

2_DonorsChoose_EDA_TSNE

February 5, 2019

1 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,

How to scale current manual processes and resources to screen 500,000 projects so that they can
How to increase the consistency of project vetting across different volunteers to improve +
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.

1.1 About the DonorsChoose Data Set

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

Feature	Description
project_id	A unique identifier for the proposed project. Example: p036502

project_title | Title of the project. **Examples:**

Art Will Make You Happy!

First Grade Fun

project_grade_category | Grade level of students for which the project is targeted. One of the following enumerated values:

Grades PreK-2

Grades 3-5

Grades 6-8

Grades 9-12

project_subject_categories | One or more (comma-separated) subject categories for the project from the following enumerated list of values:

Applied Learning

Care & Hunger
Health & Sports
History & Civics
Literacy & Language
Math & Science
Music & The Arts
Special Needs
Warmth

Examples:

Music & The Arts

Literacy & Language, Math & Science

school_state | State where school is located ([Two-letter U.S. postal code](#)). **Example:** WY
project_subject_subcategories | One or more (comma-separated) subject subcategories for the project. **Examples:**

Literacy

Literature & Writing, Social Sciences

project_resource_summary | An explanation of the resources needed for the project. **Example:**

My students need hands on literacy materials to manage sensory needs!

project_essay_1 | First application essay

project_essay_2 | Second application essay **project_essay_3** | Third application essay

project_essay_4 | Fourth application essay **project_submitted_datetime** | Datetime when project application was submitted. **Example:** 2016-04-28 12:43:56.245

teacher_id | A unique identifier for the teacher of the proposed project. **Example:** bdf8baa8fedef6bfeec7ae4ff1c15c56

teacher_prefix | Teacher's title. One of the following enumerated values:

nan

Dr.

Mr.

Mrs.

Ms.

Teacher.

teacher_number_of_previously_posted_projects | Number of project applications previously submitted by the same teacher. **Example:** 2

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

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

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

Feature	Description
<code>quantity</code>	Quantity of the resource required. Example: 3
<code>price</code>	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
<code>project_is_approved</code>	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.

1.1.1 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

```

1.2 1.1 Reading Data

```
In [2]: project_data = pd.read_csv('train_data.csv')
resource_data = pd.read_csv('resources.csv')
```

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

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

The attributes of data : ['Unnamed: 0' 'id' 'teacher_id' 'teacher_prefix' 'school_state'
 'project_submitted_datetime' 'project_grade_category'
 'project_subject_categories' 'project_subject_subcategories'

```
'project_title' 'project_essay_1' 'project_essay_2' 'project_essay_3'  
'project_essay_4' 'project_resource_summary'  
'teacher_number_of_previously_posted_projects' 'project_is_approved']
```

```
In [4]: print("Number of data points in train data", resource_data.shape)  
print(resource_data.columns.values)  
resource_data.head(2)
```

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

```
Out[4]:      id                      description  quantity  \  
0  p233245  LC652 - Lakeshore Double-Space Mobile Drying Rack      1  
1  p069063          Bouncy Bands for Desks (Blue support pipes)      3  
  
      price  
0   149.00  
1   14.95
```

2 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-g
```

```
y_value_counts = project_data['project_is_approved'].value_counts()  
print("Number of projects thar are approved for funding ", y_value_counts[1], ", (", (y_value_counts[1]) * 100 / len(project_data), "%)")  
print("Number of projects thar are not approved for funding ", y_value_counts[0], ", (", (y_value_counts[0]) * 100 / len(project_data), "%)")  
  
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(arrowsize=15,  
bbox=bbox_props, zorder=0, va="center"))  
  
for i, p in enumerate(wedges):  
    ang = (p.theta2 - p.theta1)/2. + p.theta1  
    y = np.sin(np.deg2rad(ang))  
    x = np.cos(np.deg2rad(ang))  
    horizontalalignment = {-1: "right", 1: "left"}[int(np.sign(x))]  
    connectionstyle = "angle,angleA=0,angleB={}{}".format(ang)
```

```

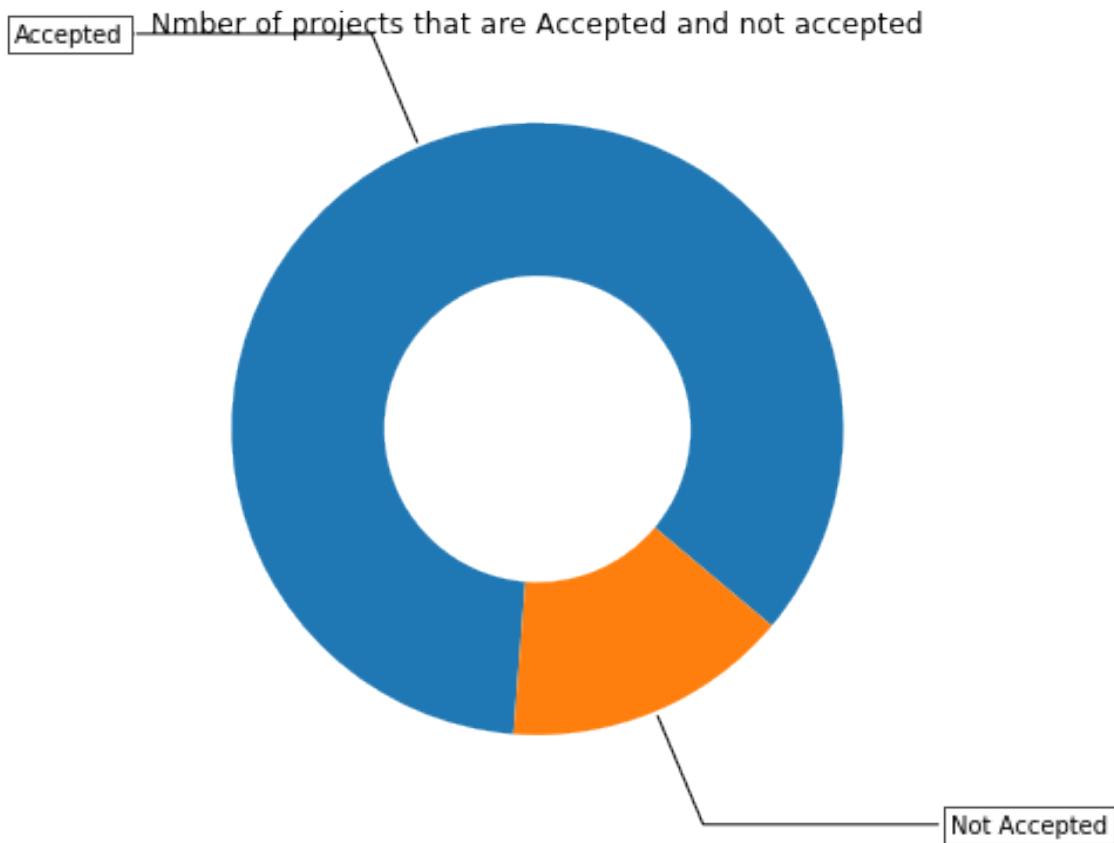
kw["arrowprops"].update({"connectionstyle": connectionstyle})
ax.annotate(recipe[i], xy=(x, y), xytext=(1.35*np.sign(x), 1.4*y),
           horizontalalignment=horizontalalignment, **kw)

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

plt.show()

```

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



2.0.1 1.2.1 Univariate Analysis: School State

In [7]: # 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))
# if you have data which contain only 0 and 1, then the mean = percentage (think about it)
temp.columns = ['state_code', 'num_proposals']

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

```

```

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

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

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

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

Out[7]: '# How to plot US state heatmap: <https://datascience.stackexchange.com/a/9620>\n\nscl =

In [8]: # <https://www.csi.cuny.edu/sites/default/files/pdf/administration/ops/2letterstabrev.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))

	state_code	num_proposals
46	VT	0.800000
7	DC	0.802326
43	TX	0.813142
26	MT	0.816327
18	LA	0.831245

```

States with highest % approvals
   state_code  num_proposals
30          NH      0.873563
35          OH      0.875152
47          WA      0.876178
28          ND      0.888112
 8          DE      0.897959

```

As we can see %approval is different for different states hence state code is an important feature

```

In [9]: #stacked bar plots matplotlib: https://matplotlib.org/gallery/lines_bars_and_markers/b
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 approved vs rejected')
    plt.xticks(ind, list(data[xtick].values))
    plt.legend((p1[0], p2[0]), ('total', 'accepted'))
    plt.show()

In [10]: def univariate_barplots(data, col1, col2='project_is_approved', top=False):
    # Count number of zeros in dataframe python: https://stackoverflow.com/a/51540521
    temp = pd.DataFrame(project_data.groupby(col1)[col2].agg(lambda x: x.eq(1).sum()))

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

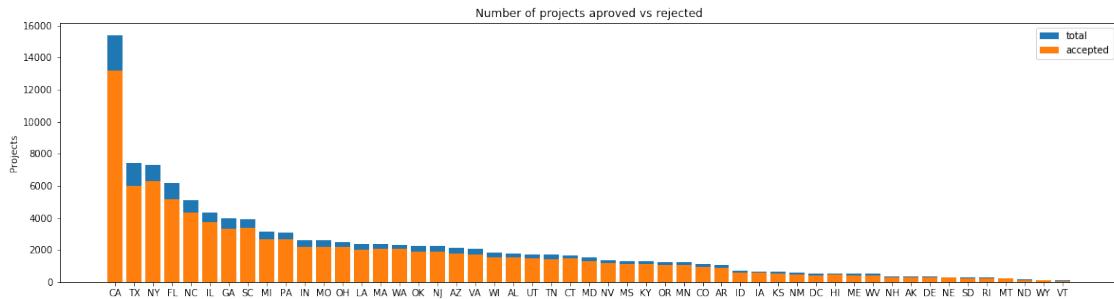
    temp.sort_values(by=['total'], inplace=True, ascending=False)

    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 [11]: univariate_barplots(project_data, 'school_state', 'project_is_approved', False)

```

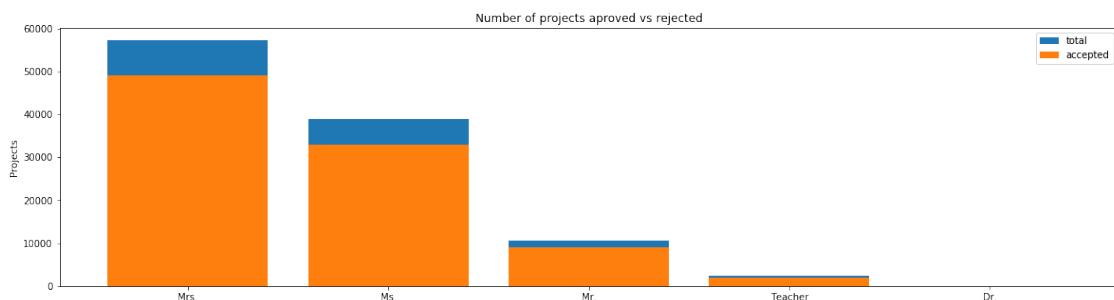


	school_state	project_is_approved	total	Avg
4	CA		13205	0.858136
43	TX		6014	0.813142
34	NY		6291	0.859661
9	FL		5144	0.831690
27	NC		4353	0.855038
<hr/>				
	school_state	project_is_approved	total	Avg
39	RI		243	0.852632
26	MT		200	0.816327
28	ND		127	0.888112
50	WY		82	0.836735
46	VT		64	0.800000

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

2.0.2 1.2.2 Univariate Analysis: teacher_prefix

In [12]: `univariate_barplots(project_data, 'teacher_prefix', 'project_is_approved', top=False)`



	teacher_prefix	project_is_approved	total	Avg
2	Mrs.		48997	0.855559
3	Ms.		32860	0.843537

```

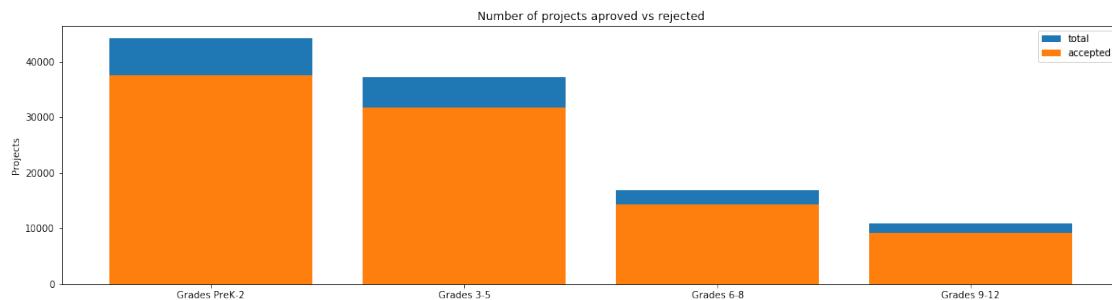
1           Mr.          8960  10648  0.841473
4         Teacher        1877   2360  0.795339
0           Dr.           9     13  0.692308
=====
teacher_prefix  project_is_approved  total      Avg
2           Mrs.          48997  57269  0.855559
3           Ms.          32860  38955  0.843537
1     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

```

If Teacher is either Mrs., Ms. or Mr. then chances of project approval is high

2.0.3 1.2.3 Univariate Analysis: project_grade_category

In [13]: `univariate_barplots(project_data, 'project_grade_category', 'project_is_approved', top=4)`



```

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

```

projects are posted mostly for Grades PreK-2 and Grades 3-5

2.0.4 1.2.4 Univariate Analysis: project_subject_categories

In [15]: `categories = list(project_data['project_subject_categories'].values)`
`# remove special characters from list of strings python: https://stackoverflow.com/a/`

```

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

```

In [16]: project_data['clean_categories'] = cat_list
project_data.drop(['project_subject_categories'], axis=1, inplace=True)
project_data.head(2)

Out[16]:

	Unnamed: 0	id	teacher_id	teacher_prefix
0	160221	p253737	c90749f5d961ff158d4b4d1e7dc665fc	Mrs.
1	140945	p258326	897464ce9ddc600bcfd1151f324dd63a	Mr.

	school_state	project_submitted_datetime	project_grade_category
0	IN	2016-12-05 13:43:57	Grades PreK-2
1	FL	2016-10-25 09:22:10	Grades 6-8

	project_subject_subcategories
0	ESL, Literacy
1	Civics & Government, Team Sports

	project_title
0	Educational Support for English Learners at Home
1	Wanted: Projector for Hungry Learners

	project_essay_1
0	My students are English learners that are work...
1	Our students arrive to our school eager to lea...

	project_essay_2	project_essay_3
0	\"The limits of your language are the limits o...	NaN
1	The projector we need for our school is very c...	NaN

	project_essay_4	project_resource_summary
0	NaN My students need opportunities to practice begin...	
1	NaN My students need a projector to help with view...	

