

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	
<code>project_id</code>	A unique identifier for the proposed project.
<code>project_title</code>	Title of the project. • Art Will •
<code>project_grade_category</code>	Grade level of students for which the project is targeted. • • • •

Feature	Description
<code>project_subject_categories</code>	One or more (comma-separated) subject categories from the following enum: <ul style="list-style-type: none"> Literacy & Language Mathematics Science Social Studies Arts Health Physical Education Foreign Languages Other
<code>school_state</code>	State where school is located (Two-letter U.S. state abbreviations)
<code>project_subject_subcategories</code>	One or more (comma-separated) subject subcategories from the following enum: <ul style="list-style-type: none"> Literature & Writing Reading Language Acquisition History Geography Mathematics Science Art Musical Arts Drama Dance Visual Arts Media Arts Design Crafts Gardening Fishing Hunting Travel Volunteering Other
<code>project_resource_summary</code>	An explanation of the resources needed for the project. <ul style="list-style-type: none"> My students need hands on literacy materials.
<code>project_essay_1</code>	File path to the first essay.
<code>project_essay_2</code>	Second file path to the second essay.
<code>project_essay_3</code>	Third file path to the third essay.
<code>project_essay_4</code>	Fourth file path to the fourth essay.
<code>project_submitted_datetime</code>	Datetime when project application was submitted. Example: "2019-08-01T12:00:00"
<code>teacher_id</code>	A unique identifier for the teacher of the proposed project. Example: bdf8baa8fedef6bf
<code>teacher_prefix</code>	Teacher's title. One of the following: <ul style="list-style-type: none"> Principal Vice Principal Assistant Principal Classroom Teacher Paraprofessional Other
<code>teacher_number_of_previously_posted_projects</code>	Number of project applications previously submitted by the teacher.

* 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
<code>project_is_approved</code>	A binary flag indicating whether DonorsChoose approved the project. A value of <code>0</code> indicates the project was not approved, and a value of <code>1</code> 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
```

```
:\\Anaconda\\lib\\site-packages\\gensim\\utils.py:1197: UserWarning: detected Windows; aliasing chunkize to chunkize_serial
warnings.warn("detected Windows; aliasing chunkize to chunkize_serial")
```

1.1 Reading Data

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

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

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

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

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

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

```
Out[4]:
```

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

Checking for nan Values

```
In [5]: print('The Columns with their nan values counts are below ')
for col in project_data.columns:
    print('{col} '.format(col=col),project_data[col].isnull().sum())
```

The Columns with their nan values counts are below

Unnamed: 0 0

id 0

teacher_id 0

teacher_prefix 3

school_state 0

project_submitted_datetime 0

project_grade_category 0

project_subject_categories 0

project_subject_subcategories 0

project_title 0

project_essay_1 0

project_essay_2 0

project_essay_3 105490

project_essay_4 105490

project_resource_summary 0

teacher_number_of_previously_posted_projects 0

project_is_approved 0

The Variable teacher_prefix has 3 missing values and project essays 3 and 4 are almost in 105k range. However, for project essays it's justifiable as system got changed after few years but for

teacher prefix i think they have been mishandled.

```
In [6]: # removing 3 nan values from teacher prefix column as they seems to be outliers  
# DataFrame.dropna(axis=0, how='any', thresh=None, subset=None, inplace=False)  
project_data.dropna(subset=['teacher_prefix'],inplace=True)
```

1.2 Data Analysis

```
In [7]: # 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#s

y_value_counts = project_data['project_is_approved'].value_counts()
print("Number of projects thar are approved for funding ", y_value_counts[1], ",")
print("Number of projects thar are not approved for funding ", y_value_counts[0],)

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)

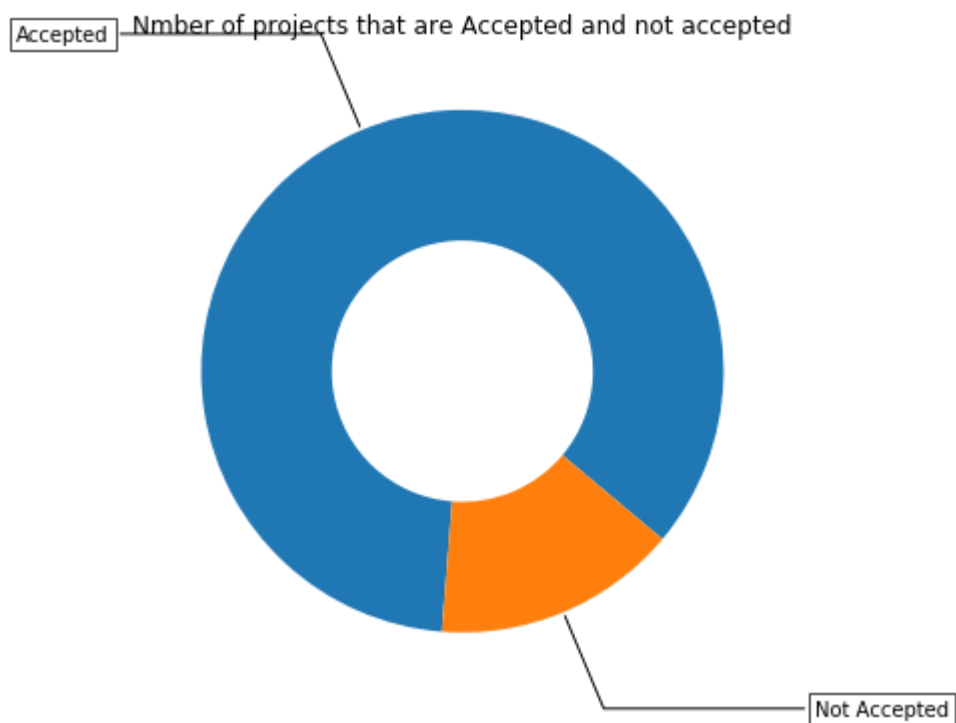
bbox_props = dict(boxstyle="square,pad=0.3", fc="w", ec="k", lw=0.72)
kw = dict(xycoords='data', textcoords='data', arrowprops=dict(arrowstyle="-"),
          bbox=bbox_props, zorder=0, va="center")

