

DonorsChoose

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

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

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

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

About the DonorsChoose Data Set

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

Feature		Description
<code>project_id</code>		A unique identifier for the proposed project. Example: p036502
<code>project_title</code>	<ul style="list-style-type: none">••	Title of the project. Examples: Art Will Make You Happy! First Grade Fun
<code>project_grade_category</code>	<ul style="list-style-type: none">••••	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
<code>project_subject_categories</code>	<ul style="list-style-type: none">•••••••••	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
	<ul style="list-style-type: none">••	Examples: Music & The Arts Literacy & Language, Math & Science
<code>school_state</code>		State where school is located (Two-letter U.S. postal code (https://en.wikipedia.org/wiki/List_of_U.S._state_abbreviations#Postal_codes)). Example: WY
<code>project_subject_subcategories</code>	<ul style="list-style-type: none">••	One or more (comma-separated) subject subcategories for the project. Examples: Literacy Literature & Writing, Social Sciences
<code>project_resource_summary</code>	<ul style="list-style-type: none">•	An explanation of the resources needed for the project. Example: My students need hands on literacy materials to manage sensory needs!
<code>project_essay_1</code>		First application essay*
<code>project_essay_2</code>		Second application essay*
<code>project_essay_3</code>		Third application essay*
<code>project_essay_4</code>		Fourth application essay*
<code>project_submitted_datetime</code>		Datetime when project application was submitted. Example: 2016-04-28 12:43:56.245
<code>teacher_id</code>		A unique identifier for the teacher of the proposed project. Example: bdf8baa8fedef6bfeec7ae4ff1c15c56
		Teacher's title. One of the following enumerated values:

	•	nan
	•	Dr.
teacher_prefix	•	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:

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

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

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

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

Notes on the Essay Data

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

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

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

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

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

In [1]:

```
%matplotlib inline
import warnings
warnings.filterwarnings("ignore")

import sqlite3
import pandas as pd
import numpy as np
import nltk
import string
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.feature_extraction.text import TfidfTransformer
from sklearn.feature_extraction.text import TfidfVectorizer

from sklearn.feature_extraction.text import CountVectorizer
from sklearn.metrics import confusion_matrix
from sklearn import metrics
from sklearn.metrics import roc_curve, auc
from nltk.stem.porter import PorterStemmer

import re
# Tutorial about Python regular expressions: https://pymotw.com/2/re/
import string
from nltk.corpus import stopwords
from nltk.stem import PorterStemmer
from nltk.stem.wordnet import WordNetLemmatizer

from gensim.models import Word2Vec
from gensim.models import KeyedVectors
import pickle

from tqdm import tqdm
import os

from plotly import plotly
import plotly.offline as offline
import plotly.graph_objs as go
offline.init_notebook_mode()
from collections import Counter
```

1.1 Reading Data

In [2]:

```
project_data = pd.read_csv('train_data.csv',nrows=30000)
resource_data = pd.read_csv('resources.csv')
```

In [3]:

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

Number of data points in train data (30000, 17)

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

In [4]:

```
print("Number of data points in train data", resource_data.shape)
print(resource_data.columns.values)
resource_data.head(2)
```

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

Out[4]:

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

1.2 Data Analysis

In [5]:

```
# PROVIDE CITATIONS TO YOUR CODE IF YOU TAKE IT FROM ANOTHER WEBSITE.
# https://matplotlib.org/gallery/pie_and_polar_charts/pie_and_donut_labels.html#sphx-glr-gallery-pie-and-polar-charts-pie-and-donut-labels-py

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

fig, ax = plt.subplots(figsize=(6, 6), subplot_kw=dict(aspect="equal"))
recipe = ["Accepted", "Not Accepted"]

data = [y_value_counts[1], y_value_counts[0]]

wedges, texts = ax.pie(data, wedgeprops=dict(width=0.5), startangle=-40)

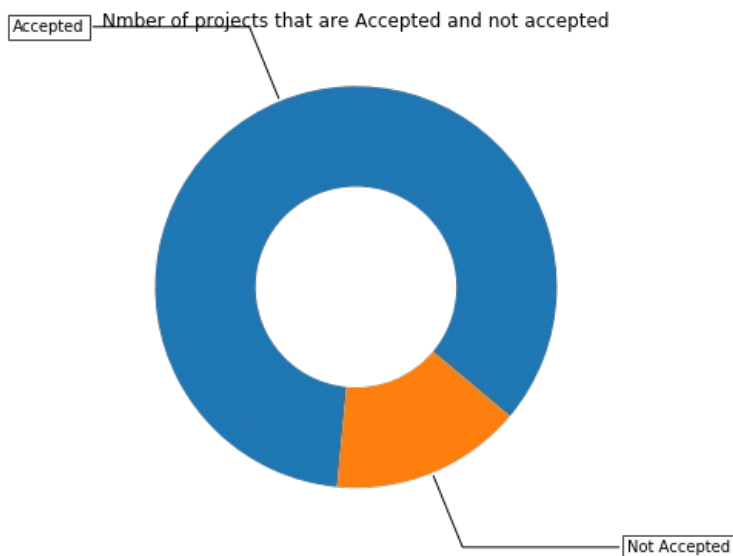
bbox_props = dict(boxstyle="square,pad=0.3", fc="w", ec="k", lw=0.72)
kw = dict(xycoords='data', textcoords='data', arrowprops=dict(arrowstyle="-"),
          bbox=bbox_props, zorder=0, va="center")

for i, p in enumerate(wedges):
    ang = (p.theta2 - p.theta1)/2. + p.theta1
    y = np.sin(np.deg2rad(ang))
    x = np.cos(np.deg2rad(ang))
    horizontalalignment = {-1: "right", 1: "left"}[int(np.sign(x))]
    connectionstyle = "angle,angleA=0,angleB={}".format(ang)
    kw["arrowprops"].update({"connectionstyle": connectionstyle})
    ax.annotate(recipe[i], xy=(x, y), xytext=(1.35*np.sign(x), 1.4*y),
                horizontalalignment=horizontalalignment, **kw)

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

plt.show()
```

Number of projects that are approved for funding 25380 , (84.6 %)
Number of projects that are not approved for funding 4620 , (15.4 %)



OBSERVATION 1:

- To observing the data(to exploring the whole data analysis) ,as we can see that the number of approved projects for funding is near about 85% ,also to see in the same obsevation in doughnut chart that is blue part is showing the approved project .
- The number of project that are not approved for funding is near about 15% ,as we can also see that in the doughnut chart Orange part is showing the number of not approved project .

1.2.1 Univariate Analysis: School State

In [6]:

```
# Pandas dataframe groupby count, mean: https://stackoverflow.com/a/19385591/4084039

temp = pd.DataFrame(project_data.groupby("school_state")["project_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']

# How to plot US state heatmap: https://datascience.stackexchange.com/a/9620

scl = [[0.0, 'rgb(242,240,247)'],[0.2, 'rgb(218,218,235)'],[0.4, 'rgb(188,189,220)'],\
       [0.6, 'rgb(158,154,200)'],[0.8, 'rgb(117,107,177)'],[1.0, 'rgb(84,39,143)']]

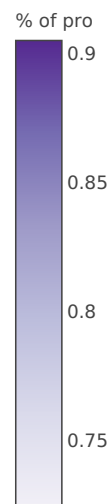
data = [ dict(
    type='choropleth',
    colorscale = scl,
    autocolorscale = False,
    locations = temp['state_code'],
    z = temp['num_proposals'].astype(float),
    locationmode = 'USA-states',
    text = temp['state_code'],
    marker = dict(line = dict (color = 'rgb(255,255,255)',width = 2)),
    colorbar = dict(title = "% of pro")
) ]

layout = dict (
    title = 'Project Proposals % of Acceptance Rate by US States',
    geo = dict(
        scope='usa',
        projection=dict( type='albers usa' ),
        showlakes = True,
        lakecolor = 'rgb(255, 255, 255)',
    ),
)

fig = go.Figure(data=data, layout=layout)
offline.iplot(fig, filename='us-map-heat-map')
```



Project Proposals % of Acceptance Rate by US States



OBSERVATION 2:

- To above heat map is showing the number of approvals and not approvals rate states wise.
- To analyse the school states , we can see in the US maps that below 85% (85% is the average approvals rate) is Vermont, District of Columbia, Texas, Montana, Louisiana are the states, which represent the minority of the project.
- The top highest rate of approvals which are above 85% is New Hampshire, Ohio, Washington, North Dakota, Delaware, are the states which account for the majority of the projects.
- The dark color zone which represent maximum acceptance rate and light color zone represent minimum acceptance rate.
- - maximum:DE state with 89.7% & minimum:VT state with 80.0%

In [7]:

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

States with lowest % approvals

	state_code	num_proposals
26	MT	0.723077
7	DC	0.775510
50	WY	0.787879
37	OR	0.798319
0	AK	0.802083

=====

States with highest % approvals

	state_code	num_proposals
17	KY	0.881159
47	WA	0.884323
30	NH	0.886364
32	NM	0.887324
8	DE	0.905263

