_summary'] = digits_in_summary
```

In [58]:

#Since we have our new column,let's check if this new feature or column is useful to predict the acceptance rate of a project proposal

univariate_barplots(project_data, 'digits_in_summary', 'proje ct_is_approved')



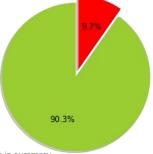
Number of projects aproved vs rejected

```
al Avg
0 0 82563 980
12 0.842376
1 1 10143 112
36 0.902723
```

In [59]:

```
accepted_digits = project_data.groupby(['digits_in_summary','
project_is_approved']).size()[1][1]
rejected_digits= project_data.groupby(['digits_in_summary','p
roject_is_approved']).size()[1][0]
# How to plot pie chart :https://stackoverflow.com/guestions/
39969089/how-to-create-pie-chart
labels = 'accepted projects containing digits in summary', 'r
ejected projects containing digits'
sizes = [accepted_digits, rejected_digits]
colors = ['yellowgreen', 'red']
explode = (0, 0.1)
plt.pie(sizes, explode=explode, labels=labels, colors=colors,
        autopct='%1.1f%%', shadow=True, startangle=90)
plt.axis('equal')
plt.title('Pie chart showing the ratio of accepted projects c
ontaining digits in project summary')
plt.show()
```

Pie chart showing the ratio of accepted projects containing digits in project summary rejected projects containing digits



accepted projects containing digits in summary

4

lacksquare

SUMMARY:

- It is not mandatory for project proposals to have numeric digits in their project summary because from the stats it is clear that a majority of approved projects do not have numeric digits in their project summary.
- However, from the pie chart, we can observe that among the project proposals with numeric digits in their summary, 90% of them got accepted and about 10% got rejected

1.3 Text preprocessing

1.3.1 Essay Text

```
In [52]:
project_data.head(2)
                                                          Out[52]:
   Unnamed:
                  id
                                          teacher_id
0
      160221 p253737 c90749f5d961ff158d4b4d1e7dc665fc
1
      140945 p258326 897464ce9ddc600bced1151f324dd63a
2 rows × 21 columns
                                                          In [60]:
# printing some random essays.
print(project_data['essay'].values[0])
print("="*50)
print(project_data['essay'].values[150])
print("="*50)
```