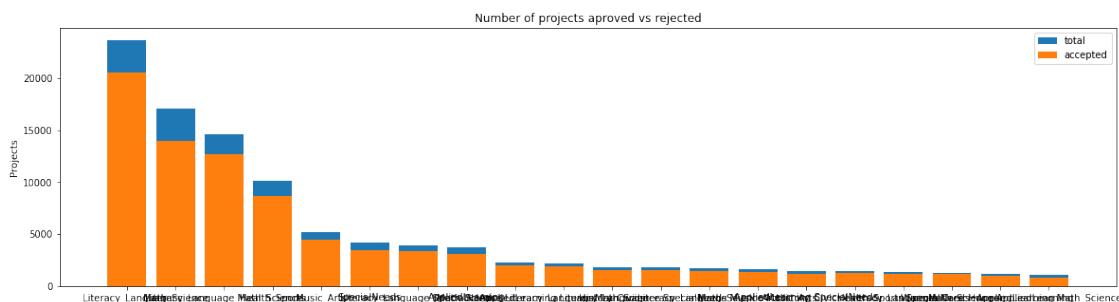
```

teacher_number_of_previously_posted_projects project_is_approved \
0 0 0
1 7 1

clean_categories
0 Literacy_Language
1 History_Civics Health_Sports

```

```
In [18]: univariate_barplots(project_data, 'clean_categories', 'project_is_approved', top=20)
```



	clean_categories	project_is_approved	total	Avg
24	Literacy_Language	20520	23655	0.867470
32	Math_Science	13991	17072	0.819529
28	Literacy_Language	12725	14636	0.869432
8	Health_Sports	8640	10177	0.848973
40	Music_Arts	4429	5180	0.855019
=====				
	clean_categories	project_is_approved	total	Avg
19	History_Civics	1271	1421	0.894444
14	Health_Sports	1215	1391	0.873473
50	Warmth	1212	1309	0.925833
33	Math_Science	1019	1220	0.835222
4	AppliedLearning	855	1052	0.812731

There is lot of variability in %project approval per clean categories. If the project fall under Warm care_hunger then will be high chances of project approval

```
In [19]: # count of all the words in corpus python: https://stackoverflow.com/a/22898595/40840
from collections import Counter
my_counter = Counter()
for word in project_data['clean_categories'].values:
    my_counter.update(word.split())
```

```
In [20]: # dict sort by value python: https://stackoverflow.com/a/613218/4084039  
cat dict = dict(my counter)
```

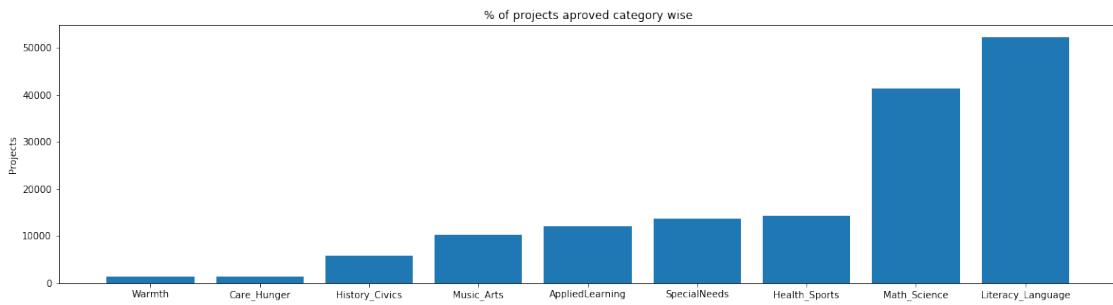
```

sorted_cat_dict = dict(sorted(cat_dict.items(), key=lambda kv: kv[1]))

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

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

```



Lot of Projects which approved are mainly from math science and Literacy language category

```
In [22]: for i, j in sorted_cat_dict.items():
    print("{:20} :{:10}".format(i,j))
```

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

2.0.5 1.2.5 Univariate Analysis: project_subject_subcategories

```
In [23]: sub_catogories = list(project_data['project_subject_subcategories'].values)
# remove special characters from list of strings python: https://stackoverflow.com/a/

# https://www.geeksforgeeks.org/removing-stop-words-nltk-python/
# https://stackoverflow.com/questions/23669024/how-to-strip-a-specific-word-from-a-string-in-python
# 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", "Warm
        if 'The' in j.split(): # this will split each of the category based on space
            j=j.replace('The','') # if we have the words "The" we are going to replace
        j = j.replace(' ','') # we are placeing all the ' '(space) with ''(empty) ex:
        temp +=j.strip()+" "# abc ".strip() will return "abc", remove the trailing sp
        temp = temp.replace('&','_')
    sub_cat_list.append(temp.strip())

```

In [24]: project_data['clean_subcategories'] = sub_cat_list
project_data.drop(['project_subject_subcategories'], axis=1, inplace=True)
project_data.head(2)

Out[24]:

	Unnamed: 0	id	teacher_id	teacher_prefix
0	160221	p253737	c90749f5d961ff158d4b4d1e7dc665fc	Mrs.
1	140945	p258326	897464ce9ddc600bcfd1151f324dd63a	Mr.

	school_state	project_submitted_datetime	project_grade_category	project_title
0	IN	2016-12-05 13:43:57	Grades PreK-2	Educational Support for English Learners at Home
1	FL	2016-10-25 09:22:10	Grades 6-8	Wanted: Projector for Hungry Learners

	project_essay_1
0	My students are English learners that are work...
1	Our students arrive to our school eager to lea...

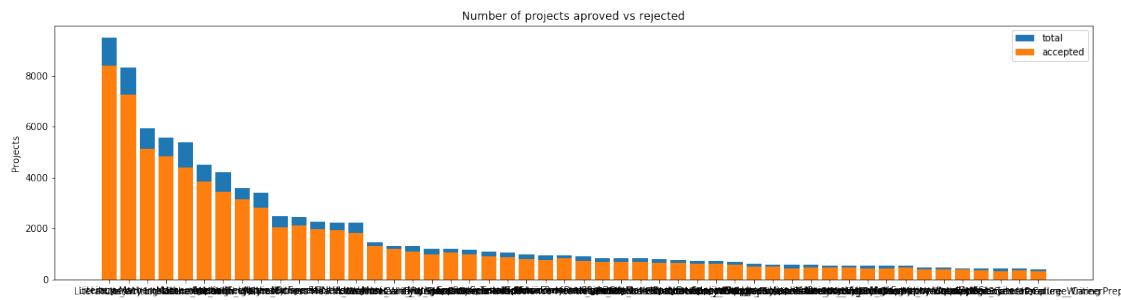
	project_essay_2	project_essay_3
0	\"The limits of your language are the limits o...	Nan
1	The projector we need for our school is very c...	Nan

	project_essay_4	project_resource_summary
0	NaN My students need opportunities to practice beg...	
1	NaN My students need a projector to help with view...	

	teacher_number_of_previously_posted_projects	project_is_approved
0	0	0
1	7	1

	clean_categories	clean_subcategories
0	Literacy_Language	ESL Literacy
1	History_Civics Health_Sports	Civics_Government TeamSports

```
In [25]: univariate_barplots(project_data, 'clean_subcategories', 'project_is_approved', top=50)
```



```
clean_subcategories  project_is_approved  total      Avg
317          Literacy                  8371  9486  0.882458
319        Literacy Mathematics       7260  8325  0.872072
331 Literature_Writing Mathematics   5140  5923  0.867803
318    Literacy Literature_Writing    4823  5571  0.865733
342          Mathematics                 4385  5379  0.815207
=====
clean_subcategories  project_is_approved  total      Avg
196 EnvironmentalScience Literacy        389   444  0.876126
127                      ESL                349   421  0.828979
79          College_CareerPrep            343   421  0.814727
17 AppliedSciences Literature_Writing   361   420  0.859524
3   AppliedSciences College_CareerPrep   330   405  0.814815
```

In subcategories, Literacy had the max no of project approved also having the higher approval %

```
In [26]: # count of all the words in corpus python: https://stackoverflow.com/a/22898595/408400
```

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

```
In [27]: # dict sort by value python: https://stackoverflow.com/a/613218/4084039
```

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

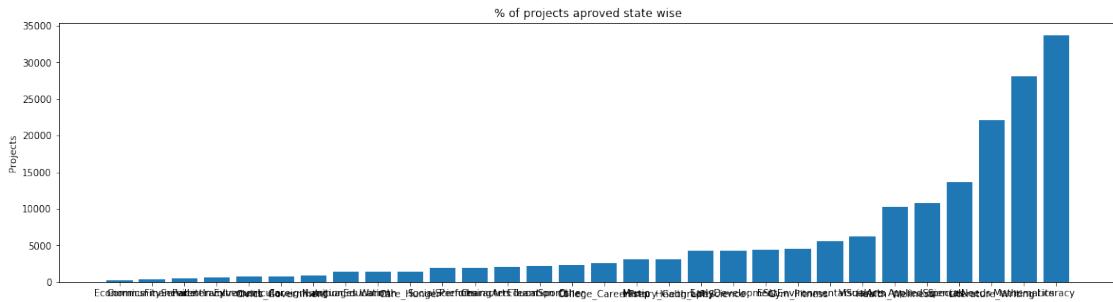
```
ind = np.arange(len(sorted_sub_cat_dict))
plt.figure(figsize=(20,5))
p1 = plt.bar(ind, list(sorted_sub_cat_dict.values()))

plt.ylabel('Projects')
```

```

plt.title('% of projects approved state wise')
plt.xticks(ind, list(sorted_sub_cat_dict.keys()))
plt.show()

```



State wise Literature_Writing had the highest approval with 22179 no of approved project then Mathematics with no of approved project as 28074 and after then we had Literacy with 22179 no of approved project

```
In [28]: for i, j in sorted_sub_cat_dict.items():
    print("{:20} :{:10}".format(i,j))
```

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

```

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

```

2.0.6 1.2.6 Univariate Analysis: Text features (Title)

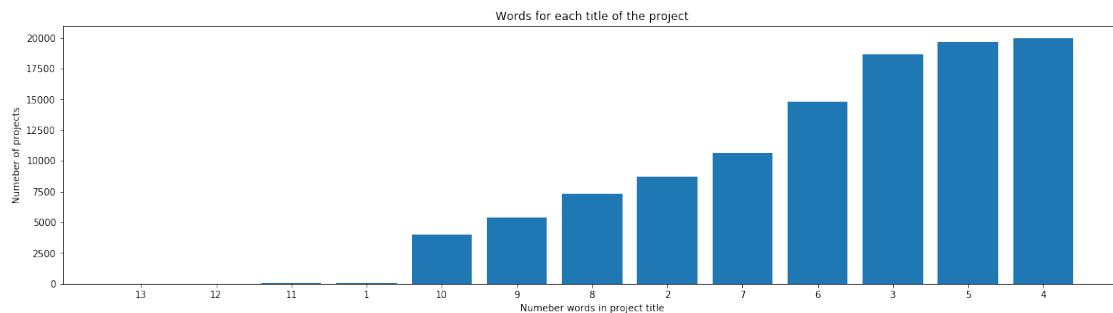
```

In [29]: #How to calculate number of words in a string in DataFrame: https://stackoverflow.com
word_count = project_data['project_title'].str.split().apply(len).value_counts()
word_dict = dict(word_count)
word_dict = dict(sorted(word_dict.items(), key=lambda kv: kv[1]))

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

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