In [8]:

```
#stacked bar plots matplotlib: https://matplotlib.org/gallery/lines\_bars\_and\_markers/bar\_stacked.html
def stack_plot(data, xtick, col2='project_is_approved', col3='total'):
    ind = np.arange(data.shape[0])

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

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

In [9]:

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

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

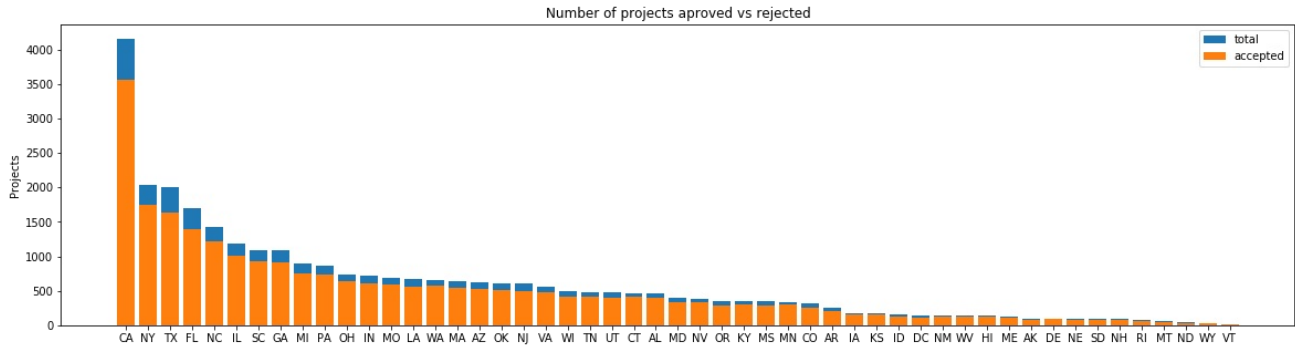
    temp.sort_values(by=['total'],inplace=True, ascending=False)

    if top:
        temp = temp[0:top]

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

In [10]:

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



	school_state	project_is_approved	total	Avg
4	CA	3564	4157	0.857349
34	NY	1745	2035	0.857494
43	TX	1638	2002	0.818182
9	FL	1399	1693	0.826344
27	NC	1219	1432	0.851257
=====				
39	RI	64	75	0.853333
26	MT	47	65	0.723077
28	ND	36	41	0.878049
50	WY	26	33	0.787879
46	VT	13	15	0.866667

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

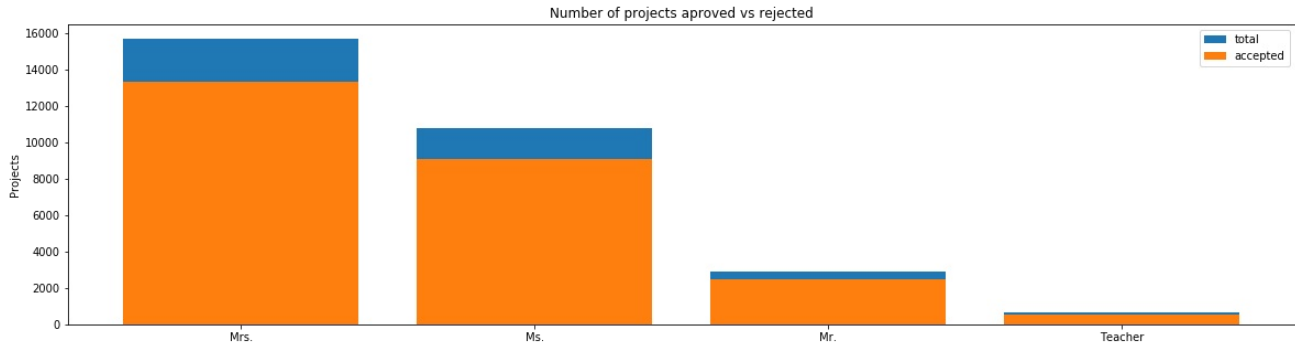
OBSERVATION 3:

- In above univariate analysis is between the features "project_is_approved" and "school state" to find the acceptance rate
- To analyse the above data, the lowest rate of acceptance is 80% so we can say that every state which has greater than 80% is success rate of acceptance.
- The states which has giving higher number of submission has more chances to acceptance rate ,but those states who has less number of submission also the chances is same like VT & WY.

1.2.2 Univariate Analysis: teacher_prefix

In [11]:

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



	teacher_prefix	project_is_approved	total	Avg
1	Mrs.	13310	15682	0.848744
2	Ms.	9099	10779	0.844141
0	Mr.	2457	2895	0.848705
3	Teacher	513	643	0.797823
=====				
1	Mrs.	13310	15682	0.848744
2	Ms.	9099	10779	0.844141
0	Mr.	2457	2895	0.848705
3	Teacher	513	643	0.797823

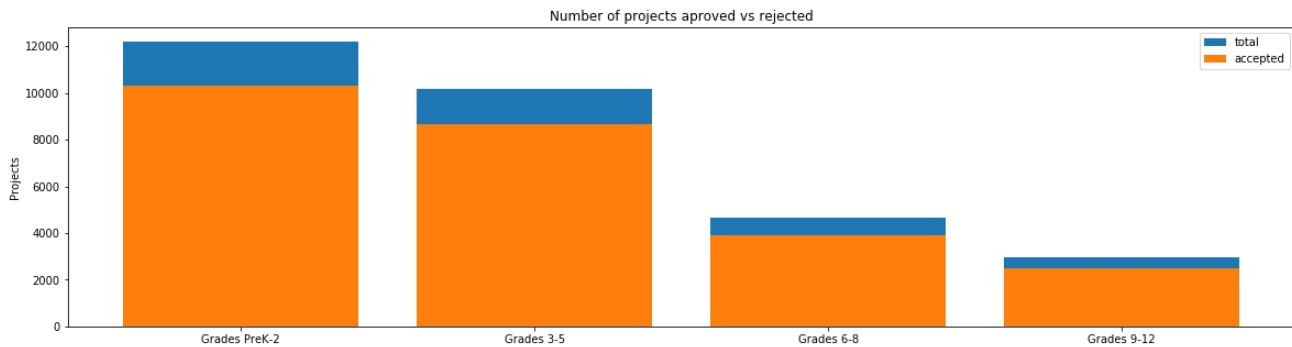
OBSERVATION 4:

- In above univariate analysis is between the features "**project_is_approved**" and "**teacher_prefix**" to find the acceptance rate
- The various teacher_prefix is used Mrs.,Ms.,teacher,Dr.,who submitted the project for approval.
- to analyse the data we can see highest number of submission and highest number of approval is from Mrs.
- Dr. has very less number of submission.

1.2.3 Univariate Analysis: project_grade_category

In [12]:

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



```
project_grade_category project_is_approved total Avg
3 Grades PreK-2 10324 12204 0.845952
0 Grades 3-5 8663 10160 0.852657
1 Grades 6-8 3897 4663 0.835728
2 Grades 9-12 2496 2973 0.839556
=====
project_grade_category project_is_approved total Avg
3 Grades PreK-2 10324 12204 0.845952
0 Grades 3-5 8663 10160 0.852657
1 Grades 6-8 3897 4663 0.835728
2 Grades 9-12 2496 2973 0.839556
```

OBSERVATION 5:

- Above univariate analysis is b/w '**project_is_approved**', "**project_grade_category**", features to find the acceptance rate among the students who is studying in various grades
- We explore the different categories of approved and not approved projects based on Category.
- All the grades are 80% above.
- Grades PreK-2 is the most popular followed by Grades 3-5, Grades 6-8 and Grades 9-12
- grades 3-5 is average number of approvals.
- The less number of submission and approval is grade 9-12 with 83%.

1.2.4 Univariate Analysis: project_subject_categories

In [13]:

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

# https://www.geeksforgeeks.org/removing-stop-words-nltk-python/
# https://stackoverflow.com/questions/23669024/how-to-strip-a-specific-word-from-a-string
# https://stackoverflow.com/questions/8270092/remove-all-whitespace-in-a-string-in-python
cat_list = []
for i in catogories:
    temp = ""
    # consider we have text like this "Math & Science, Warmth, Care & Hunger"
    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", "&
        ", "Science"
            j=j.replace('The', '') # if we have the words "The" we are going to replace it with ''(i.e removing 'T
            he')
        j = j.replace(' ', '') # we are placeing all the ' '(space) with ''(empty) ex:"Math & Science"=>"Math&Scie
        nce"
        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 [14]:

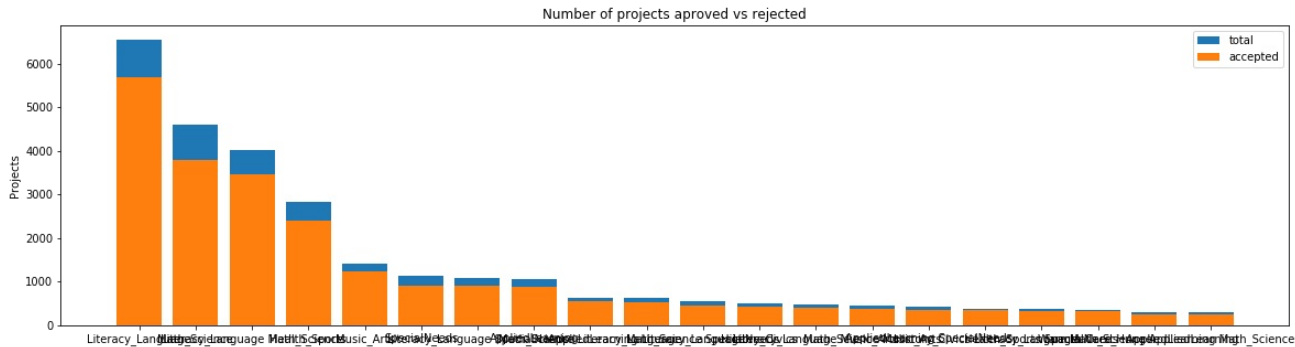
```
project_data['clean_categories'] = cat_list
project_data.drop(['project_subject_categories'], axis=1, inplace=True)
project_data.head(2)
```

Out[14]:

Unnamed: 0	id	teacher_id	teacher_prefix	school_state	project_submitted_datetime	pro
0	160221	p253737	c90749f5d961ff158d4b4d1e7dc665fc	Mrs.	IN	2016-12-05 13:43:57
1	140945	p258326	897464ce9ddc600bcd1151f324dd63a	Mr.	FL	2016-10-25 09:22:10

In [15]:

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



	clean_categories	project_is_approved	total	Avg
23	Literacy_Language	5680	6548	0.867440
31	Math_Science	3786	4607	0.821793
27	Literacy_Language Math_Science	3475	4017	0.865073
8	Health_Sports	2398	2839	0.844664
39	Music_Arts	1228	1426	0.861150
=====				
	clean_categories	project_is_approved	total	Avg
19	History_Civics Literacy_Language	355	386	0.919689
14	Health_Sports SpecialNeeds	317	378	0.838624
49	Warmth Care_Hunger	333	359	0.927577
32	Math_Science AppliedLearning	252	301	0.837209
4	AppliedLearning Math_Science	240	294	0.816327

OBSERVATION 6:

- Above univariate analysis is between "project_is_approved" ,"clean_categories" feautres to find the acceptance rate among the categories under which project is applied.
- Here we can see that the highest number of approvals(i.e. 92.5%) is for Warmth_care_Hunger .Most of the people interested for this project.
- The number of lowest acceptance rate is for AppliedLearning Math_Science i.e. 81.2%.

In [16]:

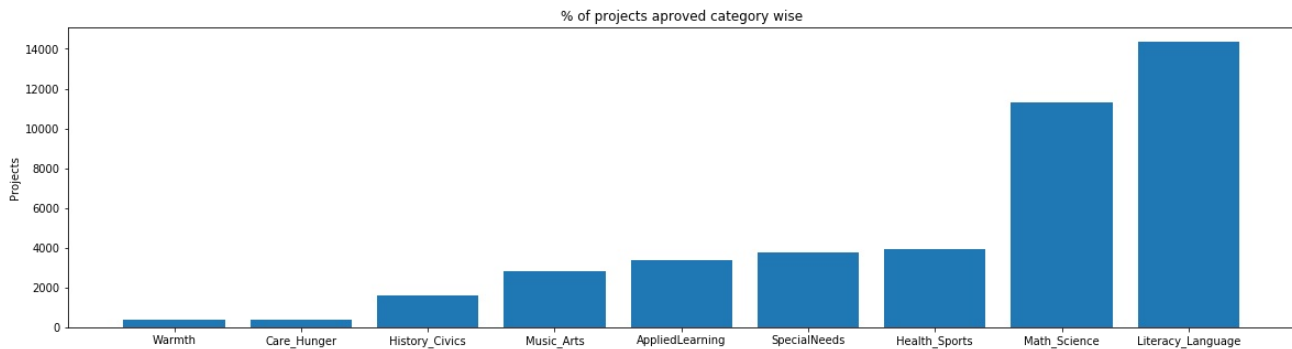
```
# count of all the words in corpus python: https://stackoverflow.com/a/22898595/4084039
from collections import Counter
my_counter = Counter()
for word in project_data['clean_categories'].values:
    my_counter.update(word.split())
```

In [17]:

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

ind = np.arange(len(sorted_cat_dict))
plt.figure(figsize=(20,5))
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()
```



OBSERVATION 7:

- In Barplot ,we can see all the individual categories are in incresing order.
- To analyse the barplot we can say ,most of the projects are in **math_science and Literacy_language** .
- **Warmth and care_Hunger** are the very less number of projects submission,but highest number of approval rate.