```
print(project_data['essay'].values[1000])
print("="*50)
print(project_data['essay'].values[20000])
print("="*50)
print(project_data['essay'].values[99999])
print("="*50)
```

My students are English learners that are work ing on English as their second or third langua ges. We are a melting pot of refugees, immigra nts, and native-born Americans bringing the gi ft of language to our school. \r\n\r\n We have over 24 languages represented in our English Learner program with students at every level o f mastery. We also have over 40 countries rep resented with the families within our school. Each student brings a wealth of knowledge and experiences to us that open our eyes to new c ultures, beliefs, and respect.\"The limits of your language are the limits of your world.\"-Ludwig Wittgenstein Our English learner's hav e a strong support system at home that begs fo r more resources. Many times our parents are learning to read and speak English along side of their children. Sometimes this creates bar riers for parents to be able to help their chi ld learn phonetics, letter recognition, and ot her reading skills.\r\n\r\nBy providing these dvd's and players, students are able to contin ue their mastery of the English language even if no one at home is able to assist. All fami lies with students within the Level 1 proficie ncy status, will be a offered to be a part of this program. These educational videos will b e specially chosen by the English Learner Teac her and will be sent home regularly to watch. The videos are to help the child develop earl y reading skills.\r\n\r\nParents that do not h ave access to a dvd player will have the oppor tunity to check out a dvd player to use for th e year. The plan is to use these videos and e ducational dvd's for the years to come for oth er EL students.\r\nnannan

====

The 51 fifth grade students that will cycle th rough my classroom this year all love learning , at least most of the time. At our school, 97 .3% of the students receive free or reduced pr ice lunch. Of the 560 students, 97.3% are mino rity students. \r\nThe school has a vibrant co mmunity that loves to get together and celebra te. Around Halloween there is a whole school p arade to show off the beautiful costumes that students wear. On Cinco de Mayo we put on a bi g festival with crafts made by the students, d ances, and games. At the end of the year the s chool hosts a carnival to celebrate the hard w ork put in during the school year, with a dunk tank being the most popular activity. My stude nts will use these five brightly colored Hokki stools in place of regular, stationary, 4-leg ged chairs. As I will only have a total of ten in the classroom and not enough for each stud ent to have an individual one, they will be us ed in a variety of ways. During independent re ading time they will be used as special chairs students will each use on occasion. I will ut ilize them in place of chairs at my small grou p tables during math and reading times. The re st of the day they will be used by the student s who need the highest amount of movement in t heir life in order to stay focused on school. r\n\r\nWhenever asked what the classroom is mi

ssing, my students always say more Hokki Stool s. They can't get their fill of the 5 stools w e already have. When the students are sitting in group with me on the Hokki Stools, they are always moving, but at the same time doing the ir work. Anytime the students get to pick wher e they can sit, the Hokki Stools are the first to be taken. There are always students who he ad over to the kidney table to get one of the stools who are disappointed as there are not e nough of them. \r\n\r\nWe ask a lot of student s to sit for 7 hours a day. The Hokki stools w ill be a compromise that allow my students to do desk work and move at the same time. These stools will help students to meet their 60 min utes a day of movement by allowing them to act ivate their core muscles for balance while the y sit. For many of my students, these chairs w ill take away the barrier that exists in schoo ls 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 the emed 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 and reduced-price lunch to qualify. Our school is an \"open classroom\" concept, which is very unique as there are no walls separating the classrooms. These 9 and 10 year old students are very eager learners; they ar

e like sponges, absorbing all the information and experiences and keep on wanting more. With these resources such as the comfy red throw pi llows and the whimsical nautical hanging decor and the blue fish nets, I will be able to hel p create the mood in our classroom setting to be one of a themed nautical environment. Creat ing a classroom environment is very important in the success in each and every child's educa tion. The nautical photo props will be used wi th each child as they step foot into our class room for the first time on Meet the Teacher ev ening. I'll take pictures of each child with t hem, have them developed, and then hung in our classroom ready for their first day of 4th gr This kind gesture will set the tone befo re even the first day of school! The nautical thank you cards will be used throughout the ye ar by the students as they create thank you ca rds to their team groups.\r\n\r\nYour generous donations will help me to help make our class room a fun, inviting, learning environment fro m day one.\r\n\r\nIt costs lost of money out o f my own pocket on resources to get our classr oom ready. Please consider helping with this p roject to make our new school year a very succ essful one. Thank you!nannan

====

My kindergarten students have varied disabilit ies ranging from speech and language delays, c ognitive delays, gross/fine motor delays, to a utism. They are eager beavers and always striv e to work their hardest working past their lim itations. \r\n\r\nThe materials we have are th e ones I seek out for my students. I teach in a Title I school where most of the students re

ceive free or reduced price lunch. Despite th eir disabilities and limitations, my students love coming to school and come eager to learn and explore. Have you ever felt like you had an ts 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 chai rs are the answer and I love then because they develop their core, which enhances gross moto r and in Turn fine motor skills. \r\nThey also want to learn through games, my kids don't wa nt to sit and do worksheets. They want to lear n to count by jumping and playing. Physical en gagement is the key to our success. The number toss and color and shape mats can make that h appen. My students will forget they are doing work and just have the fun a 6 year old deserv es.nannan

====

The mediocre teacher tells. The good teacher e xplains. The superior teacher demonstrates. Th e great teacher inspires. -William A. Ward\r\n \r\nMy school has 803 students which is makeup is 97.6% African-American, making up the larg est segment of the student body. A typical sch ool in Dallas is made up of 23.2% African-Amer ican students. Most of the students are on fre e or reduced lunch. We aren't receiving doctor s, lawyers, or engineers children from rich ba ckgrounds or neighborhoods. As an educator I a m inspiring minds of young children and we foc us not only on academics but one smart, effect ive, efficient, and disciplined students with good character. In our classroom we can utilize the Bluetooth for swift transitions during cl

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

====

In [61]:

```
# 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 [62]:
```

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

My kindergarten students have varied disabilit ies ranging from speech and language delays, c ognitive delays, gross/fine motor delays, to a utism. They are eager beavers and always striv e to work their hardest working past their lim itations. \r\n\r\nThe materials we have are th e ones I seek out for my students. I teach in a Title I school where most of the students re ceive free or reduced price lunch. Despite th eir disabilities and limitations, my students love coming to school and come eager to learn and explore. Have you ever felt like you had an ts 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 chai rs are the answer and I love then because they develop their core, which enhances gross moto r and in Turn fine motor skills. \r\nThey also want to learn through games, my kids do not w ant to sit and do worksheets. They want to lea rn to count by jumping and playing. Physical e ngagement is the key to our success. The numbe r 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 deser ves.nannan

====

```
# \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 disabilit ies ranging from speech and language delays, c ognitive delays, gross/fine motor delays, to a utism. They are eager beavers and always striv e to work their hardest working past their lim The materials we have are the on itations. es I seek out for my students. I teach in a Ti tle I school where most of the students receiv e 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 i n 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 mov e as they learn or so they say. Wobble chairs a re the answer and I love then because they dev elop their core, which enhances gross motor an d 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 engagem ent 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.na nnan

```
#remove spacial character: https://stackoverflow.com/a/584354
7/4084039
sent = re.sub('[^A-Za-z0-9]+', ' ', sent)
print(sent)
```

My kindergarten students have varied disabilit ies ranging from speech and language delays co gnitive delays gross fine motor delays to auti sm They are eager beavers and always strive to work their hardest working past their limitat ions The materials we have are the ones I seek out for my students I teach in a Title I scho ol where most of the students receive free or reduced price lunch Despite their disabilities and limitations my students love coming to sc hool 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 a nd I love then because they develop their core which enhances gross motor and in Turn fine m otor skills They also want to learn through ga mes my kids do not want to sit and do workshee ts 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 ma ts can make that happen My students will forge t they are doing work and just have the fun a 6 year old deserves nannan