```



Project that have 3-4 words are more in numbers

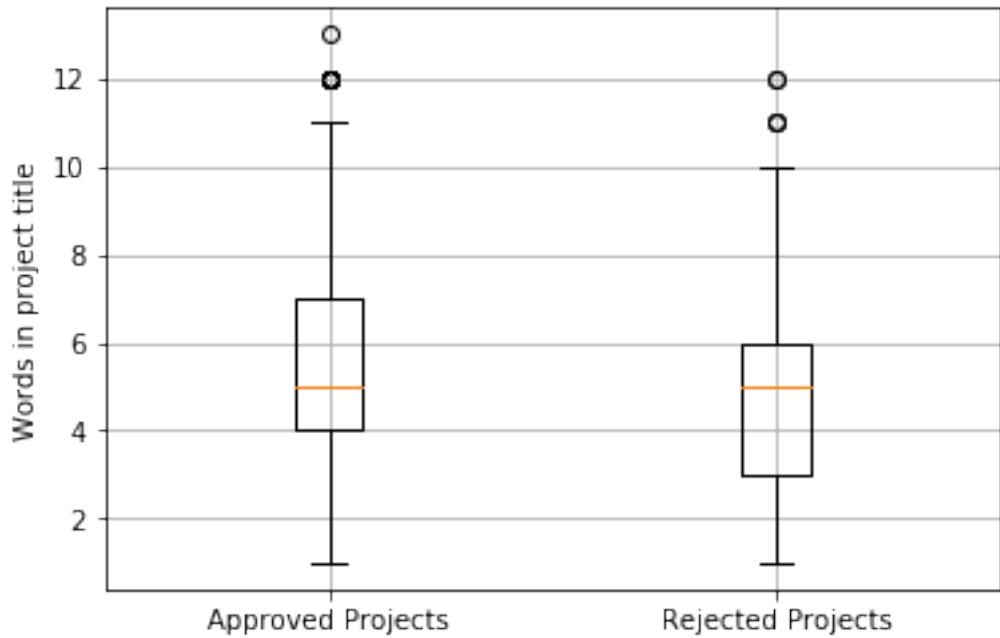
```

In [30]: approved_title_word_count = project_data[project_data['project_is_approved']==1]['project_title_word_count']
approved_title_word_count = approved_title_word_count.values

rejected_title_word_count = project_data[project_data['project_is_approved']==0]['project_title_word_count']
rejected_title_word_count = rejected_title_word_count.values

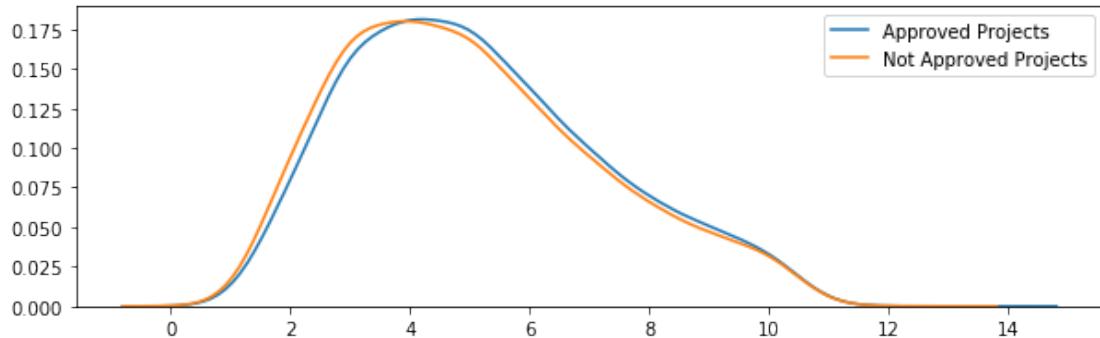
In [32]: # https://glowingpython.blogspot.com/2012/09/boxplot-with-matplotlib.html
plt.boxplot([approved_title_word_count, rejected_title_word_count])
plt.xticks([1,2],('Approved Projects','Rejected Projects'))
plt.ylabel('Words in project title')
plt.grid()
plt.show()

```



If the no. of words in project title is more than 6 then chances of getting approved will be high similarly if the words in project titler is less than 4 then the chances of getting rejected is more.

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



If the word count is below 4 then chances of project getting rejected is sliglthy more and vice versa if word count is above 4

2.0.7 1.2.7 Univariate Analysis: Text features (Project Essay's)

In [34]: # merge two column text dataframe:

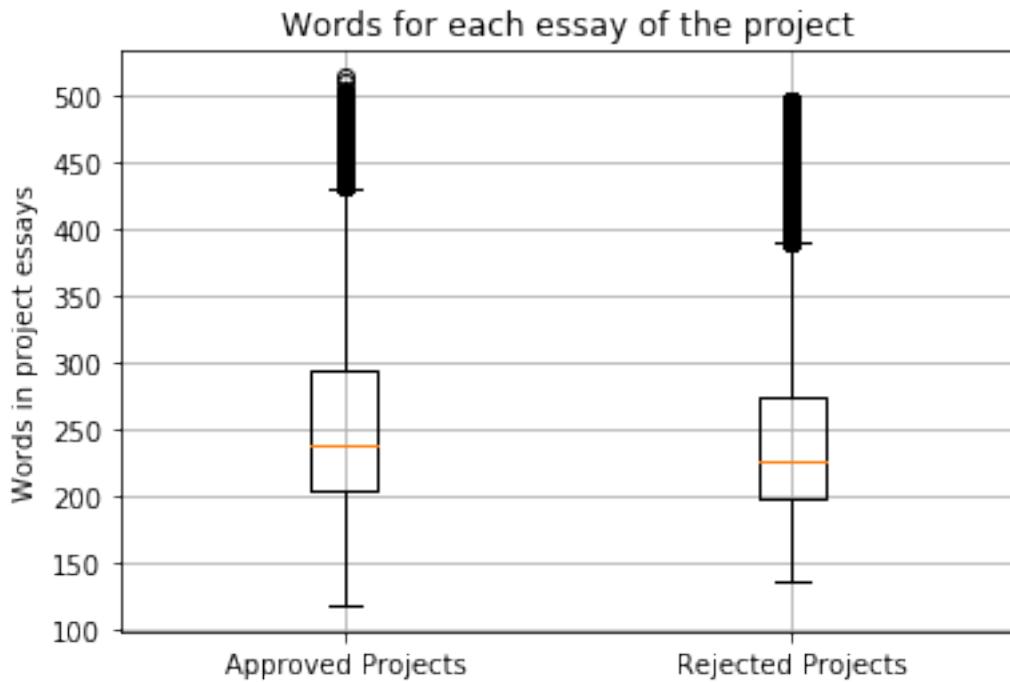
```
project_data["essay"] = project_data["project_essay_1"].map(str) + \
    project_data["project_essay_2"].map(str) + \
    project_data["project_essay_3"].map(str) + \
    project_data["project_essay_4"].map(str)
```

In [35]: approved_word_count = project_data[project_data['project_is_approved']==1]['essay'].str.split().apply(len).sum()
approved_word_count = approved_word_count.values

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

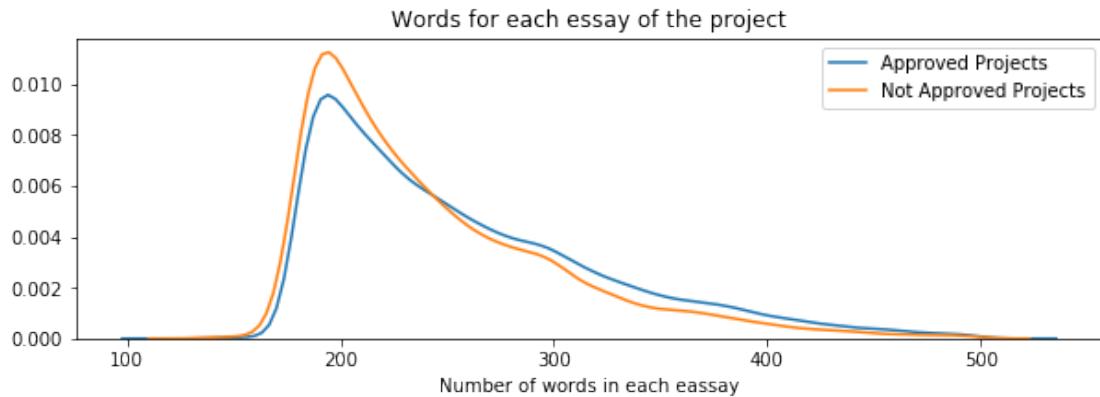
In [36]: # <https://glowingpython.blogspot.com/2012/09/boxplot-with-matplotlib.html>

```
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()
```



If words in project essay more than 275 then chances is more of getting approved but there is also outlier present So to be on safer side we can say if word count of project essay is between 275 to 280 then chances of getting approved is more

```
In [37]: plt.figure(figsize=(10,3))
sns.distplot(approved_word_count, hist=False, label="Approved Projects")
sns.distplot(rejected_word_count, hist=False, label="Not Approved Projects")
plt.title('Words for each essay of the project')
plt.xlabel('Number of words in each eassay')
plt.legend()
plt.show()
```



Project which are rejected has around 170 to 240 word count of project essay and those which are accepted have the word count between 300 to 400

2.0.8 1.2.8 Univariate Analysis: Cost per project

```
In [40]: # we get the cost of the project using resource.csv file
resource_data.head(2)
```

```
Out[40]:      id          description  quantity \
0  p233245  LC652 - Lakeshore Double-Space Mobile Drying Rack      1
1  p069063        Bouncy Bands for Desks (Blue support pipes)      3

      price
0  149.00
1   14.95
```

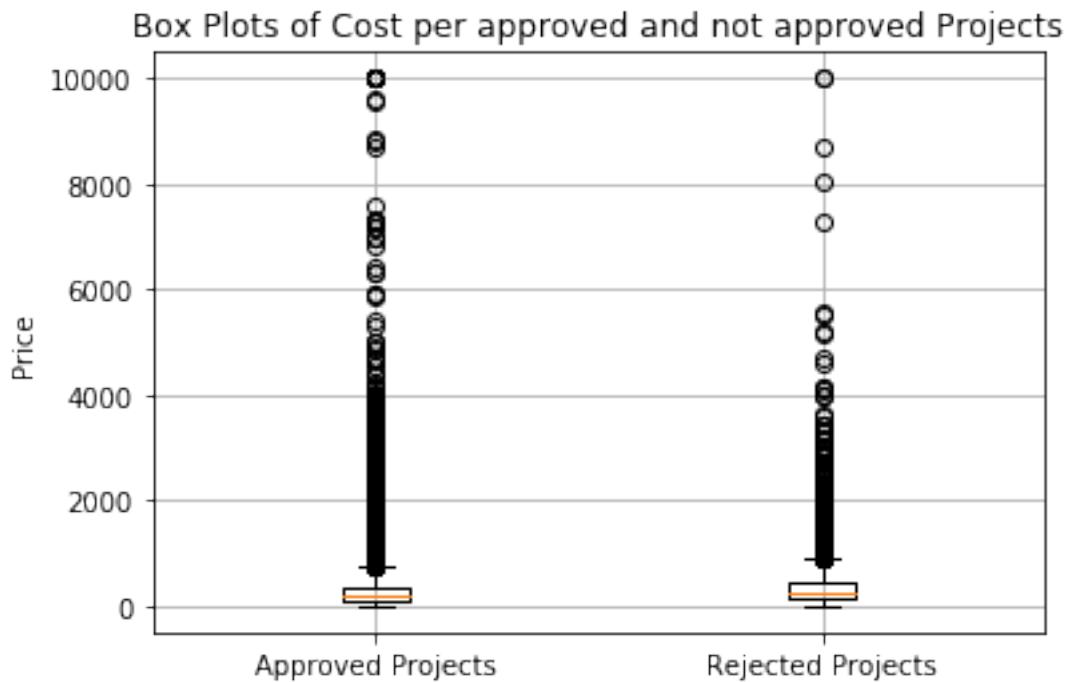
```
In [41]: # https://stackoverflow.com/questions/22407798/how-to-reset-a-dataframes-indexes-for-
price_data = resource_data.groupby('id').agg({'price':'sum', 'quantity':'sum'}).reset_
price_data.head(2)
```

```
Out[41]:      id  price  quantity
0  p000001  459.56        7
1  p000002  515.89       21
```

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

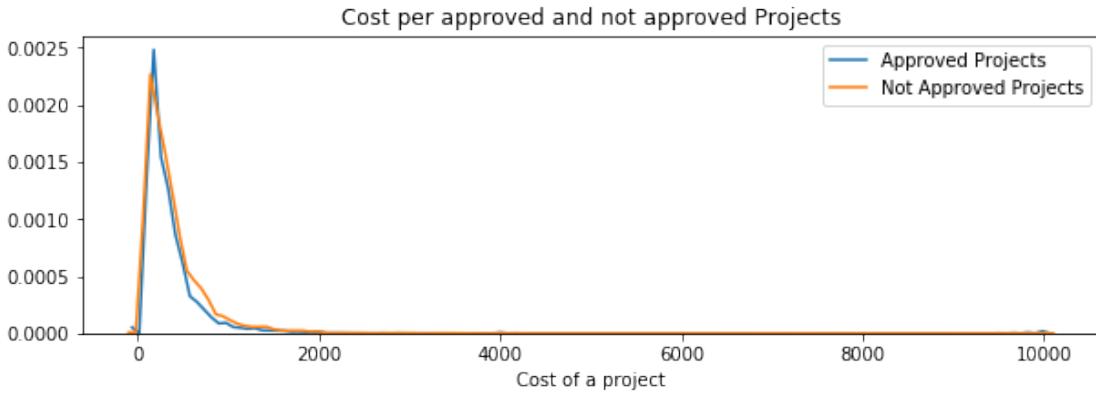
```
In [43]: approved_price = project_data[project_data['project_is_approved']==1]['price'].values
         rejected_price = project_data[project_data['project_is_approved']==0]['price'].values

In [44]: # https://glowingpython.blogspot.com/2012/09/boxplot-with-matplotlib.html
         plt.boxplot([approved_price, rejected_price])
         plt.title('Box Plots of Cost per approved and not approved Projects')
         plt.xticks([1,2],('Approved Projects','Rejected Projects'))
         plt.ylabel('Price')
         plt.grid()
         plt.show()
```



can't say anything

```
In [45]: 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()
```



If project is approved then cost of it will be slightly lesser than the project which are not approved

```
In [46]: # 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(not_approved_price,i), 3)])
print(x)

+-----+-----+-----+
| Percentile | Approved Projects | Not Approved Projects |
+-----+-----+-----+
| 0 | 0.66 | 1.97 |
| 5 | 13.59 | 41.9 |
| 10 | 33.88 | 73.67 |
| 15 | 58.0 | 99.109 |
| 20 | 77.38 | 118.56 |
| 25 | 99.95 | 140.892 |
| 30 | 116.68 | 162.23 |
| 35 | 137.232 | 184.014 |
| 40 | 157.0 | 208.632 |
| 45 | 178.265 | 235.106 |
| 50 | 198.99 | 263.145 |
| 55 | 223.99 | 292.61 |
| 60 | 255.63 | 325.144 |
| 65 | 285.412 | 362.39 |
| 70 | 321.225 | 399.99 |
| 75 | 366.075 | 449.945 |
```