In [18]:

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

```
Warmth           :      384
Care_Hunger      :      384
History_Civics   :     1583
Music_Arts       :     2832
AppliedLearning  :     3374
SpecialNeeds     :     3751
Health_Sports    :     3918
Math_Science     :    11318
Literacy_Language :    14356
```

1.2.5 Univariate Analysis: project_subject_subcategories

In [19]:

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

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

sub_cat_list = []
for i in sub_catogories:
    temp = ""
    # consider we have text like this "Math & Science, Warmth, Care & Hunger"
    for j in i.split(','): # it will split it in three parts ["Math & Science", "Warmth", "Care & Hunger"]
        if 'The' in j.split(): # this will split each of the catogory based on space "Math & Science"=> "Math", "&
", "Science"
            j=j.replace('The','') # if we have the words "The" we are going to replace it with ''(i.e removing 'T
he')
        j = j.replace(' ','') # we are placeing all the ' '(space) with ''(empty) ex:"Math & Science"=>"Math&Scie
nce"
        temp +=j.strip()+" #" " abc ".strip() will return "abc", remove the trailing spaces
        temp = temp.replace('&','_')
    sub_cat_list.append(temp.strip())
```

In [20]:

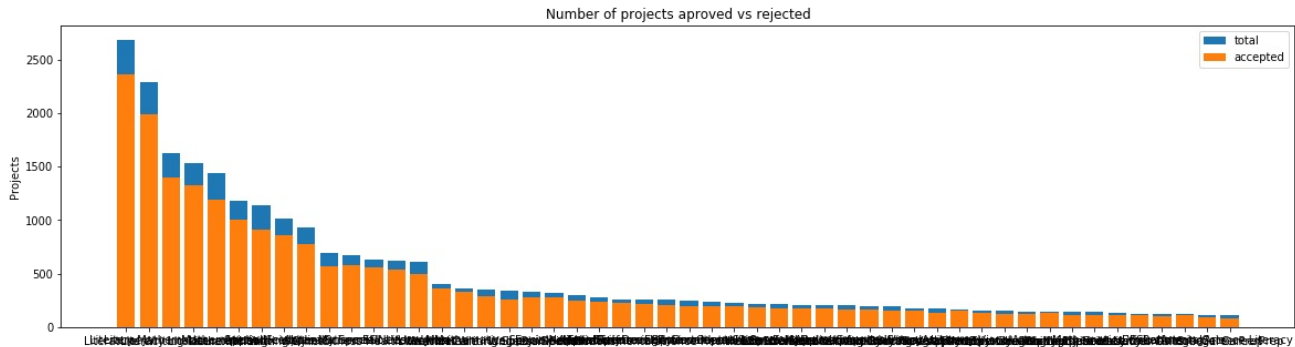
```
project_data['clean_subcategories'] = sub_cat_list
project_data.drop(['project_subject_subcategories'], axis=1, inplace=True)
project_data.head(2)
```

Out[20]:

Unnamed: 0	id	teacher_id	teacher_prefix	school_state	project_submitted_datetime	pro
0	160221	p253737	c90749f5d961ff158d4b4d1e7dc665fc	Mrs.	IN	2016-12-05 13:43:57
1	140945	p258326	897464ce9ddc600bcd1151f324dd63a	Mr.	FL	2016-10-25 09:22:10

In [21]:

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



	clean_subcategories	project_is_approved	total	Avg
279	Literacy	2366	2685	0.881192
281	Literacy Mathematics	1994	2289	0.871123
293	Literature Writing Mathematics	1401	1627	0.861094
280	Literacy Literature Writing	1331	1532	0.868799
303	Mathematics	1188	1440	0.825000
=====				
119	ESL	109	128	0.851562
3	AppliedSciences College_CareerPrep	100	121	0.826446
340	PerformingArts	109	120	0.908333
177	EnvironmentalScience Literacy	94	113	0.831858
75	College_CareerPrep	86	111	0.774775

OBSERVATION 8:

- Above univariate analysis is between "project_is_approved" and "clean_sub_categories" features to find the approval rate among the "sub_categories".
- To analyse the above graph ,the highest number of approval rate is for "Literacy" and "Literacy Mathematics",which is 88% and 87%.
- the lowest number acceptance rate is 81% which belong in "AppliedScience college_careerPrep" and "*"college _CareerPrep.

In [22]:

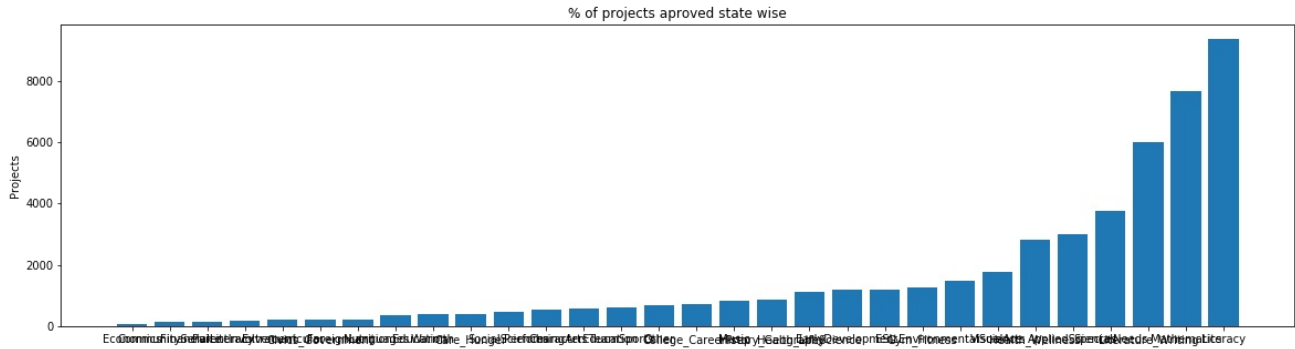
```
# count of all the words in corpus python: https://stackoverflow.com/a/22898595/4084039
from collections import Counter
my_counter = Counter()
for word in project_data['clean_subcategories'].values:
    my_counter.update(word.split())
```

In [23]:

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

ind = np.arange(len(sorted_sub_cat_dict))
plt.figure(figsize=(20,5))
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 [24]:

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

Economics	:	83
CommunityService	:	138
FinancialLiteracy	:	154
ParentInvolvement	:	172
Extracurricular	:	203
Civics_Government	:	227
ForeignLanguages	:	228
NutritionEducation	:	366
Warmth	:	384
Care_Hunger	:	384
SocialSciences	:	479
PerformingArts	:	536
CharacterEducation	:	577
TeamSports	:	601
Other	:	685
College_CareerPrep	:	722
Music	:	839
History_Geography	:	862
Health_LifeScience	:	1108
EarlyDevelopment	:	1197
ESL	:	1206
Gym_Fitness	:	1259
EnvironmentalScience	:	1495
VisualArts	:	1754
Health_Wellness	:	2825
AppliedSciences	:	2996
SpecialNeeds	:	3751
Literature_Writing	:	6014
Mathematics	:	7673
Literacy	:	9370

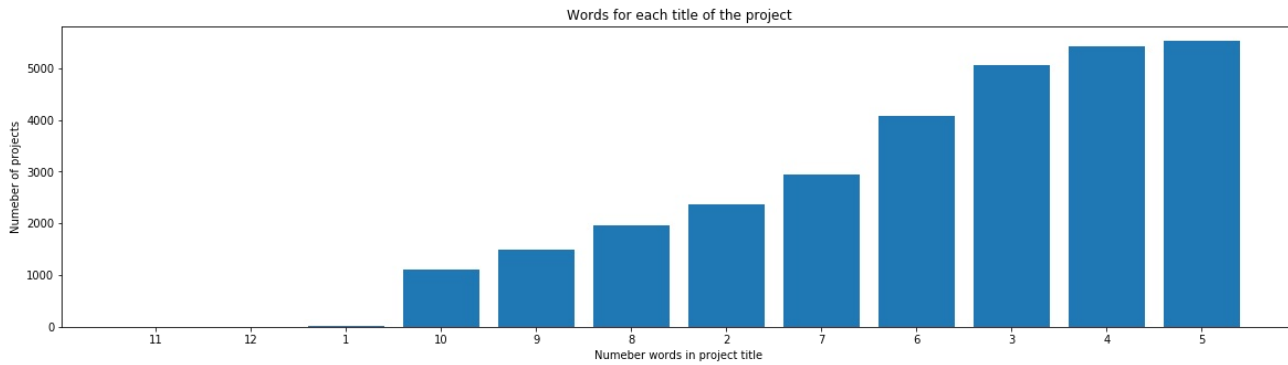
1.2.6 Univariate Analysis: Text features (Title)

In [25]:

```
#How to calculate number of words in a string in DataFrame: https://stackoverflow.com/a/37483537/4084039
word_count = project_data['project_title'].str.split().apply(len).value_counts()
word_dict = dict(word_count)
word_dict = dict(sorted(word_dict.items(), key=lambda kv: kv[1]))

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

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



OBSERVATION 9:

- Above barplot univariate Analysis between "**number_words_in_project_title**" and "**number_of_projects**".
- to analyse the barplot ,most of the number of projects are having the number of words > 4.
- ALSO their are very less number of projects with having number of words is 13,12,11,1.
- The role of the title is define that what project is all about.