In [65]:

```
# 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', 'b
etween', 'into', 'through', 'during', 'before', 'after',\
            'above', 'below', 'to', 'from', 'up', 'down', 'in
', 'out', 'on', 'off', 'over', 'under', 'again', 'further',\
            'then', 'once', 'here', 'there', 'when', 'where',
'why', 'how', 'all', 'any', 'both', 'each', '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', "co
uldn't", 'didn', "didn't", 'doesn', "doesn'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"]
```

In [66]:

```
# Combining all the above statemennts
from tqdm import tqdm
preprocessed_essays = []
# tqdm is for printing the status bar
```

In [67]:

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

Out[67]:

'my kindergarten students varied disabilities ranging speech language delays cognitive delay s gross fine motor delays autism they eager be avers always strive work hardest working past limitations the materials ones i seek students i teach title i school students receive free reduced price lunch despite disabilities limit ations students love coming school come eager learn explore have ever felt like ants pants n eeded groove move meeting this kids feel time the want able move learn say wobble chairs ans wer i love develop core enhances gross motor t urn fine motor skills they also want learn gam es kids not want sit worksheets they want lear n count jumping playing physical engagement ke y success the number toss color shape mats mak e happen my students forget work fun 6 year ol d deserves nannan'

1.3.2 Project title Text

In [68]:

1. 4 Preparing data for models

```
In [69]:
project_data.columns
                                                       Out[69]:
Index(['Unnamed: 0', 'id', 'teacher_id', 'teac
her_prefix', 'school_state',
       'project_submitted_datetime', 'project_
title', 'project_essay_1',
       'project_essay_2', 'project_essay_3', '
project_essay_4',
       'project_resource_summary',
       'teacher_number_of_previously_posted_pr
ojects', 'project_is_approved',
       'project_grade_category', 'clean_catego
ries', 'clean_subcategories',
       'essay', 'price', 'quantity', 'digits_i
n_summary'],
      dtype='object')
we are going to consider
      - school_state : categorical data
      - clean_categories : categorical data
      - clean_subcategories : categorical data
      - project_grade_category : categorical data
      - teacher_prefix : categorical data
      - project_title : text data
      - text : text data
      - project_resource_summary: text data
      - quantity : numerical
      - teacher_number_of_previously_posted_projects : nu
```

```
merical
- price : numerical
```

1.4.1 Vectorizing Categorical data

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

In [70]:

```
# we use count vectorizer to convert the values into one hot
encoded features
from sklearn.feature_extraction.text import CountVectorizer
vectorizer = CountVectorizer(vocabulary=list(sorted_cat_dict.
keys()), lowercase=False, 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 encoding ",categories_on
e_hot.shape)
['Warmth', 'Care_Hunger', 'History_Civics', 'M
usic_Arts', 'AppliedLearning', 'SpecialNeeds',
 'Health_Sports', 'Math_Science', 'Literacy_La
nguage']
Shape of matrix after one hot encoding (10924
8, 9)
```

In [71]:

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

```
vectorizer = CountVectorizer(vocabulary=list(sorted_sub_cat_d
ict.keys()), lowercase=False, binary=True)
vectorizer.fit(project_data['clean_subcategories'].values)
print(vectorizer.get_feature_names())

sub_categories_one_hot = vectorizer.transform(project_data['c
lean_subcategories'].values)
print("Shape of matrix after one hot encoding ",sub_categorie
s_one_hot.shape)
```

['Economics', 'CommunityService', 'FinancialLi teracy', 'ParentInvolvement', 'Extracurricular ', 'Civics_Government', 'ForeignLanguages', 'N utritionEducation', 'Warmth', 'Care_Hunger', 'SocialSciences', 'PerformingArts', 'CharacterE ducation', 'TeamSports', 'Other', 'College_Car eerPrep', 'Music', 'History_Geography', 'Healt h_LifeScience', 'EarlyDevelopment', 'ESL', 'Gy m_Fitness', 'EnvironmentalScience', 'VisualArt s', 'Health_Wellness', 'AppliedSciences', 'Spe cialNeeds', 'Literature_Writing', 'Mathematics ', 'Literacy']
Shape of matrix after one hot encoding (10924 8, 30)

In [72]:

```
# Please do the similar feature encoding with state, teacher_
prefix and project_grade_category also

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

In [73]:

```
school_state_cat_dict = dict(my_counter)
```

```
sorted_school_state_cat_dict = dict(sorted(school_state_cat_d
ict.items(), key=lambda kv: kv[1]))
```

In [74]:

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

vectorizer = CountVectorizer(vocabulary=list(sorted_school_st
ate_cat_dict.keys()), lowercase=False, binary=True)
vectorizer.fit(project_data['school_state'].values)
print(vectorizer.get_feature_names())

school_state_category_one_hot = vectorizer.transform(project_
data['school_state'].values)
print("Shape of matrix after one hot encoding ", school_state_
category_one_hot.shape)
```