	80		411.67		519.282	
	85		479.0		618.276	
	90		593.11		739.356	
	95		801.598		992.486	
	100		9999.0		9999.0	

Project which are approved are having the lower cost than the cost of Rejected projects

1.2.9 Univariate Analysis: teacher_number_of_previously_posted_projects

2.0.9 Let's see how many data is stored for each teacher who has submitted project proposal

```
In [48]: print("Number of unique teachers = {}".format(project_data['teacher_id'].nunique()))
temp = project_data['teacher_id'].value_counts().sort_values(ascending=False).rename_
temp.head()
```

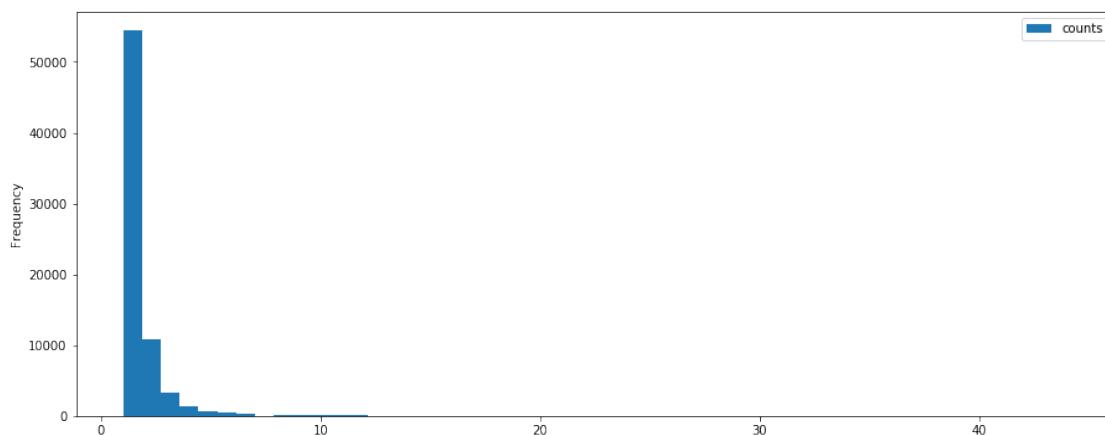
Number of unique teachers = 72168

```
Out[48]:      unique_teacher_id  counts
0  fa2f220b537e8653fb48878ebb38044d    44
1  df8a4b7ad173b57f7ac52e447cc24043    42
2  1f64dcec848be8e95c4482cc845706b2    42
3  7b17c95da53e3d1f011f84232ad01238    34
4  ae67d8bbc64ec3bf7fd2db1297721160    33
```

2.0.10 Graph between number of teachers and no of project posted (according to our dataset)

```
In [49]: temp.plot(kind='hist', bins=50, figsize=(15,6))
print('No. of teacher posted only 1 project is: ', (sum(temp['counts']==1)))
```

No. of teacher posted only 1 project is: 54429



As you can see the no. of teacher posted only one project is more and is equal to 54429

2.0.11 Let's see the history related to previously posted project of teacher with a given id

```
In [50]: project_data[project_data['teacher_id']=='5de88a574b531a9b11b0b63f664d13e9'][['teacher_id', 'project_submitted_datetime', 'project_title', 'teacher_number_of_previously_posted_projects', 'project_is_approved', 'clean_categories']]
```

Out [50]:

teacher_id	project_submitted_datetime	project_title	teacher_number_of_previously_posted_projects	project_is_approved	clean_categories
103386	2016-08-02 11:45:06	Flexible and Healthy seating options!	11	0	Health_Sports
90064	2016-08-04 15:52:32	Tablets to ensure differentiation and financia...	11	0	Math_Science
16253	2016-08-24 21:10:29	Literacy Program for Kinders!	15	0	AppliedLearning Literacy_Language
65426	2016-10-11 08:27:51	Laptop for Centers and Projects!	16	1	Literacy_Language Math_Science
96751	2016-12-19 15:23:32	Snacks for my students!	20	0	Health_Sports

2.0.12 creating a dataframe consisting of teacher with a unique id and the no of projects posted by them

```
In [51]: teacher_data = project_data.copy()
teacher_data.sort_values('teacher_number_of_previously_posted_projects', ascending=False)
teacher_data['Total_number_of_project_submitted_so_far'] = teacher_data['teacher_number_of_previously_posted_projects']
temp1 = teacher_data[['teacher_id', 'Total_number_of_project_submitted_so_far']].sort_values('Total_number_of_project_submitted_so_far', ascending=False)
```

2.0.13 List of 5 teachers who posted Max no of projects:

```
In [52]: temp1.tail(5)[-1]
```

Out [52]:

teacher_id
88015 fa2f220b537e8653fb48878ebb38044d
106026 fa2f220b537e8653fb48878ebb38044d
50805 fa2f220b537e8653fb48878ebb38044d
72233 fa2f220b537e8653fb48878ebb38044d

```
13777    fa2f220b537e8653fb48878ebb38044d
```

```
      Total_number_of_project_submitted_so_far
88015                  452
106026                  438
50805                   434
72233                   433
13777                   429
```

2.0.14 Histogram plot for total no. of project posted

```
In [53]: # Code taken from https://stackoverflow.com/a/6353051/6660373. Modified as per requirement
import matplotlib.pyplot as plt
import numpy as np
from matplotlib.ticker import FormatStrFormatter
binwidth = 50
data = temp1['Total_number_of_project_submitted_so_far'].values
fig, ax = plt.subplots(figsize=(15,6))
counts, bins, patches = ax.hist(data, facecolor='yellow', edgecolor='gray', bins=range(0, 1500, 50))

# Set the ticks to be at the edges of the bins.
ax.set_xticks(bins)
# Set the xaxis's tick labels to be formatted with 1 decimal place...
ax.xaxis.set_major_formatter(FormatStrFormatter('%.0f'))

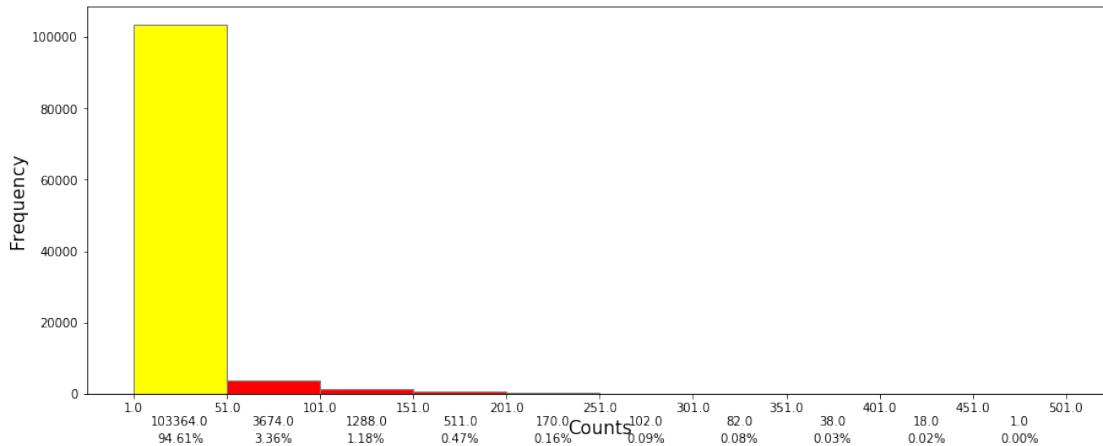
# Change the colors of bars at the edges...
twentyfifth, seventyfifth = np.percentile(data, [25, 75])
for patch, rightside, leftside in zip(patches, bins[1:], bins[:-1]):
    if rightside < twentyfifth:
        patch.set_facecolor('green')
    elif leftside > seventyfifth:
        patch.set_facecolor('red')

# Label the raw counts and the percentages below the x-axis...
bin_centers = 0.5 * np.diff(bins) + bins[:-1]
for count, x in zip(counts, bin_centers):
    # Label the raw counts
    ax.annotate(str(count), xy=(x, 0), xycoords=('data', 'axes fraction'),
               xytext=(0, -18), textcoords='offset points', va='top', ha='center')

    # Label the percentages
    percent = "{0:.2f}%".format(100 *float(count) / counts.sum())
    ax.annotate(percent, xy=(x, 0), xycoords=('data', 'axes fraction'),
               xytext=(0, -32), textcoords='offset points', va='top', ha='center')

ax.set_xlabel('Counts', fontsize=15)
ax.set_ylabel('Frequency', fontsize=15)    # relative to plt.rcParams['font.size']
```

```
# Give ourselves some more room at the bottom of the plot
plt.subplots_adjust(bottom=0.15)
plt.show()
```



94% of the teacher posted project in 1 to 51 numbers

2.0.15 Let's see how many projects are approved for each teacher

```
In [54]: col1 = 'teacher_id'
col2 = 'project_is_approved'
total = 'total_project_present_in_data'
tp = pd.DataFrame(project_data.groupby(col1)[col2].agg(lambda x: x.eq(1).sum())).reset_index()

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

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

print(tp.head(5))
print("=="*50)
print(tp.tail(5))

teacher_id  project_is_approved \
70484    fa2f220b537e8653fb48878ebb38044d      44
8702     1f64dcec848be8e95c4482cc845706b2      40
62925    df8a4b7ad173b57f7ac52e447cc24043      40
34570    7b17c95da53e3d1f011f84232ad01238      34
49149    ae67d8bbc64ec3bf7fd2db1297721160      31

total_project_present_in_data      Avg
70484           44  1.000000
```

```

8702           42  0.952381
62925          42  0.952381
34570          34  1.000000
49149          33  0.939394
=====
              teacher_id  project_is_approved \
27489  61fdee5b0c34ea70671f52b3c3a01b71           1
27490  61ff263134490840c2a7f3b11e21cf0e           1
27491  61ff51943e34dba224a319b49a4d44fa           0
27492  61ff58cafffe3bfef6e929db19e808aeb           1
72167  ffff8e040521f62207881376ecc964d5           1

      total_project_present_in_data   Avg
27489                           1  1.0
27490                           1  1.0
27491                           1  0.0
27492                           1  1.0
72167                           1  1.0

```

Here are the list of 5 teachers with highest no of submission and with lowest no of submission respectively. As you can see some of the teacher has got their all the project accepted

```

In [47]: #univariate_barplots(project_data, 'teacher_id', 'project_is_approved', False)

In [55]: teacher_data.drop(['project_is_approved'], axis = 1, inplace = True, errors = 'ignore')

teacher_proj_data = pd.concat([teacher_data.set_index('teacher_id'),tp.set_index('teacher_id')])
teacher_proj_data[['teacher_id','Total_number_of_project_submitted_so_far','project_is_approved']].sort_values('Total_number_of_project_submitted_so_far', ascending=False).head(5)

Out[55]:
              teacher_id  Total_number_of_project_submitted_so_far \
0  c90749f5d961ff158d4b4d1e7dc665fc                               1
1  897464ce9ddc600bcfd1151f324dd63a                               8
2  3465aa82da834c0582ebd0ef8040ca0                             2
3  f3cb9bffbb169bef1a77b243e620b60                             5
4  be1f7507a41f8479dc06f047086a39ec                             2

      project_is_approved  total_project_present_in_data       Avg
0                      0                           1  0.000000
1                      4                           4  1.000000
2                      2                           3  0.666667
3                      2                           2  1.000000
4                      1                           1  1.000000

```

```

In [56]: approved_proj = project_data[project_data['project_is_approved']==1]['teacher_number_of_projects']

rejected_proj = project_data[project_data['project_is_approved']==0]['teacher_number_of_projects']
# https://glowingpython.blogspot.com/2012/09/boxplot-with-matplotlib.html

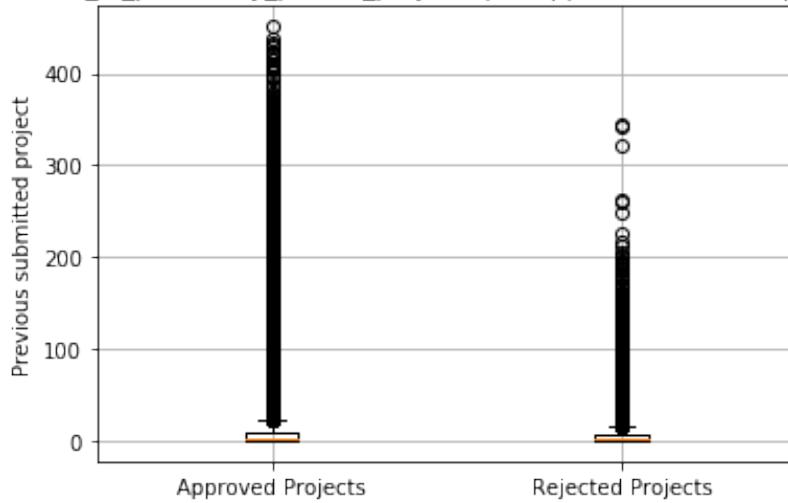
```

```

plt.boxplot([approved_proj, rejected_proj])
plt.title('Box Plots of no_of_previously_posted_projects per approved and not approved')
plt.xticks([1,2],('Approved Projects','Rejected Projects'))
plt.ylabel('Previous submitted project')
plt.grid()
plt.show()

```

Box Plots of no_of_previously_posted_projects per approved and not approved Projects



can't say anything

```

In [57]: plt.figure(figsize=(10,3))
sns.kdeplot(approved_proj,label="Approved Projects", bw=0.6)
sns.kdeplot(rejected_proj,label="Not Approved Projects", bw=0.6)
plt.title('no. of previously posted projects per approved and not approved Projects')
plt.xlabel('no. of previously posted projects')
plt.legend()
plt.show()