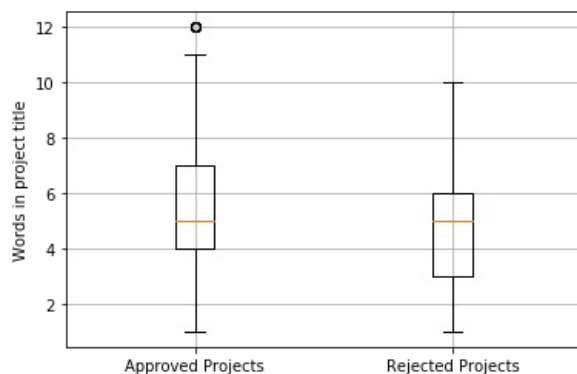
In [26]:

```
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 [27]:

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

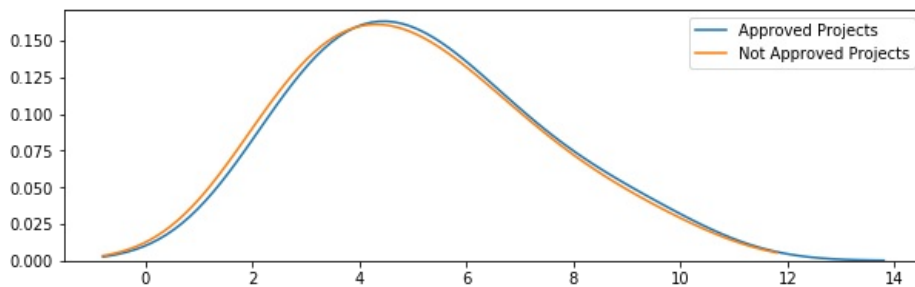


OBSERVATION 10:

- The above boxplot is between **"Approved_Project"** and **"Rejected_project"** vs **"Words_in_project_title"**.
- Both Approved_projects and Rejected_projects classified with 25 percentile and 75 percentile.
- The mean of both the group is almost similar.
- Rejected 25 percentile lied below Approved percentile with words in project title.
- To analyse the things we can say that there is more chance of rejection if project title has less num of words.

In [28]:

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



OBSERVATION 11:

- In above PDF is comparing the Probabilities between Approved projects Vs Not Approved projects.
- Blue line is shows Approved Projects and Orange line is shows Not Approved Projects.
- Blue line is always ahead of orange line.
- Peak of both is same that means Both lines mean override.
- To conclude the things we say that there is more chance of rejection if project title has less number of words.

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

In [29]:

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

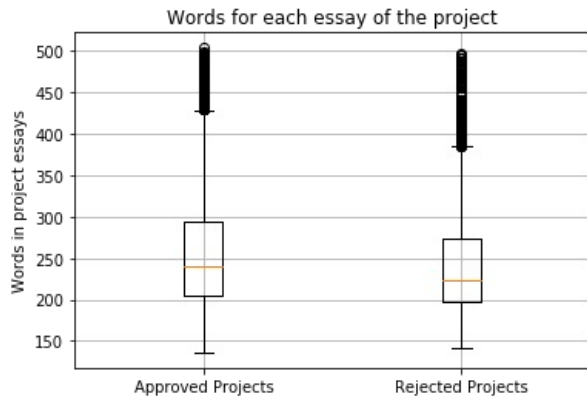
In [30]:

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

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

In [31]:

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

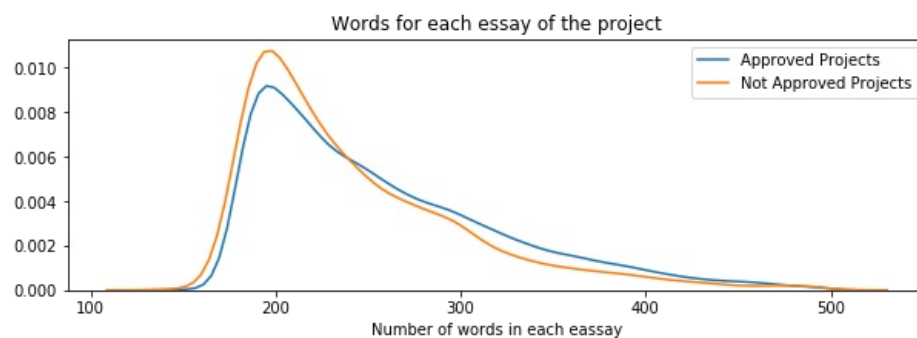


OBSERVATION 12

- This boxplot represent the Approval projects or Rejected projects between Words in projects essays
- To observe the things that more number of words it means that the person has written well or in brief, these may exist possibility to get approved the projects.
- and if the number of words is limited than might exist chance to get Rejection of the projects.
- We can observe from above boxplot - mean of Approved projects: **240 words** - mean of rejected projects: **225 words**
- There also exist difference in 75 percentile. -75 percentile of approved: **295 words** -75 percentile of rejected: **265 words**

In [32]:

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



OBSERVATION 13

- In above PDF is comparing the Probabilities b/w Approved projects vs Not Approved projects
- Blue line is shows Approved Projects and Orange line is shows Not Approved Projects.
- Blue line is always ahead of orange line.
- To conclude that there is more chance of rejection if words for each essay of the project is less.

1.2.8 Univariate Analysis: Cost per project

In [33]:

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

Out[33]:

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

In [34]:

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

Out[34]:

	id	price	quantity
0	p000001	459.56	7
1	p000002	515.89	21

In [35]:

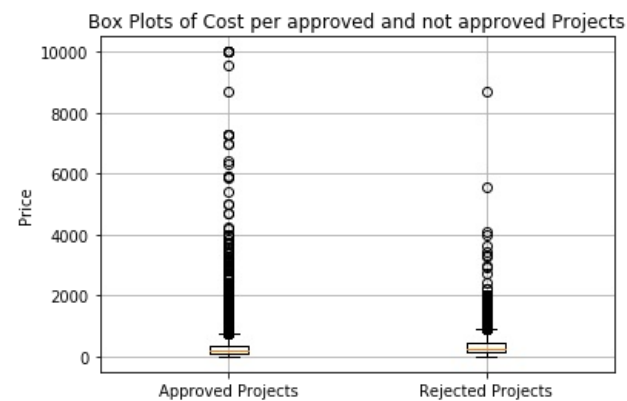
```
# join two dataframes in python:
project_data = pd.merge(project_data, price_data, on='id', how='left')
```

In [36]:

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

In [37]:

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

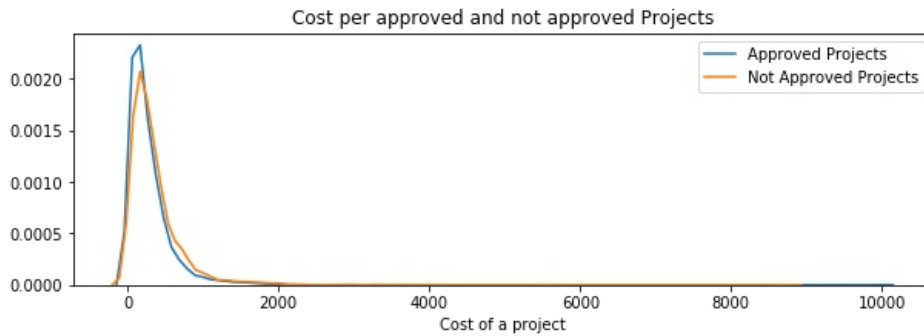


OBSERVATION 14

- Above boxplot represent the cost between approved projects vs not approved projects.
- Their is most of the points are overlapped both approved and not approved projects.
- Nothing is much clear, but Approved Projects is somehow dense

In [38]:

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



OBSERVATION 15

- Above PDF is comparing the cost per b/w Approved projects vs Not Approved projects.
- we can observing the gap between orange & blue line just before the 1000 along x-axis.
- To observe the things it may clear that if the cost project is affordable by the people than only there is chance to get approval of the projects.
- So it is the very important feature to take care while writing the projects.
- Cost should be affordable by the donors.
- It can conclude that there is more chance of rejection if the cost you are demanding is more.

In [39]:

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

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

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

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

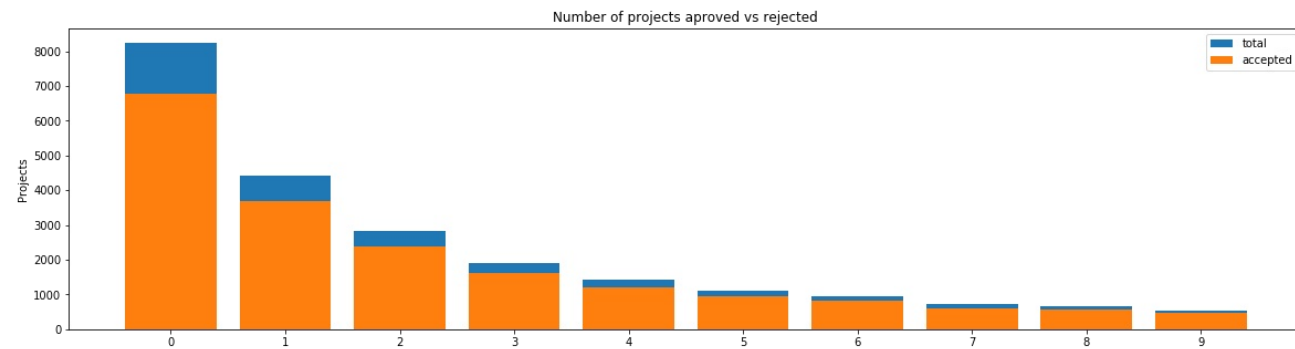
Percentile	Approved Projects	Not Approved Projects
0	0.66	1.97
5	13.858	39.625
10	34.0	71.053
15	58.429	96.349
20	78.7	116.408
25	99.99	136.938
30	117.56	158.215
35	137.93	179.99
40	157.0	203.948
45	178.422	230.376
50	199.0	255.945
55	223.739	284.823
60	254.898	314.21
65	284.791	353.083
70	320.993	392.25
75	367.482	435.568
80	412.232	497.496
85	479.0	596.679
90	598.491	722.792
95	811.176	967.552
100	9999.0	8719.69

1.2.9 Univariate Analysis: teacher_number_of_previously_posted_projects

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

In [40]:

```
univariate_barplots(project_data, 'teacher_number_of_previously_posted_projects', 'project_is_approved', top=10)
```



teacher_number_of_previously_posted_projects	project_is_approved	total	\
0	0	6768	8241
1	1	3671	4421
2	2	2382	2838
3	3	1613	1916
4	4	1213	1438

Avg

0	0.821260
1	0.830355
2	0.839323
3	0.841858
4	0.843533

teacher_number_of_previously_posted_projects	project_is_approved	total	\
5	5	937	1115
6	6	817	959
7	7	609	719
8	8	577	663
9	9	463	531

Avg

5	0.840359
6	0.851929
7	0.847010
8	0.870287
9	0.871940

OBSERVATION 16:

- The above stacked bar plot represent the Approval rate of **"teacher_number_of_previously_posted_projects"**
- To observe the data their is not much different in between the **"total_number_of_project"** and **"approval_of_the_projects"**.
- Highest acceptance rate for **teacher_number_of_previously_posted_projects 9** with **86.77%**

In [41]:

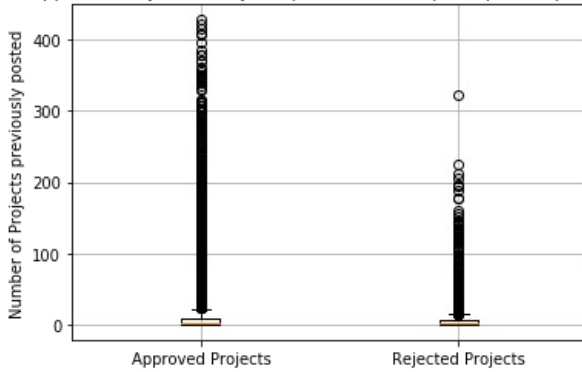
```
approved_details = project_data[project_data['project_is_approved']==1]['teacher_number_of_previously_posted_projects'].values

rejected_details = project_data[project_data['project_is_approved']==0]['teacher_number_of_previously_posted_projects'].values
```

In [42]:

```
plt.boxplot([approved_details, rejected_details])
plt.title('Box Plots of Approved/Rejected Projects per Number of prev. posted projects by teachers')
plt.xticks([1,2],('Approved Projects','Rejected Projects'))
plt.ylabel('Number of Projects previously posted')
plt.grid()
plt.show()
```

Box Plots of Approved/Rejected Projects per Number of prev. posted projects by teachers

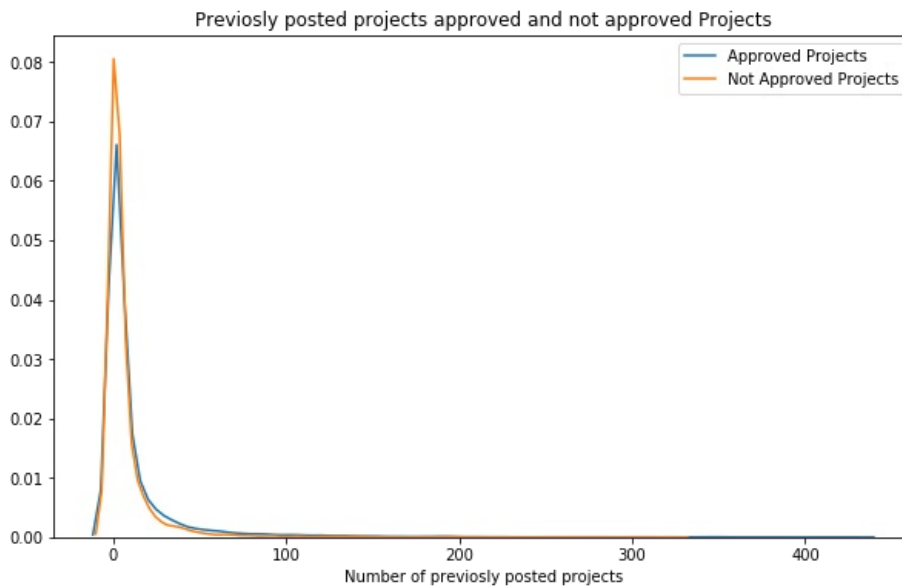


In [43]:

```
plt.figure(figsize=(10,6))
sns.distplot(approved_details, hist=False, label="Approved Projects")
sns.distplot(rejected_details, hist=False, label="Not Approved Projects")
plt.title('Previously posted projects approved and not approved Projects')
plt.xlabel('Number of previously posted projects')
plt.legend()
plt.show
```

Out[43]:

```
<function matplotlib.pyplot.show(*args, **kw)>
```



OBSERVATION 17:

1. Greater the number of submissions by teachers , greater is the acceptance rate
2. Thus, it is a great platform to accept more and more project ideas.

In [44]:

```
from prettytable import PrettyTable

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

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

Percentile	Approved Projects	Not Approved Projects
0	0.0	0.0
5	0.0	0.0
10	0.0	0.0
15	0.0	0.0
20	0.0	0.0
25	0.0	0.0
30	1.0	0.0
35	1.0	1.0
40	1.0	1.0
45	2.0	1.0
50	2.0	2.0
55	3.0	2.0
60	4.0	3.0
65	5.0	4.0
70	7.0	5.0
75	9.0	6.0
80	13.0	8.0
85	19.15	12.0
90	30.0	18.0
95	57.0	32.0
100	428.0	322.0

1.2.10 Univariate Analysis: project_resource_summary

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

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

1.3 Text preprocessing

In [45]:

```
resource_summaries = []

for data in project_data["project_resource_summary"] :
    resource_summaries.append(data)

resource_summaries[0:10]
```

Out[45]:

```
['My students need opportunities to practice beginning reading skills in English at home.',
'My students need a projector to help with viewing educational programs',
'My students need shine guards, athletic socks, Soccer Balls, goalie gloves, and training materials
for the upcoming Soccer season.',
'My students need to engage in Reading and Math in a way that will inspire them with these Mini iPads!',
'My students need hands on practice in mathematics. Having fun and personalized journals and charts
will help them be more involved in our daily Math routines.',
'My students need movement to be successful. Being that I have a variety of students that have all
different types of needs, flexible seating would assist not only these students with special needs,
but all students.',
'My students need some dependable laptops for daily classroom use for reading and math.',
'My students need ipads to help them access a world of online resources that will spark their interest in learning.',
'My students need three devices and three management licenses for small group's easy access to newly-implemented online programs--Go Noodle Plus, for increased in-class physical activity and Light Sail, an interactive reading program.',
'My students need great books to use during Independent Reading, Read Alouds, Partner Reading and Author Studies.']
```

```
len(resource_summaries)
```

```
# https://stackoverflow.com/questions/19859282/check-if-a-string-contains-a-number
numeric_summary_values = {}

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