['VT', 'WY', 'ND', 'MT', 'RI', 'SD', 'NE', 'DE ', 'AK', 'NH', 'WV', 'ME', 'HI', 'DC', 'NM', 'KS', 'IA', 'ID', 'AR', 'CO', 'MN', 'OR', 'KY', 'MS', 'NV', 'MD', 'CT', 'TN', 'UT', 'AL', 'WI', 'VA', 'AZ', 'NJ', 'OK', 'WA', 'MA', 'LA', 'OH', 'MO', 'IN', 'PA', 'MI', 'SC', 'GA', 'IL', 'NC', 'FL', 'NY', 'TX', 'CA']

Shape of matrix after one hot encoding (10924 8, 51)

In [75]:

```
my_counter = Counter()
for project_grade in project_data['project_grade_category'].v
alues:
    my_counter.update(project_grade.split())
```

In [76]:

```
project_grade_cat_dict = dict(my_counter)
```

```
sorted_project_grade_cat_dict = dict(sorted(project_grade_cat
_dict.items(), key=lambda kv: kv[1]))
                                                      In [77]:
## we use count vectorizer to convert the values into one hot
 encoded features
vectorizer = CountVectorizer(vocabulary=list(sorted_project_g
rade_cat_dict.keys()), lowercase=False, binary=True)
vectorizer.fit(project_data['project_grade_category'].values)
print(vectorizer.get_feature_names())
project_grade_category_one_hot = vectorizer.transform(project
_data['project_grade_category'].values)
print("Shape of matrix after one hot encoding ",project_grade
_category_one_hot.shape)
['Grades_9_12', 'Grades_6_8', 'Grades_3_5', 'G
rades_PreK_2']
Shape of matrix after one hot encoding (10924
8, 4)
                                                      In [78]:
my_counter = Counter()
for teacher_prefix in project_data['teacher_prefix'].values:
    teacher_prefix = str(teacher_prefix)
    my_counter.update(teacher_prefix.split())
                                                      In [79]:
teacher_prefix_cat_dict = dict(my_counter)
sorted_teacher_prefix_cat_dict = dict(sorted(teacher_prefix_c
at_dict.items(), key=lambda kv: kv[1]))
```

In [80]:

#https://stackoverflow.com/questions/39303912/tfidfvectorizer

```
-in-scikit-learn-valueerror-np-nan-is-an-invalid-document
#ValueError: np.nan is an invalid document, expected byte or
unicode string.
vectorizer = CountVectorizer(vocabulary=list(sorted_teacher_p
refix_cat_dict.keys()), lowercase=False, binary=True)
vectorizer.fit(project_data['teacher_prefix'].values.astype("
U"))
print(vectorizer.get_feature_names())

teacher_prefix_categories_one_hot = vectorizer.transform(proj
ect_data['teacher_prefix'].values.astype("U"))
print("Shape of matrix after one hot encoding ",teacher_prefi
x_categories_one_hot.shape)
```

```
['Dr.', 'Teacher', 'Mr.', 'Ms.', 'Mrs.']
Shape of matrix after one hot encoding (10924
8, 5)
```

1.4.2 Vectorizing Text data

1.4.2.1 Bag of words

8, 16623)

In [81]:

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

Shape of matrix after one hot encoding (10924

1.4.2.2 Bag of Words on $project_tit \leq$

In [82]:

```
#you can vectorize the title also
#before you vectorize the title make sure you preprocess it
vectorizer = CountVectorizer(min_df=10)
project_title_bow = vectorizer.fit_transform(preprocessed_tit
les)
print("Shape of matrix after one hot encoding ",project_title
_bow.shape)
```

Shape of matrix after one hot encoding (10924 8, 3329)

1.4.2.3 TFIDF vectorizer

In [83]:

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

Shape of matrix after one hot encoding (10924 8, 16623)

1.4.2.4 TFIDF Vectorizer on $project_tit \leq$

In [84]:

```
# Similarly you can vectorize for title also
vectorizer = TfidfVectorizer(min_df=10)
```

```
project_titles_tfidf = vectorizer.fit_transform(preprocessed_
titles)
print("Shape of matrix after one hot encoding ",project_title
s_tfidf.shape)
```

Shape of matrix after one hot encoding (10924 8, 3329)

1.4.2.5 Using Pretrained Models: Avg W2V

In [80]:

```
111
# Reading glove vectors in python: https://stackoverflow.com/
a/38230349/4084039
def loadGloveModel(gloveFile):
   print ("Loading Glove Model")
   f = open(gloveFile, 'r', encoding="utf8")
   model = \{\}
   for line in tqdm(f):
       splitLine = line.split()
       word = splitLine[0]
       embedding = np.array([float(val) for val in splitLine
[1:]]
       model[word] = embedding
   print ("Done.",len(model)," words loaded!")
   return model
model = loadGloveModel('glove.42B.300d.txt')
Output:
Loading Glove Model
1917495it [06:32, 4879.69it/s]
Done. 1917495 words loaded!
```

```
words = []
for i in preproced_texts:
   words.extend(i.split(' '))
for i in preproced_titles:
   words.extend(i.split(' '))
print("all the words in the coupus", len(words))
words = set(words)
print("the unique words in the coupus", len(words))
inter_words = set(model.keys()).intersection(words)
print("The number of words that are present in both glove vec
tors and our coupus", \
      len(inter_words), "(", np.round(len(inter_words)/len(word
s)*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.je
ssicayung.com/how-to-use-pickle-to-save-and-load-variables-in
-python/
import pickle
with open('glove_vectors', 'wb') as f:
   pickle.dump(words_courpus, f)
IIII
```