```



Project which are accepted have 30 to 90 no. of previously posted project

```
In [58]: # http://zetcode.com/python/prettytable/
from prettytable import PrettyTable

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

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_proj,i), 3), np.round(np.percentile(i
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 | 3.0 |
| 70 | 7.0 | 4.0 |
| 75 | 9.0 | 6.0 |
| 80 | 13.0 | 8.0 |
| 85 | 19.0 | 11.0 |
| 90 | 30.0 | 17.0 |
| 95 | 57.0 | 31.0 |
| 100 | 451.0 | 345.0 |
+-----+-----+-----+
```

Project which are accepted have the no. of previously posted project higher than that of rejected ones

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

1.2.10 Univariate Analysis: project_resource_summary

```
In [59]: #How to calculate number of words in a string in DataFrame: https://stackoverflow.com
word_count = project_data['project_resource_summary'].str.split().apply(len).value_coo
```

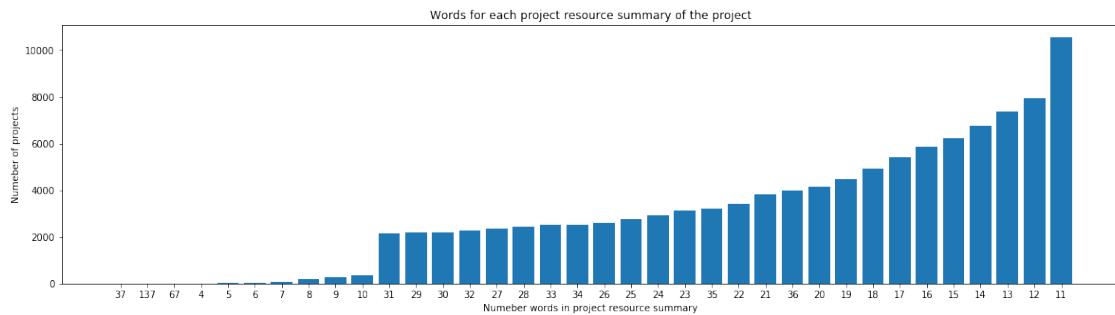
```

word_dict = dict(word_count)
word_dict = dict(sorted(word_dict.items(), key=lambda kv: kv[1]))

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

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

```



Projects which are having more no. of words are larger in numbers.

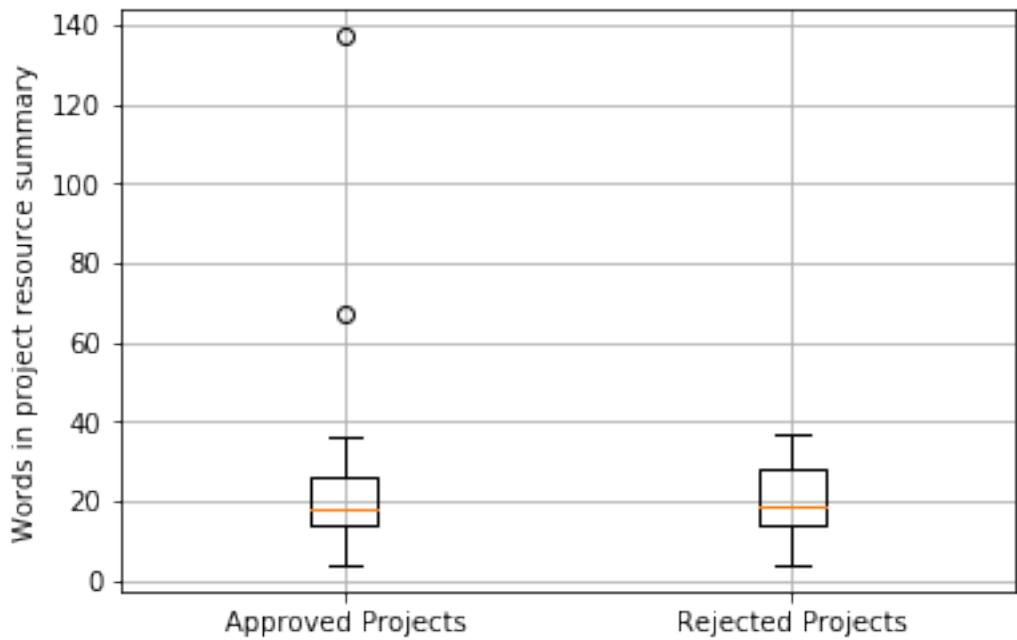
```

In [60]: approved_summary_word_count = project_data[project_data['project_is_approved']==1]['p'
approved_summary_word_count = approved_summary_word_count.values

rejected_summary_word_count = project_data[project_data['project_is_approved']==0]['p'
rejected_summary_word_count = rejected_summary_word_count.values

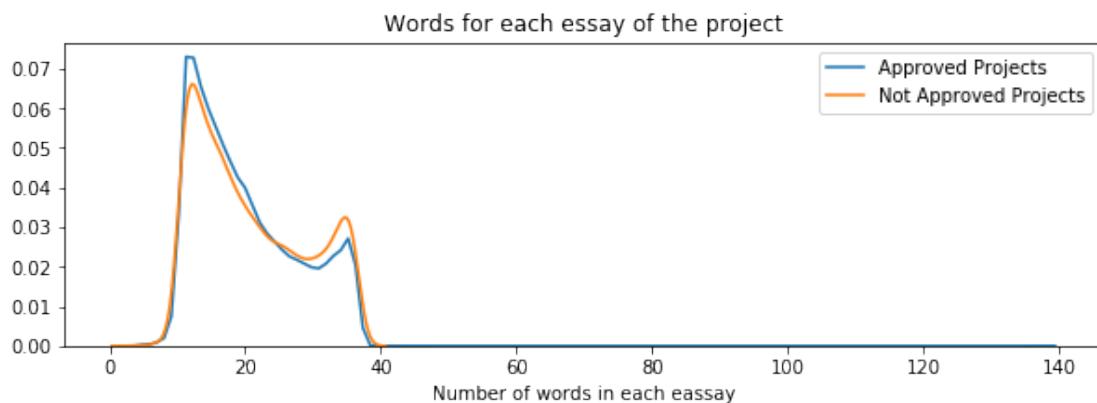
# https://glowingpython.blogspot.com/2012/09/boxplot-with-matplotlib.html
plt.boxplot([approved_summary_word_count, rejected_summary_word_count])
plt.xticks([1,2],('Approved Projects','Rejected Projects'))
plt.ylabel('Words in project resource summary')
plt.grid()
plt.show()

```



Can't infer much detail as they are overlapping

```
In [61]: plt.figure(figsize=(10,3))
sns.distplot(approved_summary_word_count, hist=False, label="Approved Projects")
sns.distplot(rejected_summary_word_count, hist=False, label="Not Approved Projects")
plt.title('Words for each essay of the project')
plt.xlabel('Number of words in each eassay')
plt.legend()
plt.show()
```



Can't infer much info

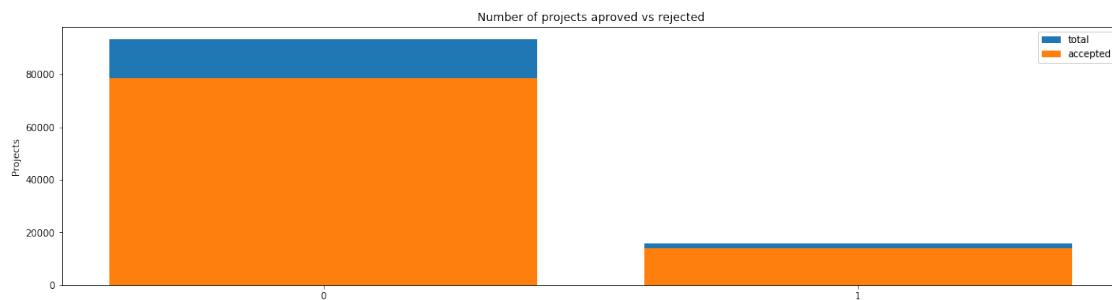
2.0.16 To find whether numerical digits in project summary exist or not

```
In [62]: #re.findall(r'\d+')
def find_num(text):
    if re.findall(r'\d+', text):
        return 1
    return 0

In [114]: project_data['numerical_digits'] = project_data['project_resource_summary'].apply(lambda x: len(re.findall(r'\d+', x)))
# https://stackoverflow.com/questions/22407798/how-to-reset-a-dataframes-indexes-for-num_data = project_data.groupby('id').agg({'numerical_digits':'sum'}).reset_index()
#project_data = pd.merge(project_data, num_data, on='id', how='left').reset_index()

In [112]: project_data['numerical_digits'] = project_data['numerical_digits'].apply(str).head(10)

In [115]: univariate_barplots(project_data, 'numerical_digits', 'project_is_approved', top=20)
```



```
numerical_digits  project_is_approved  total      Avg
0                  0                 78616  93492  0.840885
1                  1                 14090  15756  0.894263
=====
numerical_digits  project_is_approved  total      Avg
0                  0                 78616  93492  0.840885
1                  1                 14090  15756  0.894263
```

Project which are approved have the numerical digits present but not all of the project with numerical digit guaranteed that the project will get approved. We can say that chances of getting the project approved is high when there is numerical digit present in the 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.

2.1 1.3 Text preprocessing

2.1.1 1.3.1 Essay Text

In [66]: project_data.head(2)

```
Out[66]:   Unnamed: 0          id              teacher_id teacher_prefix \
0      160221  p253737  c90749f5d961ff158d4b4d1e7dc665fc      Mrs.
1      140945  p258326  897464ce9ddc600bcfd1151f324dd63a      Mr.

                           school_state project_submitted_datetime project_grade_category \
0                  IN            2016-12-05 13:43:57           Grades PreK-2
1                  FL            2016-10-25 09:22:10           Grades 6-8

                           project_title \
0  Educational Support for English Learners at Home
1  Wanted: Projector for Hungry Learners

                           project_essay_1 \
0  My students are English learners that are work...
1  Our students arrive to our school eager to lea...

                           project_essay_2      ...
0  \"The limits of your language are the limits o...
1  The projector we need for our school is very c...
                           ...
                           ...
                           project_resource_summary \
0          NaN  My students need opportunities to practice beg...
1          NaN  My students need a projector to help with view...
                           ...
                           ...
                           teacher_number_of_previously_posted_projects  project_is_approved \
0                               0                      0
1                               7                      1

                           clean_categories      clean_subcategories \
0  Literacy_Language          ESL Literacy
1  History_Civics Health_Sports  Civics_Government TeamSports

                           essay  price  quantity \
0  My students are English learners that are work...  154.6      23
1  Our students arrive to our school eager to lea...  299.0      1

                           numerical_digits
0                          0
1                          0

[2 rows x 21 columns]
```

In [67]: # printing some random essays.
print(project_data['essay'].values[0])

```

print("=="*50)
print(project_data['essay'].values[150])
print("=="*50)
print(project_data['essay'].values[1000])
print("=="*50)
print(project_data['essay'].values[20000])
print("=="*50)
print(project_data['essay'].values[99999])
print("=="*50)

```

My students are English learners that are working on English as their second or third language

The 51 fifth grade students that will cycle through my classroom this year all love learning, a

How do you remember your days of school? Was it in a sterile environment with plain walls, row

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

The mediocre teacher tells. The good teacher explains. The superior teacher demonstrates. The g

In [68]: # <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 [69]: sent = decontracted(project_data['essay'].values[20000])

```

print(sent)
print("=="*50)

```

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

```
In [70]: # \r \n \t remove from string python: http://texthandler.com/info/remove-line-breaks-
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, cogni-

```
In [71]: #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 cogni-

```
In [72]: # 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'", "you'll", "you'd", 'your', 'yours', 'yourself', 'yourselves', 'he', 'him', 'she', "she's", 'her', 'hers', 'herself', 'it', "it's", 'its', 'itself', 'theirs', 'themselves', 'what', 'which', 'who', 'whom', 'this', 'that', "that's", 'am', 'is', 'are', 'was', 'were', 'be', 'been', 'being', 'have', 'has', 'had', 'did', 'doing', 'a', 'an', 'the', 'and', 'but', 'if', 'or', 'because', 'as', 'at', 'by', 'for', 'with', 'about', 'against', 'between', 'into', 'through', 'above', 'below', 'to', 'from', 'up', 'down', 'in', 'out', 'on', 'off', 'on', 'then', 'once', 'here', 'there', 'when', 'where', 'why', 'how', 'all', 'any', 'most', 'other', 'some', 'such', 'only', 'own', 'same', 'so', 'than', 'too', 's', 't', 'can', 'will', 'just', 'don', "don't", 'should', "should've", 're', 've', 'y', 'ain', 'aren', "aren't", 'couldn', "couldn't", 'didn', "didn't", "hadn't", 'hasn', "hasn't", 'haven', "haven't", 'isn', "isn't", 'ma', 'mig', "mustn't", 'needn', "needn't", 'shan', "shan't", 'shouldn', "shouldn't", 'won', "won't", 'wouldn', "wouldn't"]
```

```
In [73]: # Combining all the above statemennts
from tqdm import tqdm
preprocessed_essays = []
# tqdm is for printing the status bar
for sentance in tqdm(project_data['essay'].values):
    sent = decontracted(sentance)
    sent = sent.replace('\r', ' ')
    sent = sent.replace('\n', ' ')
    sent = sent.replace('\t', ' ')
    sent = re.sub('[^A-Za-z0-9]+', ' ', sent)
    # https://gist.github.com/sebleier/554280
    sent = ' '.join(e for e in sent.split() if e not in stopwords)
    preprocessed_essays.append(sent.lower().strip())
```