```
numeric_digits = {}

for c in tqdm(range(len(resource_summaries))):
    if c in numeric_summary_values.keys():
        numeric_digits[c] = numeric_summary_values[c]
    else:
        numeric_digits[c] = 0
```

```
for i in range(20):
    print(numeric_digits[i])
```

```
digit_in_summary = []

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

```
print(digit_in_summary[0:100])
```

```
project_data['digit_in_summary'] = digit_in_summary
```

In [53]:

```
project_data.head(10)
```

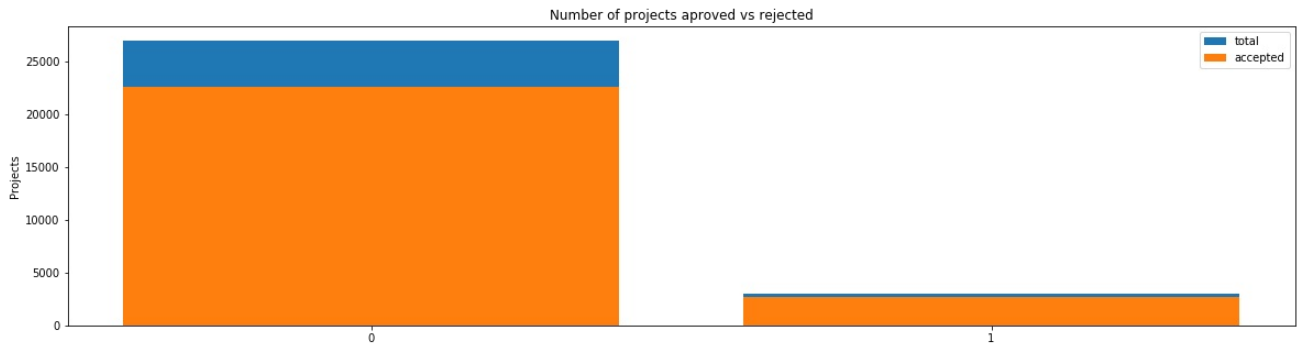
Out[53]:

	Unnamed: 0	id	teacher_id	teacher_prefix	school_state	project_submitted_datetime	pr
0	160221	p253737	c90749f5d961ff158d4b4d1e7dc665fc	Mrs.	IN	2016-12-05 13:43:57	
1	140945	p258326	897464ce9ddc600bcd1151f324dd63a	Mr.	FL	2016-10-25 09:22:10	
2	21895	p182444	3465aaf82da834c0582ebd0ef8040ca0	Ms.	AZ	2016-08-31 12:03:56	
3	45	p246581	f3cb9bffbba169bef1a77b243e620b60	Mrs.	KY	2016-10-06 21:16:17	
4	172407	p104768	be1f7507a41f8479dc06f047086a39ec	Mrs.	TX	2016-07-11 01:10:09	
5	141660	p154343	a50a390e8327a95b77b9e495b58b9a6e	Mrs.	FL	2017-04-08 22:40:43	
6	21147	p099819	9b40170bfa65e399981717ee8731efc3	Mrs.	CT	2017-02-17 19:58:56	
7	94142	p092424	5bfd3d12fae3d2fe88684bbac570c9d2	Ms.	GA	2016-09-01 00:02:15	
8	112489	p045029	487448f5226005d08d36bdd75f095b31	Mrs.	SC	2016-09-25 17:00:26	
9	158561	p001713	140eeac1885c820ad5592a409a3a8994	Ms.	NC	2016-11-17 18:18:56	

10 rows × 21 columns

In [54]:

```
univariate_barplots(project_data, 'digit_in_summary', 'project_is_approved', top=2)
```



	digit_in_summary	project_is_approved	total	Avg
0	0	22674	27011	0.839436
1	1	2706	2989	0.905320
=====				
	digit_in_summary	project_is_approved	total	Avg
0	0	22674	27011	0.839436
1	1	2706	2989	0.905320

OBSERVATION 18:

- 1. To observe the things that containing the numeric values have a very high acceptance rate of 92%.
- 2. The requirements with numerical figures ,greater will be the chance of acceptance of proposal .
- 3. It gives the clarity of quantity of resources so that nothing is wasted and can maximize the use of resources so that it will be the better help the children.

1.3.1 Essay Text

In [55]:

```
project_data.head(5)
```

Out[55]:

Unnamed: 0	id	teacher_id	teacher_prefix	school_state	project_submitted_datetime	p
0	160221 p253737	c90749f5d961ff158d4b4d1e7dc665fc	Mrs.	IN	2016-12-05 13:43:57	
1	140945 p258326	897464ce9ddc600bcd1151f324dd63a	Mr.	FL	2016-10-25 09:22:10	
2	21895 p182444	3465aaf82da834c0582ebd0ef8040ca0	Ms.	AZ	2016-08-31 12:03:56	
3	45 p246581	f3cb9bffbba169bef1a77b243e620b60	Mrs.	KY	2016-10-06 21:16:17	
4	172407 p104768	be1f7507a41f8479dc06f047086a39ec	Mrs.	TX	2016-07-11 01:10:09	

5 rows x 7 columns

```
# 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[2000])
print("="*50)
print(project_data['essay'].values[3000])
print("="*50)
```

The 51 fifth grade students that will cycle through my classroom this year all love learning, at least most of the time. At our school, 97.3% of the students receive free or reduced price lunch. Of the 560 students, 97.3% are minority students. The school has a vibrant community that loves to get together and celebrate. Around Halloween there is a whole school parade to show off the beautiful costumes that students wear. On Cinco de Mayo we put on a big festival with crafts made by the students, dances, and games. At the end of the year the school hosts a carnival to celebrate the hard work put in during the school year, with a dunk tank being the most popular activity. My students will use these five brightly colored Hokki stools in place of regular, stationary, 4-legged chairs. As I will only have a total of ten in the classroom and not enough for each student to have an individual one, they will be used in a variety of ways. During independent reading time they will be used as special chairs students will each use on occasion. I will utilize them in place of chairs at my small group tables during math and reading times. The rest of the day they will be used by the students who need the highest amount of movement in their life in order to stay focused on school. Wherever asked what the classroom is missing, my students always say more Hokki Stools. They can't get their fill of the 5 stools we already have. When the students are sitting in group with me on the Hokki Stools, they are always moving, but at the same time doing their work. Anytime the students get to pick where they can sit, the Hokki Stools are the first to be taken. There are always students who head over to the kidney table to get one of the stools who are disappointed as there are not enough of them. We ask a lot of students to sit for 7 hours a day. The Hokki stools will be a compromise that allow my students to do desk work and move at the same time. These stools will help students to meet their 60 minutes a day of movement by allowing them to activate their core muscles for balance while they sit. For many of my students, these chairs will take away the barrier that exists in schools for a child who can't sit still. nannan

Describing my students isn't an easy task. Many would say that they are inspirational, creative, and hard-working. They are all unique - unique in their interests, their learning, their abilities, and so much more. What they all have in common is their desire to learn each day, despite difficulties that they encounter. \r\nOur classroom is amazing - because we understand that everyone learns at their own pace. As the teacher, I pride myself in making sure my students are always engaged, motivated, and inspired to create their own learning! \r\nThis project is to help my students choose se

ating that is more appropriate for them, developmentally. Many students tire of sitting in chairs during lessons, and having different seats available helps to keep them engaged and learning.\r\nFlexible seating is important in our classroom, as many of our students struggle with attention, focus, and engagement. We currently have stability balls for seating, as well as regular chairs, but these stools will help students who have trouble with balance, or find it difficult to sit on a stability ball for a long period of time. We are excited to try these stools as a part of our engaging classroom community!nannan

I am a third grade teacher at Heritage Elementary in Madison Alabama. My students are unique and each possess special qualities that help others learn and grow. My students and I share experiences that help guide them into making choices that are productive and mold them into life long learners. They have an amazing ability to help others achieve their highest academic potential while ability to help children achieve their best\r\n\r\nI strive to inspire my students not only academically but personally. The kids' energies; their inquisitiveness makes inspires teaching them and pushing them harder\r\n\r\n\r\nFlexible Seating and Student-Centered Classroom Redesign is key in encouraging students to perform at their highest academic potential. My mission is to keep the focus on what's really important: the students. If student motivation and higher engagement is truly the desired end game, then I must adapt right along with my students in my classroom. Our classroom environments should be conducive to open collaboration, communication, creativity, and critical thinking. This simply cannot be done when kids are sitting in rows of desks all day.\r\n\r\n\r\n"Studies on classroom seating suggest that sustained sitting in regular classroom chairs is unhealthy for children's bodies, particularly their backs" (Schilling &Schwartz, 2004, p. 36).\r\n\r\n\r\nAllowing kids some control over where they sit turns them into problem solvers who can identify how they're feeling and choose what works best for them. Kids simply aren't meant to sit still all day long...none of us are..\r\n\r\n\r\nLastly, these key benefits are essential in promoting a healthy and safe learning environment that can be enjoyed by ALL students!\r\n\r\n\r\n1. promotes students attention spans which results in higher achievement\r\n\r\n2. makes students more actively engaged in the classroom\r\n\r\n3. gives them an active outlet without disrupting their learning\r\n\r\n4. makes them more physically fit\r\n\r\n5. helps those with ADHD and Autism, along with other special needs\r\n\r\n6. helps develop a sense of community among the students which improves their social skills\r\n\r\n7. helps them to become independent learners\r\n\r\n8. is LOVED by the students and teacher\r\n\r\n\r\n\r\nThank you so much for your generous donation!!\r\n\r\nMichele White\r\n\r\n\r\nnnannan

In [57]:

```
# 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 [58]:

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

My kindergarten students have varied disabilities ranging from speech and language delays, cognitive delays, gross/fine motor delays, to autism. They are eager beavers and always strive to work their hardest working past their limitations. \r\n\r\n\r\nThe materials we have are the ones I seek out for my students. I teach in a Title I school where most of the students receive free or reduced price lunch. Despite their disabilities and limitations, my students love coming to school and come eager to learn and explore.Have you ever felt like you had ants in your pants and you needed to groove and move as you were in a meeting? This is how my kids feel all the time. The want to be able to move as they learn or so they say.Wobble chairs are the answer and I love them because they develop their core, which enhances gross motor and in Turn fine motor skills. \r\n\r\nThey also want to learn through games, my kids do not want to sit and do worksheets. They want to learn to count by jumping and playing. Physical engagement is the key to our success. The number toss and color and shape mats can make that happen. My students will forget they are doing work and just have the fun a 6 year old deserves.nannan

In [59]:

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

My kindergarten students have varied disabilities ranging from speech and language delays, cognitive delays, gross/fine motor delays, to autism. They are eager beavers and always strive to work their hardest working past their limitations. The materials we have are the ones I seek out for my students. I teach in a Title I school where most of the students receive free or reduced price lunch. Despite their disabilities and limitations, my students love coming to school and come eager to learn and explore. Have you ever felt like you had ants in your pants and you needed to groove and move as you were in a meeting? This is how my kids feel all the time. They want to be able to move as they learn or so they say. Wobble chairs are the answer and I love them because they develop their core, which enhances gross motor and in turn fine motor skills. They also want to learn through games, my kids do not want to sit and do worksheets. They want to learn to count by jumping and playing. Physical engagement is the key to our success. The number toss and color and shape mats can make that happen. My students will forget they are doing work and just have the fun a 6 year old deserves. nannan

In [60]:

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

My kindergarten students have varied disabilities ranging from speech and language delays cognitive delays gross fine motor delays to autism They are eager beavers and always strive to work their hardest working past their limitations The materials we have are the ones I seek out for my students I teach in a Title I school where most of the students receive free or reduced price lunch Despite their disabilities and limitations my students love coming to school and come eager to learn and explore Have you ever felt like you had ants in your pants and you needed to groove and move as you were in a meeting This is how my kids feel all the time They want to be able to move as they learn or so they say Wobble chairs are the answer and I love them because they develop their core which enhances gross motor and in turn fine motor skills They also want to learn through games my kids do not want to sit and do worksheets They want to learn to count by jumping and playing Physical engagement is the key to our success The number toss and color and shape mats can make that happen My students will forget they are doing work and just have the fun a 6 year old deserves nannan

In [61]:

```
# https://gist.github.com/sebleier/554280
# we are removing the words from the stop words list: 'no', 'nor', 'not'
stopwords= ['i', 'me', 'my', 'myself', 'we', 'our', 'ours', 'ourselves', 'you', "you're", "you've", \
            "you'll", "you'd", 'your', 'yours', 'yourself', 'yourselves', 'he', 'him', 'his', 'himself', \
            'she', "she's", 'her', 'hers', 'herself', 'it', "it's", 'its', 'itself', 'they', 'them', 'their', \
            'theirs', 'themselves', 'what', 'which', 'who', 'whom', 'this', 'that', "that'll", 'these', 'those', \
            'am', 'is', 'are', 'was', 'were', 'be', 'been', 'being', 'have', 'has', 'had', 'having', 'do', 'does', \
            'did', 'doing', 'a', 'an', 'the', 'and', 'but', 'if', 'or', 'because', 'as', 'until', 'while', 'of', \
            'at', 'by', 'for', 'with', 'about', 'against', 'between', 'into', 'through', 'during', 'before', 'after', \
            'above', 'below', 'to', 'from', 'up', 'down', 'in', 'out', 'on', 'off', 'over', 'under', 'again', 'further', \
            'then', 'once', 'here', 'there', 'when', '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', "couldn'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 [62]:

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

100%|██████████| 30000/30000 [00:23<00:00, 1253.52it/s]

In [63]:

```
# after preprocessing
preprocessed_essays[20000]
```

Out[63]:

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

1.3.2 Project title Text

In [64]:

```
# similarly you can preprocess the titles also
preprocessed_title=[]
# tqdm is for printing the status bar
for sentence in tqdm(project_data['project_title'].values):
    sent = decontracted(sentence)
    sent = sent.replace('\\r', ' ')
    sent = sent.replace('\\n', ' ')
    sent = sent.replace('\\t', ' ')
    sent = re.sub('[^A-Za-z0-9]+', ' ', sent)
    # https://gist.github.com/sebleier/554280
    sent = ' '.join(e for e in sent.split() if e not in stopwords)
    preprocessed_title.append(sent.lower().strip())
```

100%|██████████| 30000/30000 [00:01<00:00, 28013.97it/s]

In [65]:

```
# after preprocessing
preprocessed_title[2000]
```

Out[65]:

'steady stools active learning'

1. 4 Preparing data for models

In [66]:

```
project_data.columns
```

Out[66]:

```
Index(['Unnamed: 0', 'id', 'teacher_id', 'teacher_prefix', 'school_state',
      'project_submitted_datetime', 'project_grade_category', 'project_title',
      'project_essay_1', 'project_essay_2', 'project_essay_3',
      'project_essay_4', 'project_resource_summary',
      'teacher_number_of_previously_posted_projects', 'project_is_approved',
      'clean_categories', 'clean_subcategories', 'essay', 'price', 'quantity',
      'digit_in_summary'],
      dtype='object')
```

we are going to consider

- school_state : categorical data
- clean_categories : categorical data
- clean_subcategories : categorical data
- project_grade_category : categorical data
- teacher_prefix : categorical data
- project_title : text data
- text : text data
- project_resource_summary: text data
- quantity : numerical
- teacher_number_of_previously_posted_projects : numerical
- price : numerical

1.4.1 Vectorizing Categorical data

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

CLEAN CATEGORIES

In [67]:

```
# 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_one_hot.shape)

['Warmth', 'Care_Hunger', 'History_Civics', 'Music_Arts', 'AppliedLearning', 'SpecialNeeds', 'Health_Sports', 'Math_Science', 'Literacy_Language']
Shape of matrix after one hot encoding (30000, 9)
```

CLEAN SUBCATEGORIES

In [68]:

```
# we use count vectorizer to convert the values into one hot encoded features
vectorizer = CountVectorizer(vocabulary=list(sorted_sub_cat_dict.keys()), lowercase=False, binary=True)
vectorizer.fit(project_data['clean_subcategories'].values)
print(vectorizer.get_feature_names())

sub_categories_one_hot = vectorizer.transform(project_data['clean_subcategories'].values)
print("Shape of matrix after one hot encoding ", sub_categories_one_hot.shape)

['Economics', 'CommunityService', 'FinancialLiteracy', 'ParentInvolvement', 'Extracurricular', 'Civics_Government', 'ForeignLanguages', 'NutritionEducation', 'Warmth', 'Care_Hunger', 'SocialSciences', 'PerformingArts', 'CharacterEducation', 'TeamSports', 'Other', 'College_CareerPrep', 'Music', 'History_Geography', 'Health_LifeScience', 'EarlyDevelopment', 'ESL', 'Gym_Fitness', 'EnvironmentalScience', 'VisualArts', 'Health_Wellness', 'AppliedSciences', 'SpecialNeeds', 'Literature_Writing', 'Mathematics', 'Literacy']
Shape of matrix after one hot encoding (30000, 30)
```

SCHOOL STATE

In [69]:

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

#State
vectorizer = CountVectorizer(lowercase=False, binary=True)
vectorizer.fit(project_data['school_state'].values)
print(vectorizer.get_feature_names())

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

['AK', 'AL', 'AR', 'AZ', 'CA', 'CO', 'CT', 'DC', 'DE', 'FL', 'GA', 'HI', 'IA', 'ID', 'IL', 'IN', 'KS',
', 'KY', 'LA', 'MA', 'MD', 'ME', 'MI', 'MN', 'MO', 'MS', 'MT', 'NC', 'ND', 'NE', 'NH', 'NJ', 'NM', 'NV',
', 'NY', 'OH', 'OK', 'OR', 'PA', 'RI', 'SC', 'SD', 'TN', 'TX', 'UT', 'VA', 'VT', 'WA', 'WI', 'WV',
', 'WY']
Shape of matrix after one hot encodig  (30000, 51)
```

TEACHER PREFIX

when we are process to remove the nan in teacher prefix the number of data points should be less from the original .

In [70]:

```
'''
#https://datascience.stackexchange.com/questions/30249/how-to-delete-entire-row-if-values-in-a-column-are-nan
project_data.isna().sum() # count NULLs before filtering
project_data = project_data[pd.notnull(project_data['teacher_prefix'])]
project_data.isna().sum() # count NULLs after removing null values from techer_prefix Column
'''
```

Out[70]:

```
Unnamed: 0      0
id              0
teacher_id      0
teacher_prefix  0
school_state    0
project_submitted_datetime  0
project_grade_category  0
project_title   0
project_essay_1  0
project_essay_2  0
project_essay_3 28986
project_essay_4 28986
project_resource_summary  0
teacher_number_of_previously_posted_projects  0
project_is_approved  0
clean_categories  0
clean_subcategories  0
essay            0
price           0
quantity        0
digit_in_summary  0
dtype: int64
```

In [71]:

```
'''
# Please do the similar feature encoding with state, teacher_prefix and project_grade_category also

#Teacher_Prefix
from sklearn.feature_extraction.text import CountVectorizer
vectorizer = CountVectorizer(lowercase=False, binary=True)
vectorizer.fit(project_data['teacher_prefix'].values.astype(str))
print(vectorizer.get_feature_names())
teacher_prefix_one_hot = vectorizer.transform(project_data['teacher_prefix'].values.astype('U'))
print("Shape of matrix after one hot encodig ",teacher_prefix_one_hot.shape)
'''

['Mr', 'Mrs', 'Ms', 'Teacher']
Shape of matrix after one hot encodig  (29999, 4)
```

In [70]:

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

In [71]:

```
teacher_prefix_dict = dict(my_counter)
sorted_teacher_prefix_dict = dict(sorted(teacher_prefix_dict.items(), key=lambda kv: kv[1]))
```

In [72]:

```
# we use count vectorizer to convert the values into one hot encoded features
# for Teacher_Prefix
vectorizer = CountVectorizer(vocabulary=list(sorted_teacher_prefix_dict.keys()), lowercase=False, binary=True)
```

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

```
['nan', 'Teacher', 'Mr.', 'Ms.', 'Mrs.']
Shape of matrix after one hot encoding (30000, 5)
```

PROJECT GRADE

In [73]:

```
project_data['project_grade_category'].value_counts()
```

Out[73]:

```
Grades PreK-2    12204
Grades 3-5       10160
Grades 6-8        4663
Grades 9-12       2973
Name: project_grade_category, dtype: int64
```

In [74]:

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

# https://www.geeksforgeeks.org/removing-stop-words-nltk-python/
# https://stackoverflow.com/questions/23669024/how-to-strip-a-specific-word-from-a-string
# https://stackoverflow.com/questions/8270092/remove-all-whitespace-in-a-string-in-python
grade_list = []
for i in grade_catogories:
    i = i.replace(' ', '_') # we are replacing the " " value into
    i = i.replace('-', '_') # we are replacing the " " value into
    grade_list.append(i)
```

In [75]:

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

In [76]:

```
project_grade_category_dict = dict(my_counter)
sorted_project_grade_category_dict = dict(sorted(project_grade_category_dict.items(), key=lambda kv: kv[1]))
```

In [77]:

```
# we use count vectorizer to convert the values into one hot encoded features  
# for Project_Grade_Category  
vectorizer = CountVectorizer(vocabulary=list(sorted(project_grade_category_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)
```

```
['9-12', '6-8', '3-5', 'PreK-2', 'Grades']  
Shape of matrix after one hot encoding (30000, 5)
```

1.4.2 Vectorizing Text data

1.4.2.1 Bag of words

In [78]:

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

Shape of matrix after one hot encoding (30000, 10006)

1.4.2.2 Bag of Words on `project_title`

In [79]:

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

Shape of matrix after one hot encoding (30000, 1536)

1.4.2.3 TFIDF vectorizer

In [80]:

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

Shape of matrix after one hot encoding (30000, 10006)

1.4.2.4 TFIDF Vectorizer on `project_title`

In [81]:

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

Shape of matrix after one hot encoding (30000, 1536)

1.4.2.5 Using Pretrained Models: Avg W2V

In [82]:

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

# =====
Output:

Loading Glove Model
1917495it [06:32, 4879.69it/s]
Done. 1917495 words loaded!

# =====

words = []
for i in preproced_texts:
    words.extend(i.split(' '))

for i in preproced_titles:
    words.extend(i.split(' '))
print("all the words in the coupus", len(words))
words = set(words)
print("the unique words in the coupus", len(words))

inter_words = set(model.keys()).intersection(words)
print("The number of words that are present in both glove vectors and our coupus", \
      len(inter_words), "(", np.round(len(inter_words)/len(words)*100, 3), "%)")

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

# stronging variables into pickle files python: http://www.jessicayung.com/how-to-use-pickle-to-save-and-load-var
iables-in-python/

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

'''
```

Out[82]:

```
'\n# Reading glove vectors in python: https://stackoverflow.com/a/38230349/4084039\ndef loadGloveMod
el(gloveFile):\n    print ("Loading Glove Model")\n    f = open(gloveFile, \'r\', encoding="utf8")\n
model = {}\n    for line in tqdm(f):\n        splitLine = line.split()\n        word = splitLine[0]\n
    n        embedding = np.array([float(val) for val in splitLine[1:]])\n        model[word] = embeddin
g\n    print ("Done.", len(model), " words loaded!")\n    return model\nmodel = loadGloveModel(\'glove
.42B.300d.txt\')\n\n# =====\nOutput:\n    \nLoading Glove Model\n1917495it [0
6:32, 4879.69it/s]\nDone. 1917495 words loaded!\n\n# =====\n\nwords = []\nfo
r i in preproced_texts:\n    words.extend(i.split(\' \''))\n\nfor i in preproced_titles:\n    words.e
xtend(i.split(\' \''))\n\nprint("all the words in the coupus", len(words))\n\nwords = set(words)\n\nprint("
the unique words in the coupus", len(words))\n\n\ninter_words = set(model.keys()).intersection(words)\
\nprint("The number of words that are present in both glove vectors and our coupus", len(inter_
words), "(", np.round(len(inter_words)/len(words)*100, 3), "%)")\n\nwords_courpus = {}\nwords_glove = se
t(model.keys())\nfor i in words:\n    if i in words_glove:\n        words_courpus[i] = model[i]\n\npri
nt("word 2 vec length", len(words_courpus))\n\n\n# stronging variables into pickle files python: htt
p://www.jessicayung.com/how-to-use-pickle-to-save-and-load-variables-in-python/\n\nimport pickle\nwi
th open(\'glove_vectors\', \'wb\') as f:\n    pickle.dump(words_courpus, f)\n\n\n'
```


In [83]:

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

In [84]:

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

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

100%|██████████| 30000/30000 [00:11<00:00, 2688.65it/s]

30000
300

1.4.2.6 Using Pretrained Models: AVG W2V on `project_title`

In [85]:

```
# Similarly you can vectorize for title also
avg_w2v_title_vectors = []; # the avg-w2v for each sentence/review is stored in this list
for sentence in tqdm(preprocessed_title): # for each review/sentence
    vector = np.zeros(300) # as word vectors are of zero length
    cnt_words = 0; # num of words with a valid vector in the sentence/review
    for word in sentence.split(): # for each word in a review/sentence
        if word in glove_words:
            vector += model[word]
            cnt_words += 1
    if cnt_words != 0:
        vector /= cnt_words
    avg_w2v_title_vectors.append(vector)

print(len(avg_w2v_title_vectors))
print(len(avg_w2v_title_vectors[0]))
```

100%|██████████| 30000/30000 [00:00<00:00, 42047.49it/s]

30000
300

1.4.2.7 Using Pretrained Models: TFIDF weighted W2V

In [86]:

```
# S = ["abc def pqr", "def def def abc", "pqr pqr def"]
tfidf_model = TfidfVectorizer()
tfidf_model.fit(preprocessed_essays)

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

In [87]:

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

print(len(tfidf_w2v_vectors))
print(len(tfidf_w2v_vectors[0]))
```

100%|██████████| 30000/30000 [01:16<00:00, 389.93it/s]

30000

300

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

In [88]:

```
# Similarly you can vectorize for title also
tfidf_w2v_title_vectors = []; # the avg-w2v for each sentence/review is stored in this list
for sentence in tqdm(preprocessed_title): # for each review/sentence
    vector = np.zeros(300) # as word vectors are of zero length
    tf_idf_weight = 0; # num of words with a valid vector in the sentence/review
    for word in sentence.split(): # for each word in a review/sentence
        if (word in glove_words) and (word in tfidf_words):
            vec = model[word] # getting the vector for each word
            # here we are multiplying idf value(dictionary[word]) and the tf value((sentence.count(word)/len(sentence.split())))
            tf_idf = dictionary[word]*(sentence.count(word)/len(sentence.split())) # getting the tfidf value for each word
            vector += (vec * tf_idf) # calculating tfidf weighted w2v
            tf_idf_weight += tf_idf
    if tf_idf_weight != 0:
        vector /= tf_idf_weight
    tfidf_w2v_title_vectors.append(vector)

print(len(tfidf_w2v_title_vectors))
print(len(tfidf_w2v_title_vectors[0]))
```

100%|██████████| 30000/30000 [00:01<00:00, 19308.35it/s]

30000

300

1.4.3 Vectorizing Numerical features

A) PRICE

In [89]:

```
# check this one: https://www.youtube.com/watch?v=0H0q0cIn3Z4&t=530s
# standardization sklearn: https://scikit-learn.org/stable/modules/generated/sklearn.preprocessing.StandardScaler.html
from sklearn.preprocessing import StandardScaler

# price_standardized = standardScaler.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 mean and variance.
price_standardized = price_scalar.transform(project_data['price'].values.reshape(-1, 1))
```

Mean : 298.59731666666666, Standard deviation : 376.0203217475084

In [90]:

```
price_standardized
```

Out[90]:

```
array([[ -0.38295089],
       [  0.00107091],
       [  0.58042789],
       ...,
       [-0.07357931],
       [-0.2623723 ],
       [-0.16126074]])
```

B) TEACHER NUMBER OF PREVIOUSLY POSTED PROJECTS

In [91]:

```
import warnings
warnings.filterwarnings('ignore')
# check this one: https://www.youtube.com/watch?v=0H0q0cIn3Z4&t=530s
# standardization sklearn: https://scikit-learn.org/stable/modules/generated/sklearn.preprocessing.StandardScaler.html
from sklearn.preprocessing import StandardScaler #Column Standardisation

# price_standardized = standardScaler.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)

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

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

Mean : 11.211633333333333, Standard deviation : 27.929644549574125

In [92]:

```
prev_posts_standardized
```

Out[92]:

```
array([[ -0.40142413],
       [-0.15079438],
       [-0.36561988],
       ...,
       [-0.25820713],
       [-0.40142413],
       [-0.15079438]])
```

C) QUANTITY

In [93]:

```
import warnings
warnings.filterwarnings('ignore')
# check this one: https://www.youtube.com/watch?v=0H0q0cln3Z4&t=530s
# standardization sklearn: https://scikit-learn.org/stable/modules/generated/sklearn.preprocessing.StandardScaler.html
from sklearn.preprocessing import StandardScaler #Column Standardisation

# price_standardized = standardScaler.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)

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

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

Mean : 16.936166666666665, Standard deviation : 26.64948076990536

In [94]:

```
quantity_standardized
```

Out[94]:

```
array([[ 0.22754039],
       [-0.59799164],
       [ 0.19001621],
       ...,
       [-0.44789491],
       [-0.48541909],
       [-0.48541909]])
```

1.4.4 Merging all the above features

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

In [95]:

```
print(categories_one_hot.shape)
print(sub_categories_one_hot.shape)
print(text_bow.shape)
print(price_standardized.shape)
```

```
(30000, 9)
(30000, 30)
(30000, 10006)
(30000, 1)
```

In [96]:

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

Out[96]:

```
(30000, 10046)
```

Assignment 2: Apply TSNE

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

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

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

In [97]:

```
#Concatenate all the categorical ,numrical features and project title(BOW)....

Y= hstack((categories_one_hot, sub_categories_one_hot, school_state_one_hot,
project_grade_category_one_hot, teacher_prefix_one_hot, price_standardized,
           quantity_standardized, prev_posts_standardized, title_bow))
Y.shape
```

Out[97]:

(30000, 1639)

In [98]:

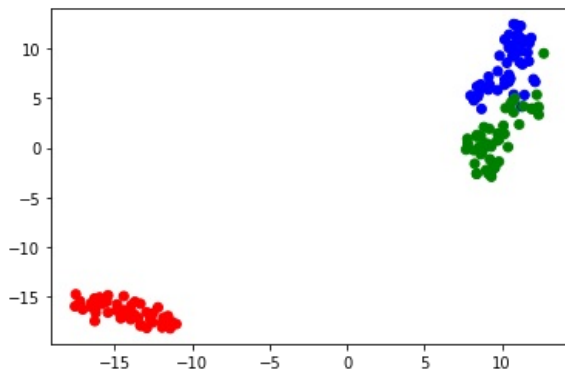
```
# this is the example code for TSNE
import numpy as np
from sklearn.manifold import TSNE
from sklearn import datasets
import pandas as pd
import matplotlib.pyplot as plt

iris = datasets.load_iris()
x = iris['data']
y = iris['target']

tsne = TSNE(n_components=2, perplexity=30, learning_rate=200)

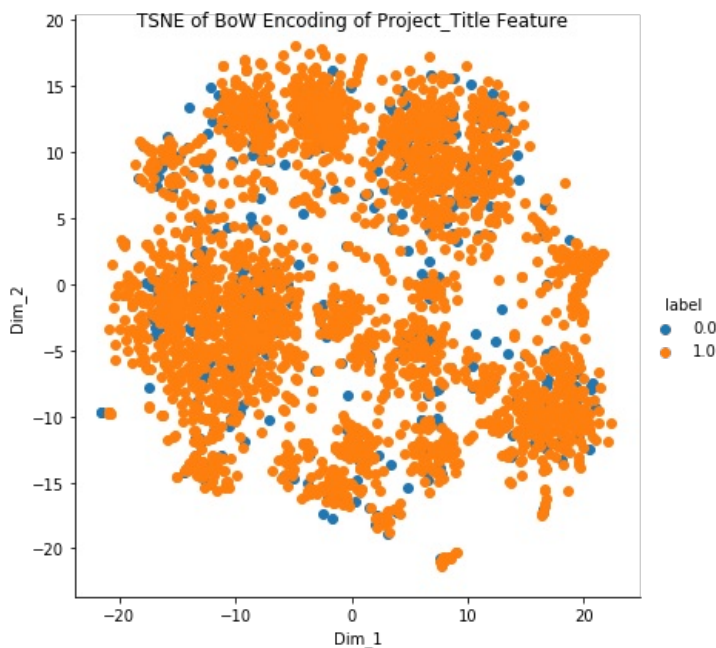
X_embedding = tsne.fit_transform(x)
# if x is a sparse matrix you need to pass it as X_embedding = tsne.fit_transform(x.toarray()) , .toarray() will
convert the sparse matrix into dense matrix

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



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 helpful to the reader
# b. Legends if needed
# c. X-axis label
# d. Y-axis label
from sklearn.manifold import TSNE
Y = Y.tocsr()
data_3000 = Y[0:3000,:]
# for conversion of sparse to dense array
new_3000 = data_3000.toarray()
labels = project_data['project_is_approved'] #The feature we need to plot
labels_3000 = labels[0:3000]
model = TSNE(n_components=2, random_state=0, perplexity=100)
tsne_data = model.fit_transform(new_3000)
#Vertical stacking labels to the tsne_data
tsne_data = np.vstack((tsne_data.T, labels_3000)).T
# Create a new data frame for plotting the result
tsne_df = pd.DataFrame(data=tsne_data, columns=("Dim_1", "Dim_2", "label"))
# Plotting the result of tsne using Seaborn
sns.FacetGrid(tsne_df, hue="label", size=6).map(plt.scatter, 'Dim_1', 'Dim_2').add_legend().fig.suptitle("TSNE of
BoW Encoding of Project_Title Feature ")
plt.show()
```



OBSERVATION :

- Bag_Of_Word is representing the encoding of project title with **blue points shows Acceptance and Orange points shows Rejection.
- Bag_of_word technique is used for converting text into binary vector form.
- To observe the things we can see that many of the points are in small cluster ,orange and blue points are so much dense .

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

In [105]:

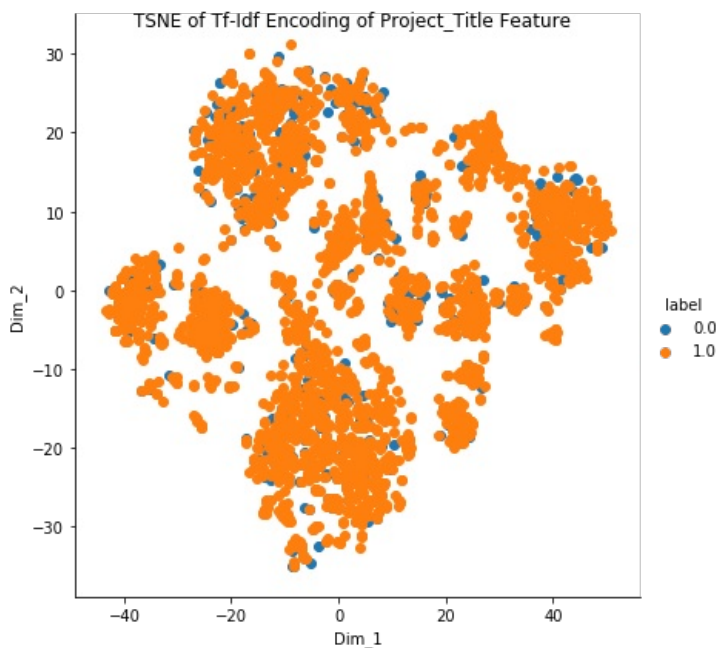
```
#Concatenate all the categorical ,numrical features and project title(tfidf)....
Y= hstack((categories_one_hot, sub_categories_one_hot, school_state_one_hot,
project_grade_category_one_hot, teacher_prefix_one_hot, price_standardized,
           quantity_standardized, prev_posts_standardized, title_tfidf))
Y.shape
```

Out[105]:

(30000, 1639)

In [106]:

```
# please write all the code with proper documentation, and proper titles for each subsection
# when you plot any graph make sure you use
# a. Title, that describes your plot, this will be very helpful to the reader
# b. Legends if needed
# c. X-axis label
# d. Y-axis label
from sklearn.manifold import TSNE
Y = Y.tocsr()
data_3000 = Y[0:3000,:]
# for conversion of sparse to dense array
new_3000 = data_3000.toarray()
labels = project_data['project_is_approved'] #The feature we need to plot
labels_3000 = labels[0:3000]
model = TSNE(n_components=2, random_state=0, perplexity=100)
tsne_data = model.fit_transform(new_3000)
#Vertical stacking labels to the tsne_data
tsne_data = np.vstack((tsne_data.T, labels_3000)).T
# Create a new data frame for plotting the result
tsne_df = pd.DataFrame(data=tsne_data, columns=("Dim_1", "Dim_2", "label"))
# Plotting the result of tsne using Seaborn
sns.FacetGrid(tsne_df, hue="label", size=6).map(plt.scatter, 'Dim_1', 'Dim_2').add_legend().fig.suptitle("TSNE of
Tf-Idf Encoding of Project_Title Feature ")
plt.show()
```



OBSERVATION:

- TFIDF (term frequency inverse document frequency) the words that are rarely occurs in the corpus have high demisionality.
- TFIDF accepts multiple words i.e. n-grams.
- The points are change according to the perplexity ,Points are more dense .

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

In [108]:

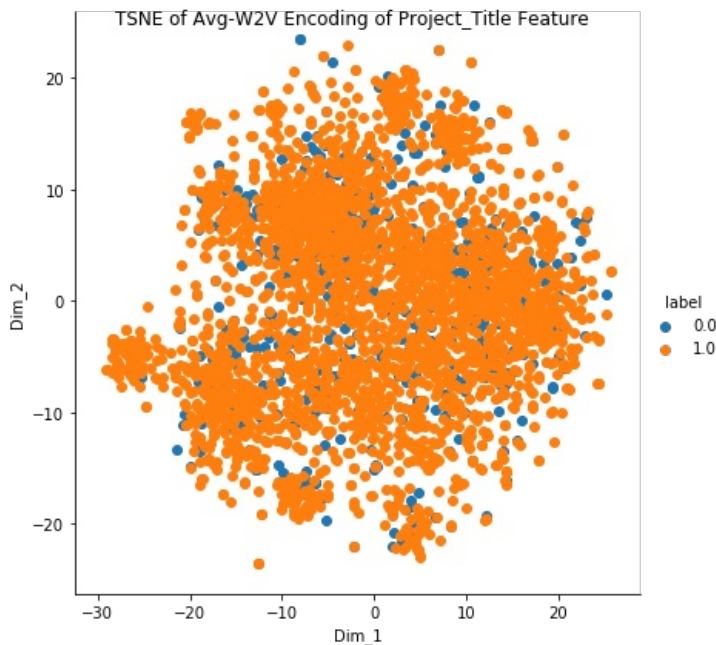
```
#Cancatenate all the categorical ,numrical features and project title(AVG W2V)....
Y= hstack((categories_one_hot, sub_categories_one_hot, school_state_one_hot,
project_grade_category_one_hot, teacher_prefix_one_hot, price_standardized,
           quantity_standardized, prev_posts_standardized, avg_w2v_title_vectors))
Y.shape
```

Out[108]:

(30000, 403)

In [109]:

```
# please write all the code with proper documentation, and proper titles for each subsection
# when you plot any graph make sure you use
# a. Title, that describes your plot, this will be very helpful to the reader
# b. Legends if needed
# c. X-axis label
# d. Y-axis label
from sklearn.manifold import TSNE
Y = Y.tocsr()
data_3000 = Y[0:3000,:]
# for conversion of sparse to dense array
new_3000 = data_3000.toarray()
labels = project_data['project_is_approved'] #The feature we need to plot
labels_3000 = labels[0:3000]
model = TSNE(n_components=2, random_state=0, perplexity=100)
tsne_data = model.fit_transform(new_3000)
#Vertical stacking labels to the tsne_data
tsne_data = np.vstack((tsne_data.T, labels_3000)).T
# Create a new data frame for plotting the result
tsne_df = pd.DataFrame(data=tsne_data, columns=("Dim_1", "Dim_2", "label"))
# Plotting the result of tsne using Seaborn
sns.FacetGrid(tsne_df, hue="label", size=6).map(plt.scatter, 'Dim_1', 'Dim_2').add_legend().fig.suptitle("TSNE of Avg-W2V Encoding of Project_Title Feature ")
plt.show()
```



OBSERVATION:

- To observing ,All the words has same weightage or given a equal number of chances .
- it is not a sparse vector like BOW or TFIDF.
- its works to convert the whole sentence into set of vectors.
- Points are more and more dense according to previous one .

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

In [110]:

```
#Concatenate all the categorical ,numrical features and project title(TFIDF Weighted W2V)....
```

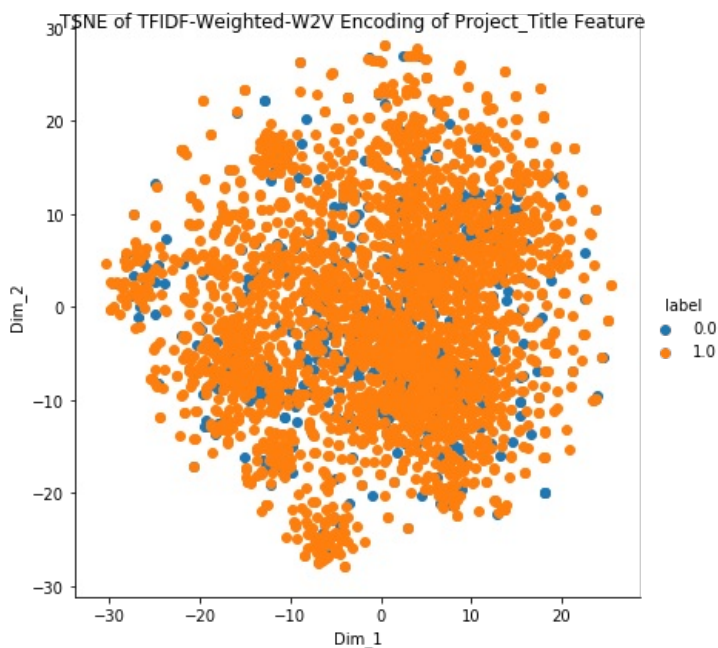
```
Y= hstack((categories_one_hot, sub_categories_one_hot, school_state_one_hot,
project_grade_category_one_hot, teacher_prefix_one_hot, price_standardized,
           quantity_standardized, prev_posts_standardized, tfidf_w2v_title_vectors))
Y.shape
```

Out[110]:

```
(30000, 403)
```

In [111]:

```
# please write all the code with proper documentation, and proper titles for each subsection
# when you plot any graph make sure you use
# a. Title, that describes your plot, this will be very helpful to the reader
# b. Legends if needed
# c. X-axis label
# d. Y-axis label
from sklearn.manifold import TSNE
Y = Y.tocsr()
data_3000 = Y[0:3000,:]
# for conversion of sparse to dense array
new_3000 = data_3000.toarray()
labels = project_data['project_is_approved'] #The feature we need to plot
labels_3000 = labels[0:3000]
model = TSNE(n_components=2, random_state=0, perplexity=100)
tsne_data = model.fit_transform(new_3000)
#Vertical stacking labels to the tsne_data
tsne_data = np.vstack((tsne_data.T, labels_3000)).T
# Create a new data frame for plotting the result
tsne_df = pd.DataFrame(data=tsne_data, columns=("Dim_1", "Dim_2", "label"))
# Plotting the result of tsne using Seaborn
sns.FacetGrid(tsne_df, hue="label", size=6).map(plt.scatter, 'Dim_1', 'Dim_2').add_legend().fig.suptitle("TSNE of
TFIDF-Weighted-W2V Encoding of Project_Title Feature ")
plt.show()
```



OBSERVATION:

- TFIDF Weighted W2V is find the nearest words using similarity function of pre-trained word embedding.
- word embedding is very useful technique .
- AVG word to vec and TFIDF weighed W2V both are similary same ,the task is different that it is also calculating each word weightage.

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

In [112]:

```
Y= hstack((categories_one_hot, sub_categories_one_hot, school_state_one_hot,
project_grade_category_one_hot, teacher_prefix_one_hot, price_standardized,
           quantity_standardized, prev_posts_standardized, title_bow,title_tfidf,avg_w2v_title_vectors,tfidf_w2v_
title_vectors))
Y.shape
```

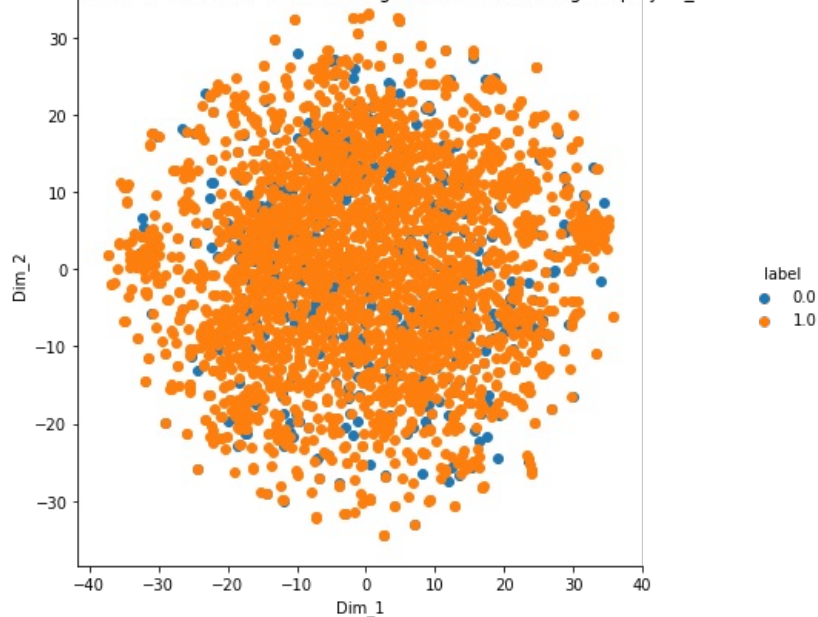
Out[112]:

(30000, 3775)

In [113]:

```
from sklearn.manifold import TSNE
Y = Y.tocsr()
data_3000 = Y[0:3000,:]
# for conversion of sparse to dense array
new_3000 = data_3000.toarray()
labels = project_data['project_is_approved'] #The feature we need to plot
labels_3000 = labels[0:3000]
model = TSNE(n_components=2, random_state=0, perplexity=100)
tsne_data = model.fit_transform(new_3000)
#Vertical stacking labels to the tsne_data
tsne_data = np.vstack((tsne_data.T, labels_3000)).T
# Create a new data frame for plotting the result
tsne_df = pd.DataFrame(data=tsne_data, columns=("Dim_1", "Dim_2", "label"))
# Plotting the result of tsne using Seaborn
sns.FacetGrid(tsne_df, hue="label", size=6).map(plt.scatter, 'Dim_1', 'Dim_2').add_legend().fig.suptitle(" TSNE with `BOW`, `TFIDF`, `AVG W2V`, `TFIDF Weighted W2V` encoding of `project_title` feature ")
plt.show()
```

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



OBSERVATION:

- Due to vital overlapping of points, the visualisation of TSNE with Bag_of_words, TF-IDF, Avg Word2Vec, TF-IDF Weighted Word2Vec does not seem to yield the expected result .
- Similar points are not forming any clusters.
- Nothing much could be concluded out of the TSNE.

2.5 Summary

- we did so many analysis which we can see that either the project is going to be approved or rejected.
- we observe that every state has the lowest rate of acceptance is 80% so we can say that every state which has greater than 80% is success rate in approval.
- Using T-SNE for the dimension reduction we did different parameter and got the result that **Avg W2V && TF-IDF W2V** is better than **Bow & TF-IDF** .
- Rest all the things i have explained it in the obseravtion .