'\n# Reading glove vectors in python: https:// stackoverflow.com/a/38230349/4084039\ndef load GloveModel(gloveFile):\n print ("Loading Gl ove Model")\n f = open(gloveFile,\'r\', enc oding="utf8")\n $model = {} \n$ for line in tqdm(f):\n splitLine = line.split()\n word = splitLine[0]\n embedding = np.array([float(val) for val in splitLine[1: model[word] = embedding\n t ("Done.",len(model)," words loaded!")\n eturn model\nmodel = loadGloveModel(\'glove.42 B.300d.txt\')\n\n# ============ \nLoading Glove Model\n1917495 =\nOutput:\n it [06:32, 4879.69it/s]\nDone. 1917495 words loaded!\n\n# =======\n\nw ords = []\nfor i in preproced_texts:\n s.extend(i.split(\' \'))\n\nfor i in preproced _titles:\n words.extend(i.split(\' \'))\npr int("all the words in the coupus", len(words)) \nwords = set(words)\nprint("the unique words in the coupus", len(words))\n\ninter_words = s et(model.keys()).intersection(words)\nprint("T he number of words that are present in both gl ove vectors and our coupus", len(inter_w ords), "(", np.round(len(inter_words)/len(words) *100,3),"%)")\n\nwords_courpus = {}\nwords_glo ve = set(model.keys())\nfor i in words:\n f i in words_glove:\n words_courpus[i] = model[i]\nprint("word 2 vec length", len(wor ds_courpus))\n\n# stronging variables into p ickle files python: http://www.jessicayung.com /how-to-use-pickle-to-save-and-load-variablesin-python/\n\nimport pickle\nwith open(\'glove _vectors\', \'wb\') as f:\n pickle.dump(wor ds_courpus, f)\n\n\n'

```
In [85]:
```

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

In [86]:

```
# average Word2Vec
# compute average word2vec for each review.
avg_w2v_vectors = []; # the avg-w2v for each sentence/review
is stored in this list
for sentence in tqdm(preprocessed_essays): # for each review/
sentence
    vector = np.zeros(300) # as word vectors are of zero leng
th
    cnt_words =0; # num of words with a valid vector in the s
entence/review
    for word in sentence.split(): # for each word in a review
/sentence
        if word in glove_words:
            vector += model[word]
            cnt_words += 1
    if cnt_words != 0:
        vector /= cnt_words
    avg_w2v_vectors.append(vector)
print(len(avg_w2v_vectors))
print(len(avg_w2v_vectors[0]))
100%| 100%| 100248/109248 [00:19<00:00, 5
730.32it/s]
```

109248

1.4.2.6 Using Pretrained Models: AVG W2V on $project_tit \leq$

In [87]:

```
# Similarly you can vectorize for title also
# average Word2Vec
# compute average word2vec for each review.
avg_w2v_titles = []; # the avg-w2v for each sentence/review i
s stored in this list
for sentence in tqdm(preprocessed_titles): # for each review/
sentence
               vector = np.zeros(300) # as word vectors are of zero leng
 th
               cnt_words =0; # num of words with a valid vector in the s
entence/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_titles.append(vector)
 print(len(avg_w2v_titles))
print(len(avg_w2v_titles[0]))
100%| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1000| 1
02240.02it/s]
109248
300
```

1.4.2.7 Using Pretrained Models: TFIDF weighted W2V

In [88]:

```
# 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(t
fidf_model.idf_)))
tfidf_words = set(tfidf_model.get_feature_names())
```

In [89]:

```
# average Word2Vec
# compute average word2vec for each review.
tfidf_w2v_vectors = []; # the avg-w2v for each sentence/revie
w 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 leng
th
    tf_idf_weight =0; # num of words with a valid vector in t
he 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 w
ord
            # here we are multiplying idf value(dictionary[wo
rd]) and the tf value((sentence.count(word)/len(sentence.spli
t())))
            tf_idf = dictionary[word]*(sentence.count(word)/1
en(sentence.split())) # getting the tfidf value for each word
            vector += (vec * tf_idf) # calculating tfidf weig
```

In [90]:

```
# Similarly you can vectorize for title also

tfidf_model = TfidfVectorizer()
tfidf_model.fit(preprocessed_titles)
# we are converting a dictionary with word as a key, and the
idf as a value
dictionary = dict(zip(tfidf_model.get_feature_names(), list(t
fidf_model.idf_)))
tfidf_words = set(tfidf_model.get_feature_names())
```