100%|| 109248/109248 [02:23<00:00, 760.67it/s]

```
In [74]: # after preprocessing
preprocessed_essays[20000]
```

```
Out[74]: 'my kindergarten students varied disabilities ranging speech language delays cognitive
```

1.3.2 Project title Text

```
In [75]: # Combining all the above statements
from tqdm import tqdm
preprocessed_titles = []
# 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_titles.append(sent.lower().strip())
```

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```
In [76]: # after preprocessing
preprocessed_titles[20000]
```

```
Out[76]: 'we need to move it while we input it'
```

2.2 1.4 Preparing data for models

```
In [77]: project_data.columns
```

```
Out[77]: 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',
                 'numerical_digits'],
                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

```

2.2.1 1.4.1 Vectorizing Categorical data

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

```
In [120]: # 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)
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', '...']
Shape of matrix after one hot encoding  (109248, 9)
```

```
In [124]: # we use count vectorizer to convert the values into one hot encoded features
vectorizer = CountVectorizer(vocabulary=list(sorted_sub_cat_dict.keys()), lowercase=False)
vectorizer.fit(project_data['clean_subcategories'].values)
print(vectorizer.get_feature_names()

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

['Economics', 'CommunityService', 'FinancialLiteracy', 'ParentInvolvement', 'Extracurricular', ...]
Shape of matrix after one hot encoding  (109248, 30)
```

```
In [122]: # Please do the similar feature encoding with state, teacher_prefix and project_grad
# we use count vectorizer to convert the values into one hot encoded features
vectorizer = CountVectorizer(vocabulary=list(sorted_cat_dict.keys()), lowercase=False)
vectorizer.fit(project_data['school_state'].values)
print(vectorizer.get_feature_names()

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

```

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

In [130]: categories_one_hot_num_dig = pd.get_dummies(project_data['numerical_digits'].values)
           print("Shape of matrix after one hot encodig ",categories_one_hot_num_dig.shape)

Shape of matrix after one hot encodig (109248, 2)

In [82]: print(project_data['teacher_prefix'].unique())
           project_data[['teacher_id','teacher_prefix']].groupby('teacher_prefix').count()

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

Out[82]:
teacher_id
teacher_prefix
Dr.          13
Mr.         10648
Mrs.        57269
Ms.         38955
Teacher      2360

In [83]: project_data['teacher_prefix']=project_data['teacher_prefix'].fillna('Mrs.')

In [84]: # 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)
           vectorizer.fit(project_data['teacher_prefix'].values)
           print(vectorizer.get_feature_names())

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

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

In [85]: # 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)
           vectorizer.fit(project_data['project_grade_category'].values)
           print(vectorizer.get_feature_names())

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

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

```

2.2.2 1.4.2 Vectorizing Text data

1.4.2.1 Bag of words

```
In [86]: # We are considering only the words which appeared in at least 10 documents(rows or p
vectorizer = CountVectorizer(min_df=10)
text_bow_essays = vectorizer.fit_transform(preprocessed_essays)
print("Shape of matrix after one hot encodig ",text_bow_essays.shape)

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

1.4.2.2 Bag of Words on project_title

```
In [87]: # you can vectorize the title also
# before you vectorize the title make sure you preprocess it
# We are considering only the words which appeared in at least 10 documents(rows or p
vectorizer = CountVectorizer(min_df=10)
text_bow_titles = vectorizer.fit_transform(preprocessed_titles)
print("Shape of matrix after one hot encodig ",text_bow_titles.shape)

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

1.4.2.3 TFIDF vectorizer

```
In [88]: from sklearn.feature_extraction.text import TfidfVectorizer
vectorizer = TfidfVectorizer(min_df=10)
text_tfidf_essays = vectorizer.fit_transform(preprocessed_essays)
print("Shape of matrix after one hot encodig ",text_tfidf_essays.shape)

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

1.4.2.4 TFIDF Vectorizer on project_title

```
In [89]: # Similarly you can vectorize for title also
from sklearn.feature_extraction.text import TfidfVectorizer
vectorizer = TfidfVectorizer(min_df=10)
text_tfidf_titles = vectorizer.fit_transform(preprocessed_titles)
print("Shape of matrix after one hot encodig ",text_tfidf_titles.shape)

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

1.4.2.5 Using Pretrained Models: Avg W2V

```
In [90]: '''
# 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 preproc_text:
    words.extend(i.split(' '))

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

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

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

# stronging variables into pickle files python: http://www.jessicayung.com/how-to-use

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

```

...

```
Out[90]: '\n# Reading glove vectors in python: https://stackoverflow.com/a/38230349/4084039\n'

In [91]: # stronging variables into pickle files python: http://www.jessicayung.com/how-to-use
          # 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 [92]: # average Word2Vec
          # compute average word2vec for each review.
          avg_w2v_vectors_essays = [] # the avg-w2v for each sentence/review is stored in this
          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_essays.append(vector)

          print(len(avg_w2v_vectors_essays))
          print(len(avg_w2v_vectors_essays[0]))
```

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109248

300

1.4.2.6 Using Pretrained Models: AVG W2V on project_title

```
In [93]: # Similarly you can vectorize for title also
          # average Word2Vec
          # compute average word2vec for each review.
          avg_w2v_vectors_titles = [] # the avg-w2v for each sentence/review is stored in this
          for sentence in tqdm(preprocessed_titles): # for each review/sentence
              vector = np.zeros(300) # as word vectors are of zero length
              cnt_words = 0; # num of words with a valid vector in the sentence/review
              for word in sentence.split(): # for each word in a review/sentence
                  if word in glove_words:
                      vector += model[word]
                      cnt_words += 1
```

```

if cnt_words != 0:
    vector /= cnt_words
avg_w2v_vectors_titles.append(vector)

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

100%|| 109248/109248 [00:04<00:00, 23938.31it/s]

```

109248
300

1.4.2.7 Using Pretrained Models: TFIDF weighted W2V

```

In [94]: # 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 [95]: # average Word2Vec
# compute average word2vec for each review.
tfidf_w2v_vectors_essays = [] # the avg-w2v for each sentence/review is stored in this list
for sentence in tqdm(preprocessed_essays): # for each review/sentence
    vector = np.zeros(300) # as word vectors are of zero length
    tf_idf_weight = 0; # num of words with a valid vector in the sentence/review
    for word in sentence.split(): # for each word in a review/sentence
        if (word in glove_words) and (word in tfidf_words):
            vec = model[word] # getting the vector for each word
            # here we are multiplying idf value(dictionary[word]) and the tf value((sentence.count(word)/len(sentence.split())))
            tf_idf = dictionary[word]*(sentence.count(word)/len(sentence.split())) #
            vector += (vec * tf_idf) # calculating tfidf weighted w2v
            tf_idf_weight += tf_idf
    if tf_idf_weight != 0:
        vector /= tf_idf_weight
    tfidf_w2v_vectors_essays.append(vector)

print(len(tfidf_w2v_vectors_essays))
print(len(tfidf_w2v_vectors_essays[0]))

100%|| 109248/109248 [08:11<00:00, 222.46it/s]

```

109248
300

1.4.2.9 Using Pretrained Models: TFIDF weighted W2V on project_title

```
In [96]: # Similarly you can vectorize for title also
# S = ["abc def pqr", "def def def abc", "pqr pqr def"]
tfidf_model = TfidfVectorizer()
tfidf_model.fit(preprocessed_titles)
# we are converting a dictionary with word as a key, and the idf as a value
dictionary = dict(zip(tfidf_model.get_feature_names(), list(tfidf_model.idf_)))
tfidf_words = set(tfidf_model.get_feature_names())

In [97]: # average Word2Vec
    # compute average word2vec for each review.
    tfidf_w2v_vectors_titles = [] # the avg-w2v for each sentence/review is stored in this list
    for sentence in tqdm(preprocessed_titles): # for each review/sentence
        vector = np.zeros(300) # as word vectors are of zero length
        tf_idf_weight = 0; # num of words with a valid vector in the sentence/review
        for word in sentence.split(): # for each word in a review/sentence
            if (word in glove_words) and (word in tfidf_words):
                vec = model[word] # getting the vector for each word
                # here we are multiplying idf value(dictionary[word]) and the tf value((sentence.count(word)/len(sentence.split())))
                tf_idf = dictionary[word]*(sentence.count(word)/len(sentence.split())) # calculating tf-idf
                vector += (vec * tf_idf) # calculating tfidf weighted w2v
                tf_idf_weight += tf_idf
        if tf_idf_weight != 0:
            vector /= tf_idf_weight
        tfidf_w2v_vectors_titles.append(vector)

    print(len(tfidf_w2v_vectors_titles))
    print(len(tfidf_w2v_vectors_titles[0]))

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109248
300
```

2.2.3 1.4.3 Vectorizing Numerical features

```
In [98]: # check this one: https://www.youtube.com/watch?v=OH0q0cln3Z4&t=530s
# standardization sklearn: https://scikit-learn.org/stable/modules/generated/sklearn.p...
from sklearn.preprocessing import StandardScaler

# price_standardized = standardScalar.fit(project_data['price'].values)
# this will rise the error
# ValueError: Expected 2D array, got 1D array instead: array=[725.05 213.03 329. ...
# 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
print(f"Mean : {price_scalar.mean_[0]}, Standard deviation : {np.sqrt(price_scalar.var_)}")

# Now standardize the data with above mean and variance.
price_standardized = price_scalar.transform(project_data['price'].values.reshape(-1, 1))

Mean : 298.1193425966608, Standard deviation : 367.49634838483496

In [99]: price_standardized

Out[99]: array([-0.3905327 ,  0.00239637,  0.59519138, ..., -0.15825829, -0.61243967, -0.51216657])

In [100]: # check this one: https://www.youtube.com/watch?v=OH0q0cln3Z4&t=530s
          # standardization sklearn: https://scikit-learn.org/stable/modules/generated/sklearn
          from sklearn.preprocessing import StandardScaler

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

          project_scalar = StandardScaler()
          project_scalar.fit(project_data['teacher_number_of_previously_posted_projects'].values)
          print(f"Mean : {project_scalar.mean_[0]}, Standard deviation : {np.sqrt(project_scalar.var_)}")

          # Now standardize the data with above mean and variance.
          project_standardized = project_scalar.transform(project_data['teacher_number_of_previously_posted_projects'].values)

C:\Users\Kriti chaudhary\Anaconda3\lib\site-packages\sklearn\utils\validation.py:475: DataConversionWarning:
Data with input dtype int64 was converted to float64 by StandardScaler.

Mean : 11.153165275336848, Standard deviation : 27.77702641477403

C:\Users\Kriti chaudhary\Anaconda3\lib\site-packages\sklearn\utils\validation.py:475: DataConversionWarning:
Data with input dtype int64 was converted to float64 by StandardScaler.

```

```
In [101]: project_standardized
```

```
Out[101]: array([[-0.40152481],  
                 [-0.14951799],  
                 [-0.36552384],  
                 ...,  
                 [-0.29352189],  
                 [-0.40152481],  
                 [-0.40152481]])
```

2.2.4 1.4.4 Merging all the above features

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

```
In [131]: print(categories_one_hot.shape)  
        print(sub_categories_one_hot.shape)  
        print(categories_one_hot_teacher_prefix.shape)  
        print(categories_one_hot_grade.shape)  
        print(categories_one_hot_num_dig.shape)  
        print(categories_one_hot_state.shape)  
        print(text_bow_essays.shape)  
        print(text_bow_titles.shape)  
        print(price_standardized.shape)  
        print(project_standardized.shape)
```

```
(109248, 9)  
(109248, 30)  
(109248, 9)  
(109248, 9)  
(109248, 2)  
(109248, 9)  
(109248, 16623)  
(109248, 3329)  
(109248, 1)  
(109248, 1)
```

```
In [ ]: '''# 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'''
```

```
In [132]: from scipy.sparse import hstack  
X = hstack((sub_categories_one_hot, categories_one_hot, categories_one_hot_state, categories_one_hot_teacher_prefix, text_bow, price_standardized))  
X.shape
```

```
Out[132]: (109248, 70)
```

```
In [133]: A = hstack((X,text_bow_titles))
B = hstack((X,text_tfidf_titles))
C = hstack((X,avg_w2v_vectors_titles))
D = hstack((X,tfidf_w2v_vectors_titles))
```

```
In [134]: m = hstack((A,B,C,D))
A.shape,B.shape,C.shape,D.shape, m.shape
```

Out[134]: ((109248, 3399), (109248, 3399), (109248, 370), (109248, 370), (109248, 7538))

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.

 In the above cells we have plotted and analyzed many features. Please observe the plots and EDA: Please complete the analysis of the feature: teacher_number_of_previously_posted_projects

- Build the data matrix using these features
 - school_state : categorical data (one hot encoding)
 - clean_categories : categorical data (one hot encoding)
 - clean_subcategories : categorical data (one hot encoding)
 - teacher_prefix : categorical data (one hot encoding)
 - project_title : text data (BOW, TFIDF, AVG W2V, TFIDF W2V)
 - price : numerical
 - teacher_number_of_previously_posted_projects : numerical

 Now, plot FOUR t-SNE plots with each of these feature sets.