1.4.2.9 Using Pretrained Models: TFIDF weighted W2V on $project_tit \leq$

In [91]:

```
# Similarly you can vectorize for title also
# average Word2Vec
# compute average word2vec for each review.
tfidf_w2v_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 leng
th
    tf_idf_weight =0; # num of words with a valid vector in t
he 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 w
ord
            # here we are multiplying idf value(dictionary[wo
rd]) and the tf value((sentence.count(word)/len(sentence.spli
t())))
            tf_idf = dictionary[word]*(sentence.count(word)/1
en(sentence.split())) # getting the tfidf value for each word
            vector += (vec * tf_idf) # calculating tfidf weig
hted w2v
            tf_idf_weight += tf_idf
    if tf_idf_weight != 0:
        vector /= tf_idf_weight
    tfidf_w2v_titles.append(vector)
print(len(tfidf_w2v_titles))
print(len(tfidf_w2v_titles[0]))
100%| 100%| 100248/109248 [00:02<00:00, 4
3812.54it/s]
109248
300
```

1.4.3 Vectorizing Numerical features

```
In [92]:
```

```
# check this one: https://www.youtube.com/watch?v=0H0q0cln3Z4
&t=530s
# standardization sklearn: https://scikit-learn.org/stable/mo
dules/generated/sklearn.preprocessing.StandardScaler.html
from sklearn.preprocessing import StandardScaler
# price_standardized = standardScalar.fit(project_data['price
'].values)
# this will rise the error
# ValueError: Expected 2D array, got 1D array instead: array=
[725.05 213.03 329. ... 399. 287.73 5.5].
# Reshape your data either using array.reshape(-1, 1)
price_scalar = StandardScaler()
price_scalar.fit(project_data['price'].values.reshape(-1,1))
# finding the mean and standard deviation of this data
print(f"Mean : {price_scalar.mean_[0]}, Standard deviation :
{np.sqrt(price_scalar.var_[0])}")
# Now standardize the data with above maen and variance.
price_standardized = price_scalar.transform(project_data['pri
ce'].values.reshape(-1, 1))
Mean : 298.1193425966608, Standard deviation :
367.49634838483496
                                                      In [93]:
price_standardized
                                                      Out[93]:
array([[-0.3905327],
       [ 0.00239637],
       [ 0.59519138],
       [-0.15825829],
```

```
[-0.51216657]])
                                                       In [94]:
quantity_scalar = StandardScaler()
## Finding the mean and standard deviation of this data
quantity_scalar.fit(project_data['quantity'].values.reshape(-1
,1))
print("Mean : {}".format(quantity_scalar.mean_[0]))
print("Standard deviation : {}".format(np.sqrt(quantity_scala
r.var_[0])))
# Now standardize the data with above maen and variance.
quantity_standardized = quantity_scalar.transform(project_dat
a['quantity'].values.reshape(-1, 1))
Mean: 16.965610354422964
Standard deviation : 26.182821919093175
                                                       In [95]:
quantity_standardized
                                                       Out[95]:
array([[ 0.23047132],
       [-0.60977424],
       [ 0.19227834],
       [-0.4951953],
       [-0.03687954],
       [-0.45700232]])
                                                       In [96]:
```

[-0.61243967],

```
previously_posted_projects_scalar = StandardScaler()

## Finding the mean and standard deviation of this data
previously_posted_projects_scalar.fit(project_data['teacher_n
umber_of_previously_posted_projects'].values.reshape(-1,1))

print("Mean : {}".format(previously_posted_projects_scalar.me
an_[0]))

print("Standard deviation : {}".format(np.sqrt(previously_posted_projects_scalar.var_[0])))

# Now standardize the data with above maen and variance.
previously_posted_projects_scalar_standardized = previously_p
osted_projects_scalar.transform(project_data['teacher_number_
of_previously_posted_projects'].values.reshape(-1, 1))
```

Mean: 11.153165275336848

Standard deviation : 27.77702641477403

1.4.4 Merging all the above features

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

```
In [97]:
```

```
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 [98]:

```
# merge two sparse matrices: https://stackoverflow.com/a/1971
0648/4084039
from scipy.sparse import hstack
# with the same hstack function we are concatinating a sparse
matrix and a dense matirx :)
X = hstack((categories_one_hot, sub_categories_one_hot, text_
bow, price_standardized))
X.shape
```

Out[98]:

(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_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 datatpoins you are using

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

```
In [99]:
#stacking all the layers of columns from the above shells
X = hstack((categories_one_hot, sub_categories_one_hot, schoo
l_state_category_one_hot, project_grade_category_one_hot, tea
cher_prefix_categories_one_hot, price_standardized, quantity_
standardized, previously_posted_projects_scalar_standardized,
 project_title_bow))
X.shape
                                                       Out[99]:
(109248, 3431)
                                                      In [100]:
#Plotting only first 5k datapoints due to computational const
raints
from sklearn.manifold import TSNE
X = X.tocsr() #Tsne only accepts dense matrices
X_{5000} = X[0:5000,:]
                                                      In [101]:
X_{5000} = X_{5000}.toarray()
model = TSNE(n_components = 2, perplexity = 50, random_state
= 0)
tsne_data_bow = model.fit_transform(X_5000)
                                                      In [102]:
```

```
labels = project_data["project_is_approved"]
labels_{5000} = labels_{0:5000}
```

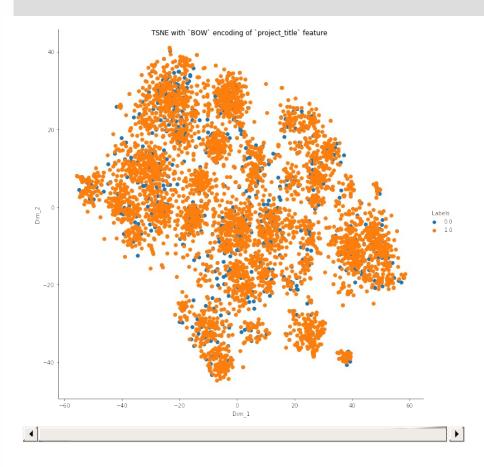
```
In [103]:
```

```
tsne_data_bow = np.vstack((tsne_data_bow.T, labels_5000)).T
tsne_df_bow = pd.DataFrame(tsne_data_bow, columns = ("Dim_1",
"Dim_2","Labels"))
```

In [104]:

```
# 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 h
elpful to the reader
    # b. Legends if needed
    # c. X-axis label
    # d. Y-axis label

sns.FacetGrid(tsne_df_bow, hue = "Labels", size = 10).map(plt
.scatter, "Dim_1", "Dim_2").add_legend().fig.suptitle("TSNE w
ith `BOW` encoding of `project_title` feature ")
plt.show()
```



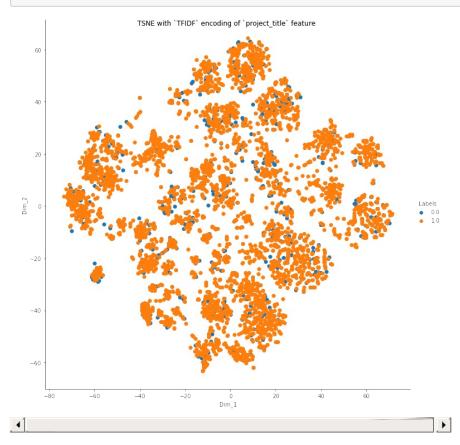
- There is a lot of overlapping in the datapoints
- The datapoints based on their class labels are not well separated,not much sense can be made out of this plot

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

```
In [105]:
X = hstack((categories_one_hot, sub_categories_one_hot, schoo
l_state_category_one_hot, project_grade_category_one_hot, tea
cher_prefix_categories_one_hot, price_standardized, quantity_
standardized, previously_posted_projects_scalar_standardized,
 project_titles_tfidf))
X.shape
                                                      Out[105]:
(109248, 3431)
                                                      In [106]:
X = X.tocsr() #Tsne only accepts dense matrices
X_5000 = X[0:5000,:]
                                                      In [107]:
X_{5000} = X_{5000}.toarray()
model = TSNE(n_components = 2, perplexity = 50, random_state
= ⊙)
tsne_data_tfidf = model.fit_transform(X 5000)
                                                      In [108]:
tsne_data_tfidf = np.vstack((tsne_data_tfidf.T, labels_5000))
. T
tsne_df_tfidf = pd.DataFrame(tsne_data_tfidf, columns = ("Dim
_1", "Dim_2", "Labels"))
```

In [109]:

```
sns.FacetGrid(tsne_df_tfidf, hue = "Labels", size = 10).map(p
lt.scatter, "Dim_1", "Dim_2").add_legend().fig.suptitle("TSNE
with `TFIDF` encoding of `project_title` feature ")
plt.show()
```



- The datapoints seem to be somewhat scattered, but the overlapping still exists.
- Not much sense can be made out of this plot

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

```
In [110]:
```

```
X = hstack((categories_one_hot, sub_categories_one_hot, school
l_state_category_one_hot, project_grade_category_one_hot, tea
cher_prefix_categories_one_hot, price_standardized, quantity_
standardized, previously_posted_projects_scalar_standardized,
 avg_w2v_titles))
X.shape
                                                      Out[110]:
(109248, 402)
                                                      In [111]:
X = X.tocsr() #Tsne only accepts dense matrices
X_{5000} = X[0:5000,:]
                                                      In [112]:
X_{5000} = X_{5000}.toarray()
model = TSNE(n_components = 2, perplexity = 50, random_state
```

In [113]:

```
tsne_data_avg_w2v = np.vstack((tsne_data_avg_w2v.T, labels_50
00)).T
tsne_df_avg_w2v = pd.DataFrame(tsne_data_avg_w2v, columns = (
"Dim_1", "Dim_2", "Labels"))
```

tsne data avg w2v = model.fit transform(X 5000)

= ⊙)

In [114]:

```
sns.FacetGrid(tsne_df_avg_w2v, hue = "Labels", size = 10).map
(plt.scatter, "Dim_1", "Dim_2").add_legend().fig.suptitle("TS
NE with `AVG W2V` encoding of `project_title` feature ")
plt.show()
```



- There is a lot of overlapping in the datapoints
- The datapoints based on their class labels are not well scattered or separated, so no proper conclusion can be drawn

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