-
 - categorical, numerical features + project_title(BOW)
 - categorical, numerical features + project_title(TFIDF)
 - categorical, numerical features + project_title(AVG W2V)
 - categorical, numerical features + project_title(TFIDF W2V)

 Concatenate all the features and Apply TNSE on the final data matrix

 Note 1: The TSNE accepts only dense matrices

 Note 2: Consider only 5k to 6k data points to avoid memory issues. If you want to use all the data, then convert the sparse matrix to a dense matrix

```
In [147]: def convert2matrix(data,no_of_points):
    k=csr_matrix(data) [:no_of_points]
    u = k.todense()
    u = u[:, :]
    return u
```

2.1 TSNE with BOW encoding of project_title feature

```

In [149]: # please write all of the code with proper documentation and proper titles for each .
          # 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
from scipy.sparse import csr_matrix
import time

time_start = time.time()
npoints = 5000
data_5000_1 = convert2matrix(A,npoints)

labels_5000_1 = project_data['project_is_approved'][:npoints]

model = TSNE(n_components=2,verbose=1, random_state=0,n_iter=2000)
# configuring the parameters
# the number of components = 2
# default perplexity = 30
# default learning rate = 200
# default Maximum number of iterations for the optimization = 1000

tsne_data = model.fit_transform(data_5000_1)
print('t-SNE done! Time elapsed: {} seconds'.format(time.time()-time_start))

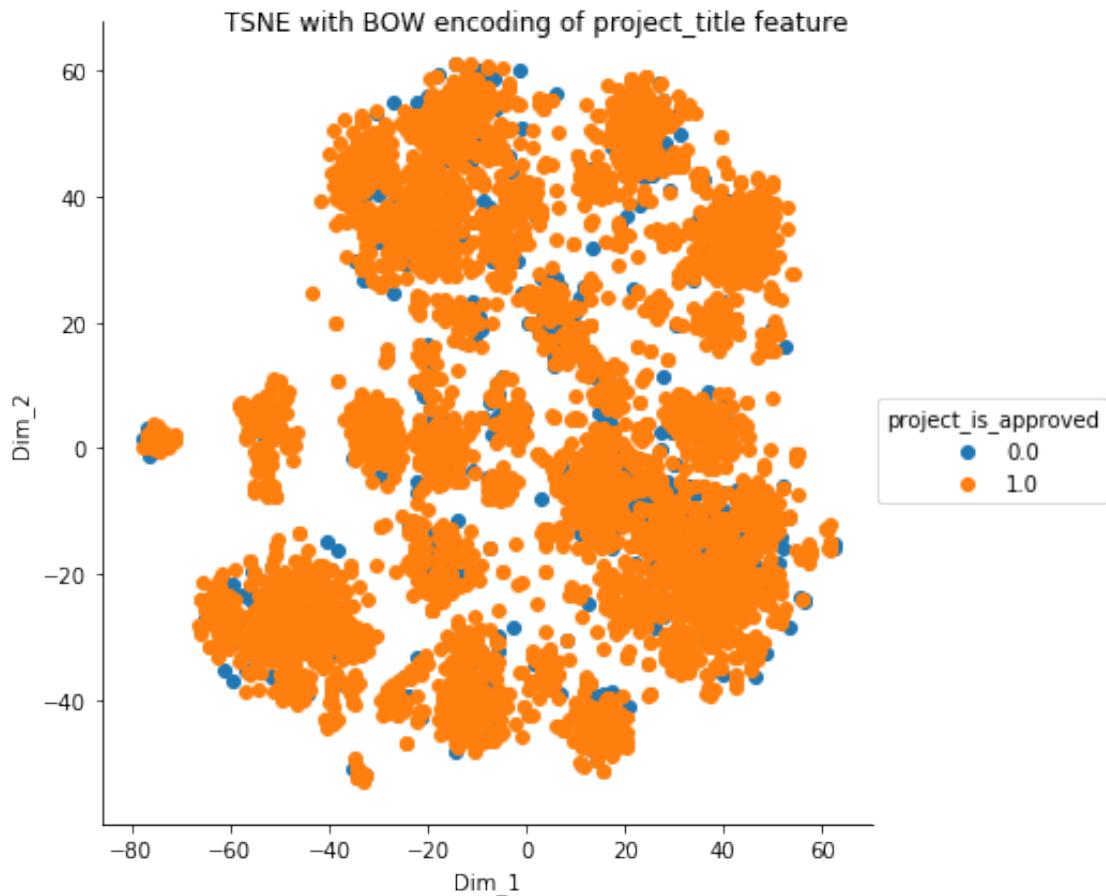
# creating a new data frame which help us in plotting the result data
tsne_data = np.vstack((tsne_data.T, labels_5000_1)).T
tsne_df = pd.DataFrame(data=tsne_data, columns=["Dim_1", "Dim_2", "project_is_approved"])

# Ploting the result of tsne
g = sns.FacetGrid(tsne_df, hue="project_is_approved", size=6).map(plt.scatter, 'Dim_1', 'Dim_2')
g.fig.suptitle('TSNE with BOW encoding of project_title feature')
plt.show()

[t-SNE] Computing 91 nearest neighbors...
[t-SNE] Indexed 5000 samples in 1.353s...
[t-SNE] Computed neighbors for 5000 samples in 205.160s...
[t-SNE] Computed conditional probabilities for sample 1000 / 5000
[t-SNE] Computed conditional probabilities for sample 2000 / 5000
[t-SNE] Computed conditional probabilities for sample 3000 / 5000
[t-SNE] Computed conditional probabilities for sample 4000 / 5000
[t-SNE] Computed conditional probabilities for sample 5000 / 5000
[t-SNE] Mean sigma: 0.838553
[t-SNE] KL divergence after 250 iterations with early exaggeration: 86.509628
[t-SNE] Error after 2000 iterations: 1.848597

```

t-SNE done! Time elapsed: 724.5056474208832 seconds



As you can see there are clusters of points but blue points and orange points are overlapping each other so it's hard to interpret. Since we are having less no. of points for class 0. But we are sure of one thing that its high-dimensional data contains sufficient variability because there are clusters of orange points. So we can use these features for project classification.

2.2 TSNE with TFIDF encoding of project_title feature

In [150]: # please write all the code with proper documentation, and proper titles for each sub

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
from scipy.sparse import csr_matrix
import time
```

```

time_start = time.time()
npoints = 5000
# Picking the top 1000 points as TSNE takes a lot of time for 15K points
data_5000_2 = convert2matrix(B,npoints)

labels_5000_2 = project_data['project_is_approved'][npoints]

model = TSNE(n_components=2,verbose=1, random_state=0,n_iter=2000)
# configuring the parameteres
# the number of components = 2
# default perplexity = 30
# default learning rate = 200
# default Maximum number of iterations for the optimization = 1000

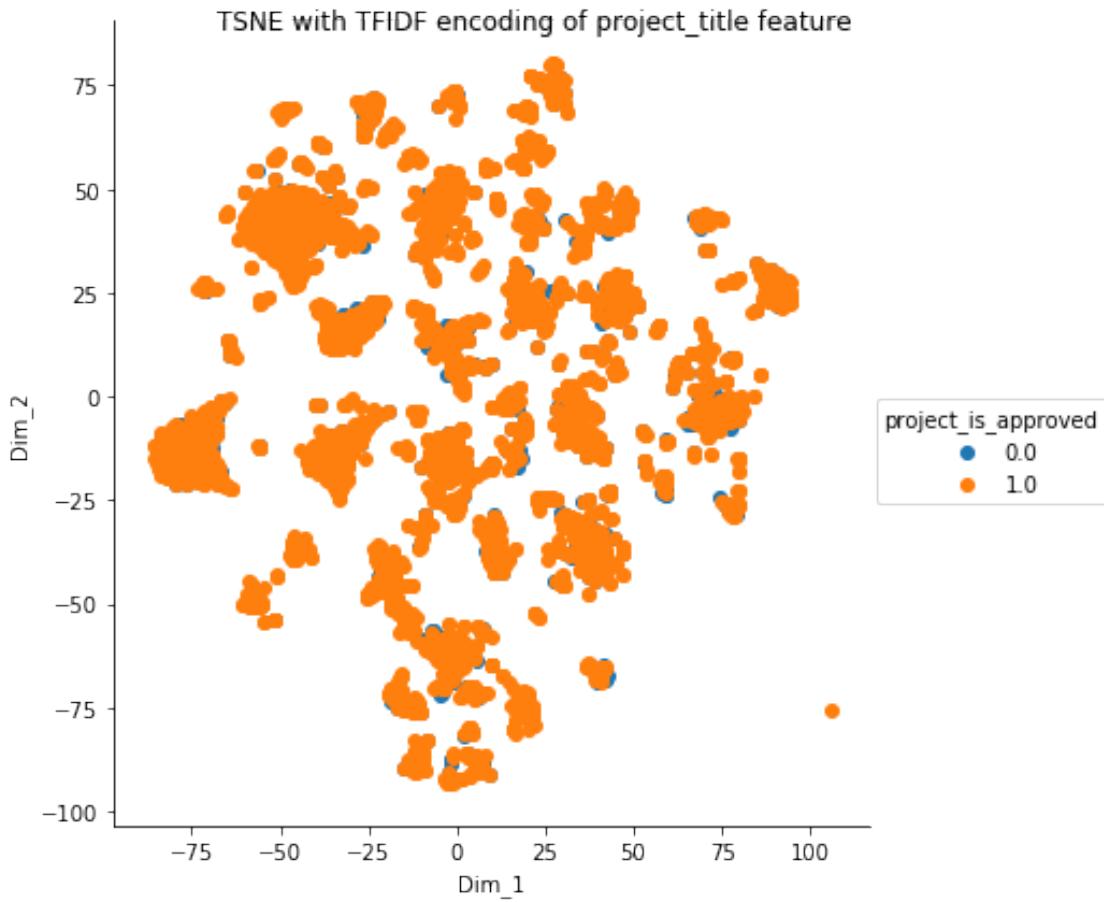
tsne_data = model.fit_transform(data_5000_2)
print('t-SNE done! Time elapsed: {} seconds'.format(time.time()-time_start))

# creating a new data frame which help us in plotting the result data
tsne_data = np.vstack((tsne_data.T, labels_5000_2)).T
tsne_df = pd.DataFrame(data=tsne_data, columns=["Dim_1", "Dim_2", "project_is_approved"])

# Ploting the result of tsne
g = sns.FacetGrid(tsne_df, hue="project_is_approved", size=6).map(plt.scatter, 'Dim_1', 'Dim_2')
g.fig.suptitle('TSNE with TFIDF encoding of project_title feature')
plt.show()

[t-SNE] Computing 91 nearest neighbors...
[t-SNE] Indexed 5000 samples in 1.236s...
[t-SNE] Computed neighbors for 5000 samples in 188.749s...
[t-SNE] Computed conditional probabilities for sample 1000 / 5000
[t-SNE] Computed conditional probabilities for sample 2000 / 5000
[t-SNE] Computed conditional probabilities for sample 3000 / 5000
[t-SNE] Computed conditional probabilities for sample 4000 / 5000
[t-SNE] Computed conditional probabilities for sample 5000 / 5000
[t-SNE] Mean sigma: 0.475031
[t-SNE] KL divergence after 250 iterations with early exaggeration: 74.491501
[t-SNE] Error after 2000 iterations: 0.952860
t-SNE done! Time elapsed: 626.5070950984955 seconds

```



Here we can see better clustering of orange points hence the data can be reduced to lower dimensions. Also these features would be best to work with for classifying the projects whether it get approved or rejected

2.3 TSNE with AVG W2V encoding of project_title feature

```
In [151]: # please write all the code with proper documentation, and proper titles for each sub
          # 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
from scipy.sparse import csr_matrix
import time

time_start = time.time()
npoints = 5000
# Picking the top 1000 points as TSNE takes a lot of time for 15K points
data_5000_3 = convert2matrix(C,npoints)
```

```

labels_5000_3 = project_data['project_is_approved'][:npoints]

model = TSNE(n_components=2, verbose=1, random_state=0, n_iter=2000)
# configuring the parameteres
# the number of components = 2
# default perplexity = 30
# default learning rate = 200
# default Maximum number of iterations for the optimization = 1000

tsne_data = model.fit_transform(data_5000_3)
print('t-SNE done! Time elapsed: {} seconds'.format(time.time()-time_start))

# creating a new data frame which help us in plotting the result data
tsne_data = np.vstack((tsne_data.T, labels_5000_3)).T
tsne_df = pd.DataFrame(data=tsne_data, columns=["Dim_1", "Dim_2", "project_is_approved"])

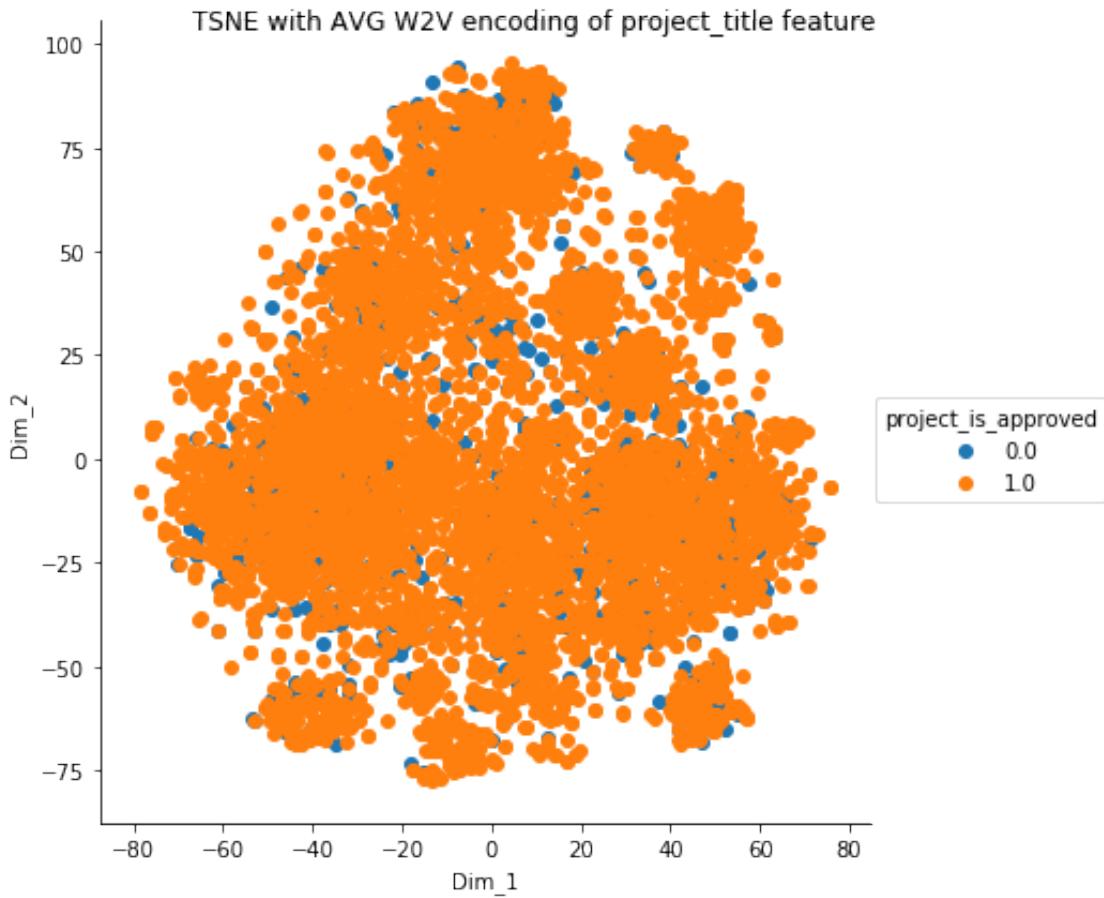
# Ploting the result of tsne
g = sns.FacetGrid(tsne_df, hue="project_is_approved", size=6).map(plt.scatter, 'Dim_1', 'Dim_2')
g.fig.suptitle('TSNE with AVG W2V encoding of project_title feature')
plt.show()

```