```
In [115]:
```

```
X = hstack((categories_one_hot, sub_categories_one_hot, schoo
l_state_category_one_hot, project_grade_category_one_hot, tea
cher_prefix_categories_one_hot, price_standardized, quantity_
standardized, previously_posted_projects_scalar_standardized,
tfidf_w2v_titles))
X.shape
```

Out[115]:

(109248, 402)

In [116]:

```
X = X.tocsr() #Tsne only accepts dense matrices
X_5000 = X[0:5000,:]
```

In [117]:

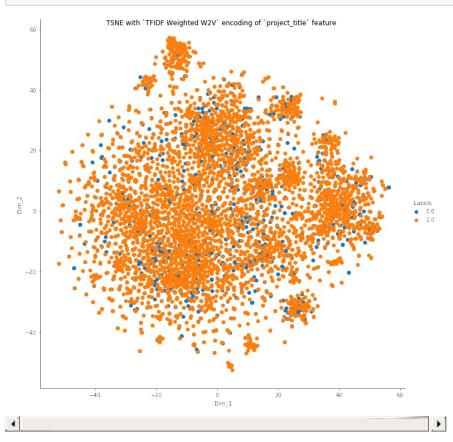
```
X_5000 = X_5000.toarray()
model = TSNE(n_components = 2, perplexity = 50, random_state
= 0)
tsne_data_tfidf_w2v = model.fit_transform(X_5000)
```

In [118]:

```
tsne_data_tfidf_w2v = np.vstack((tsne_data_tfidf_w2v.T, label
s_5000)).T
tsne_df_tfidf_w2v = pd.DataFrame(tsne_data_tfidf_w2v, columns
= ("Dim_1", "Dim_2", "Labels"))
```

In [119]:

```
sns.FacetGrid(tsne_df_tfidf_w2v, hue = "Labels", size = 10).m
ap(plt.scatter, "Dim_1", "Dim_2").add_legend().fig.suptitle("
TSNE with `TFIDF Weighted W2V` encoding of `project_title` fe
ature ")
plt.show()
```



- Even here, there seems to be a lot of overlapping in the datapoints
- The datapoints based on their class labels are not well scattered or separated, so no proper conclusion can be drawn from this plot

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

```
In [120]:
```

```
X = hstack((categories_one_hot, sub_categories_one_hot, schoo
l_state_category_one_hot, project_grade_category_one_hot, tea
cher_prefix_categories_one_hot, price_standardized, quantity_
standardized, previously_posted_projects_scalar_standardized,
project_title_bow,project_titles_tfidf, avg_w2v_titles, tfidf
_w2v_titles))
X.shape
```

Out[120]:

(109248, 7360)

In [121]:

```
X = X.tocsr() #Tsne only accepts dense matrices
X_5000 = X[0:5000,:]
```

In [122]:

```
X_5000 = X_5000.toarray()
model = TSNE(n_components = 2, perplexity = 50, random_state
= 0)
tsne_data_final = model.fit_transform(X_5000)
```

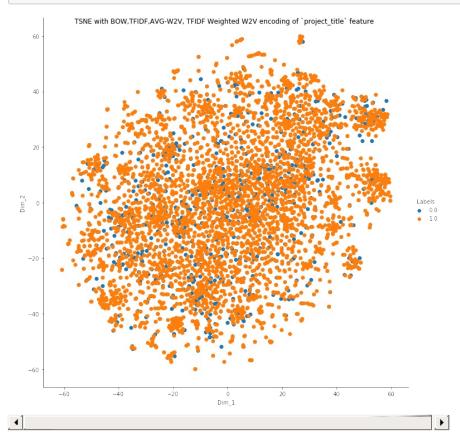
In [123]:

```
tsne_data_final = np.vstack((tsne_data_final.T, labels_5000))
.T
tsne_df_final = pd.DataFrame(tsne_data_final, columns = ("Dim
```

```
_1","Dim_2","Labels"))
```

In [124]:

```
sns.FacetGrid(tsne_df_final, hue = "Labels", size = 10).map(p
lt.scatter, "Dim_1", "Dim_2").add_legend().fig.suptitle("TSNE
with BOW,TFIDF,AVG-W2V, TFIDF Weighted W2V encoding of `proj
ect_title` feature ")
plt.show()
```



- Even here, there seems to be a lot of overlapping in the datapoints
- The datapoints based on their class labels are not

well scattered or separated,so no proper conclusion can be drawn from this plot

Conclusion

- The EDA of this dataset gives us some useful insights on how the features are correlated with the approval status of a project proposal. However, that can't be generalized, so we used T-SNE with the objective of grouping datapoints based on their class labels.
- However, The Visualisation of TSNE with Bag of Words, TF-IDF, Avg Word2Vec, TF-IDF Weighted Word2Vec does not seem to yield the expected result of clustering similar data points
- So alternate methods have to be tried on this dataset so that the approval status of thousands of project proposals could be automated instead of manually screening each of them