```

[t-SNE] Computing 91 nearest neighbors...
[t-SNE] Indexed 5000 samples in 0.171s...
[t-SNE] Computed neighbors for 5000 samples in 28.874s...
[t-SNE] Computed conditional probabilities for sample 1000 / 5000
[t-SNE] Computed conditional probabilities for sample 2000 / 5000
[t-SNE] Computed conditional probabilities for sample 3000 / 5000
[t-SNE] Computed conditional probabilities for sample 4000 / 5000
[t-SNE] Computed conditional probabilities for sample 5000 / 5000
[t-SNE] Mean sigma: 1.120775
[t-SNE] KL divergence after 250 iterations with early exaggeration: 86.365677
[t-SNE] Error after 2000 iterations: 2.318992
t-SNE done! Time elapsed: 580.4425520896912 seconds

```



We got a T-Sne graph with lots of overlapping data so odds are high that our classifier will perform badly.

2.4 TSNE with TFIDF Weighted W2V encoding of project_title feature

```
In [152]: # please write all the code with proper documentation, and proper titles for each sub
# 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
from scipy.sparse import csr_matrix
import time

time_start = time.time()
npoints = 5000
# Picking the top 1000 points as TSNE takes a lot of time for 15K points
data_5000_4 = convert2matrix(D,npoints)
```

```

labels_5000_4 = project_data['project_is_approved'][:npoints]

model = TSNE(n_components=2,verbose=1, random_state=0,n_iter=2000)
# configuring the parameters
# the number of components = 2
# default perplexity = 30
# default learning rate = 200
# default Maximum number of iterations for the optimization = 1000

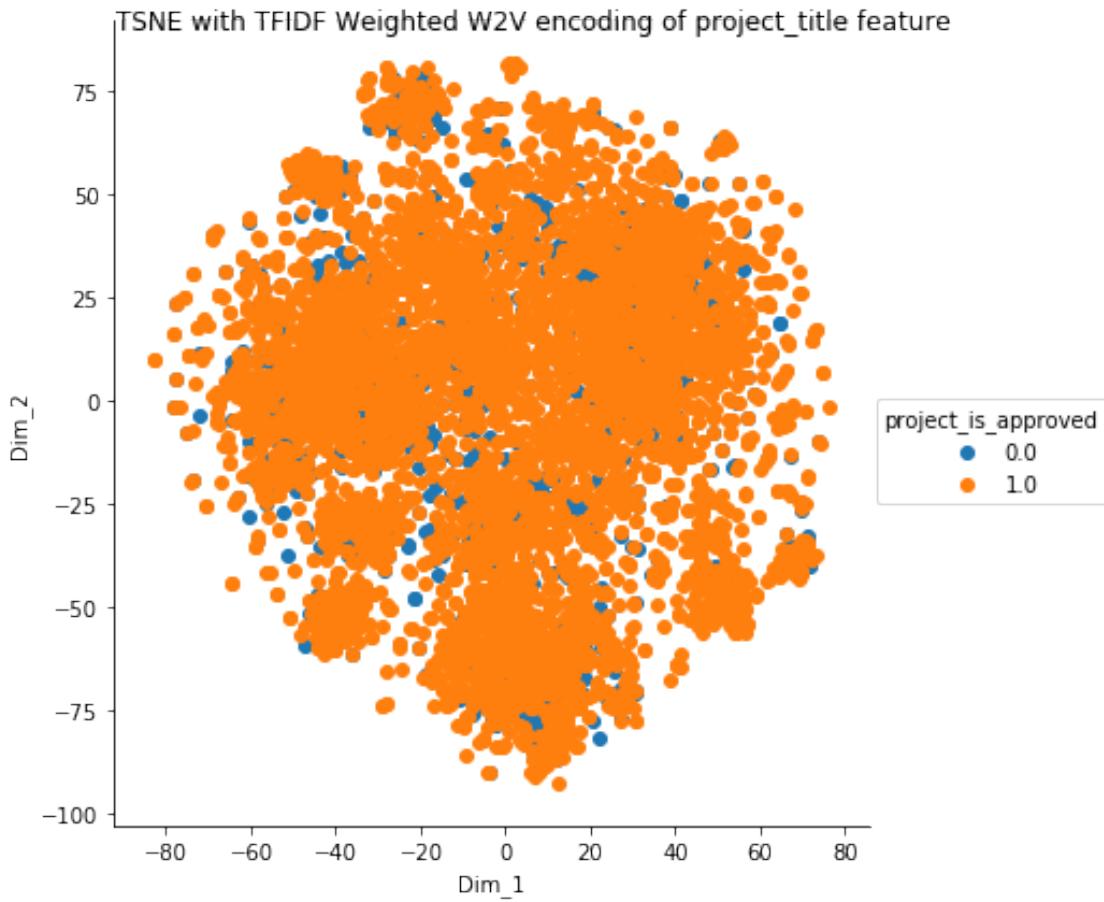
tsne_data = model.fit_transform(data_5000_4)
print('t-SNE done! Time elapsed: {} seconds'.format(time.time()-time_start))

# creating a new data frame which help us in plotting the result data
tsne_data = np.vstack((tsne_data.T, labels_5000_4)).T
tsne_df = pd.DataFrame(data=tsne_data, columns=["Dim_1", "Dim_2", "project_is_approved"])

# Plotting the result of tsne
g = sns.FacetGrid(tsne_df, hue="project_is_approved", size=6).map(plt.scatter, 'Dim_1', 'Dim_2')
g.fig.suptitle('TSNE with TFIDF Weighted W2V encoding of project_title feature')
plt.show()

[t-SNE] Computing 91 nearest neighbors...
[t-SNE] Indexed 5000 samples in 0.186s...
[t-SNE] Computed neighbors for 5000 samples in 27.740s...
[t-SNE] Computed conditional probabilities for sample 1000 / 5000
[t-SNE] Computed conditional probabilities for sample 2000 / 5000
[t-SNE] Computed conditional probabilities for sample 3000 / 5000
[t-SNE] Computed conditional probabilities for sample 4000 / 5000
[t-SNE] Computed conditional probabilities for sample 5000 / 5000
[t-SNE] Mean sigma: 1.185279
[t-SNE] KL divergence after 250 iterations with early exaggeration: 86.269173
[t-SNE] Error after 2000 iterations: 2.388887
t-SNE done! Time elapsed: 587.793505191803 seconds

```



We got a T-Sne graph with lots of overlapping data so odds are high that our classifier will perform badly.

2.5 TSNE on the final data matrix

```
In [162]: # please write all of the code with proper documentation and proper titles for each .
# 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
from scipy.sparse import csr_matrix
import time

time_start = time.time()
npoints = 5000
data_5000 = convert2matrix(m,npoints)

labels_5000 = project_data['project_is_approved'][:npoints]
```

```

model = TSNE(n_components=2,verbose=1, random_state=0,n_iter=1000)
# configuring the parameteres
# the number of components = 2
# default perplexity = 30
# default learning rate = 200
# default Maximum number of iterations for the optimization = 1000

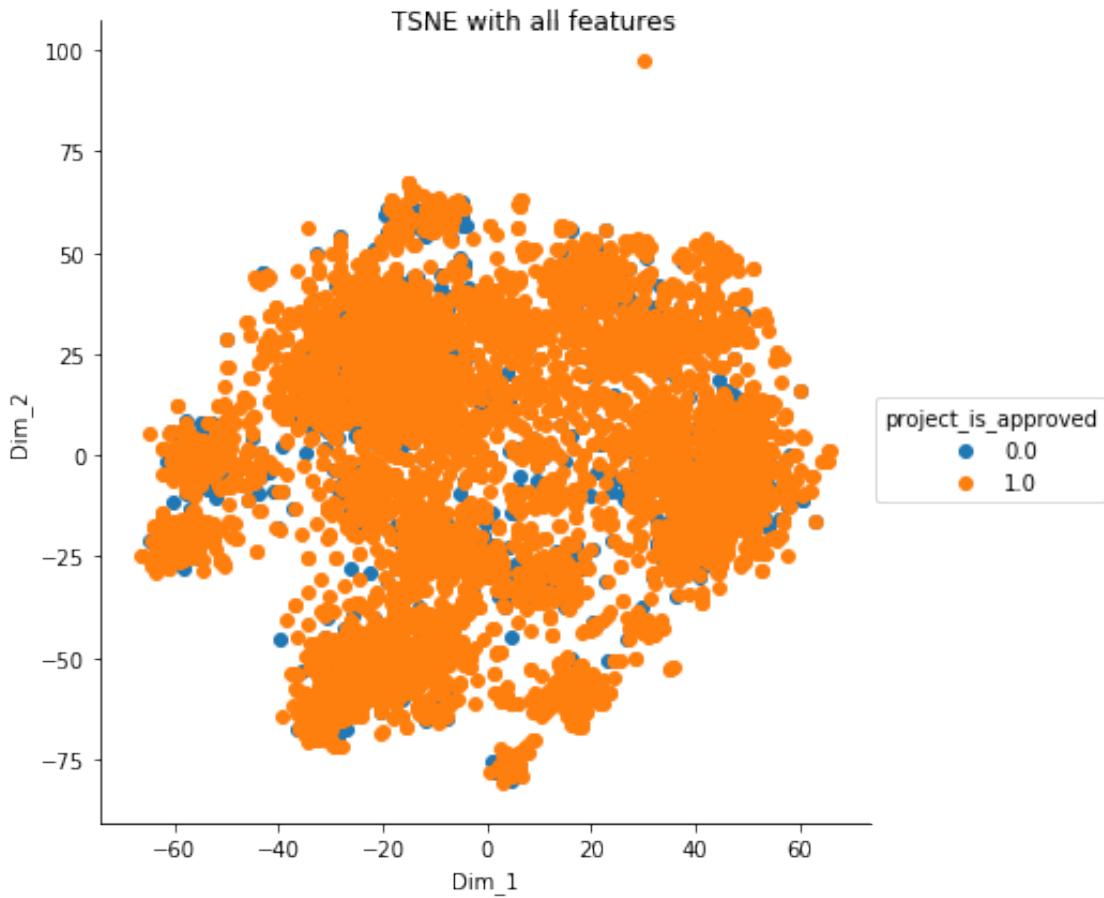
tsne_data = model.fit_transform(data_5000)
print('t-SNE done! Time elapsed: {} seconds'.format(time.time()-time_start))

# creating a new data frame which help us in plotting the result data
tsne_data = np.vstack((tsne_data.T, labels_5000)).T
tsne_df = pd.DataFrame(data=tsne_data, columns=["Dim_1", "Dim_2", "project_is_approved"])

# Ploting the result of tsne
g = sns.FacetGrid(tsne_df, hue="project_is_approved", size=6).map(plt.scatter, 'Dim_1', 'Dim_2')
g.fig.suptitle('TSNE with all features')
plt.show()

[t-SNE] Computing 91 nearest neighbors...
[t-SNE] Indexed 5000 samples in 4.317s...
[t-SNE] Computed neighbors for 5000 samples in 494.754s...
[t-SNE] Computed conditional probabilities for sample 1000 / 5000
[t-SNE] Computed conditional probabilities for sample 2000 / 5000
[t-SNE] Computed conditional probabilities for sample 3000 / 5000
[t-SNE] Computed conditional probabilities for sample 4000 / 5000
[t-SNE] Computed conditional probabilities for sample 5000 / 5000
[t-SNE] Mean sigma: 1.953713
[t-SNE] KL divergence after 250 iterations with early exaggeration: 86.496468
[t-SNE] Error after 1000 iterations: 2.182801
t-SNE done! Time elapsed: 1072.4603147506714 seconds

```



We got a T-Sne graph with lots of overlapping data so odds are high that our classifier will perform badly.

2.5 Summary

Project which were accepted: - mainly from math science and Literacy language category - subcategories as Literacy, and literacy-mathematics had the higher chances of approvance. - had no. of words in project title more than 6. - had no. of words in project essay, between 275 to 280. - had 30 to 90 no. of previously posted project. - had the numerical digits in most of the cases.

categorical, numerical features + project_title(TFIDF), under these features classifier will work better than rest of the features mentioned in other cases, by looking at Tsne plot of it we can say it.