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()
```

```
umber of projects thar are approved for funding 92703 , ( 84.85788823287108
)
umber of projects thar are not approved for funding 16542 , ( 15.142111767128
3 %)
```

**Observation :**

From Donut plot, we can see that the chances of project getting approved seems to be high as almost 85%. It means that If a teacher will post his project on this platform, he will have 85% chance for approval for his project by Donors.

1.2.1 Univariate Analysis: School State


```

In [8]: # Pandas dataframe groupby count, mean: https://stackoverflow.com/a/19385591/4084
temp = pd.DataFrame(project_data.groupby("school_state")["project_is_approved"].a
# if you have data which contain only 0 and 1, then the mean = percentage (think
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(84,39,177)']]

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')

```

```
In [9]: # https://www.csi.cuny.edu/sites/default/files/pdf/administration/ops/2Letterstab
temp.sort_values(by=['num_proposals'], inplace=True)
print("States with lowest % approvals")
print(temp.head(5))
print('='*50)
print("States with highest % approvals")
print(temp.tail(5))
```

States with lowest % approvals

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

=====

States with highest % approvals

	state_code	num_proposals
30	NH	0.873563
35	OH	0.875152
47	WA	0.876178
28	ND	0.888112
8	DE	0.897959

Observations:

The state vermont has the least acceptance rate and Delaware has the highest acceptance rate.

```
In [10]: #stacked bar plots matplotlib: https://matplotlib.org/gallery/lines_bars_and_marks
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 [11]: def univariate_barplots(data, col1, col2='project_is_approved', top=False):
# Count number of zeros in dataframe python: https://stackoverflow.com/a/5154
temp = pd.DataFrame(project_data.groupby(col1)[col2].agg(lambda x: x.eq(1).sum))

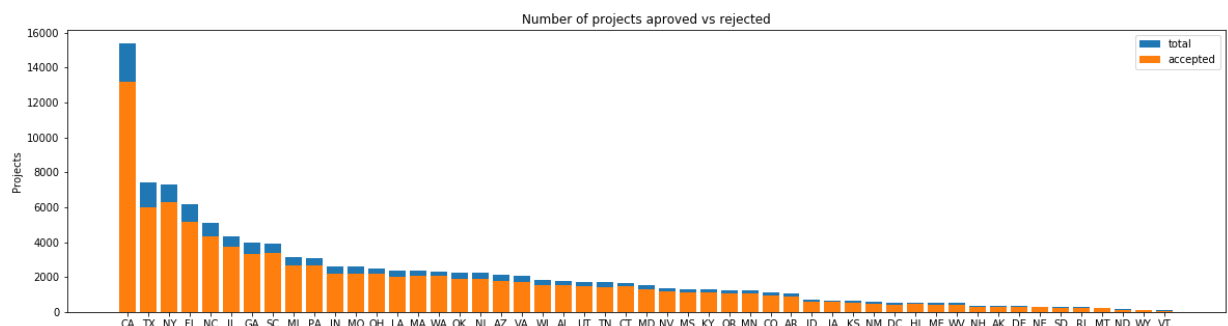
# Pandas dataframe grouby count: https://stackoverflow.com/a/19385591/4084039
temp['total'] = pd.DataFrame(project_data.groupby(col1)[col2].agg({'total': 'count'}))
temp['Avg'] = pd.DataFrame(project_data.groupby(col1)[col2].agg({'Avg': 'mean'}))

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

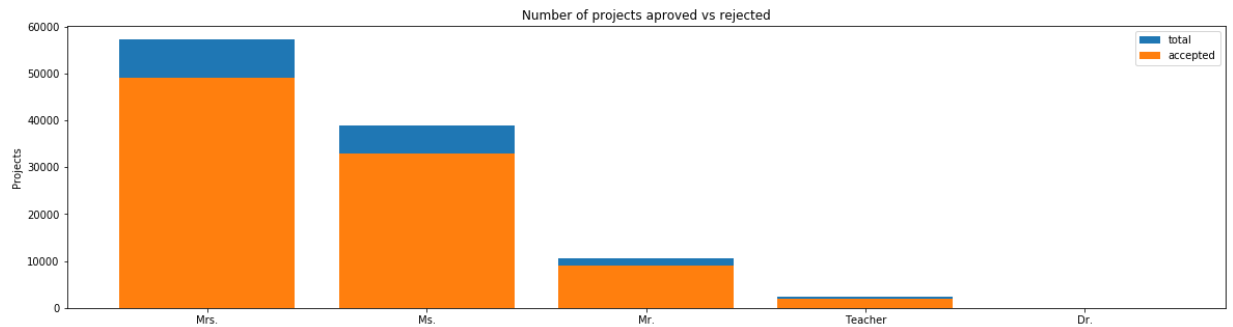


	school_state	project_is_approved	total	Avg
4	CA	13204	15387	0.858127
43	TX	6014	7396	0.813142
34	NY	6291	7318	0.859661
9	FL	5144	6185	0.831690
27	NC	4353	5091	0.855038
=====				
	school_state	project_is_approved	total	Avg
39	RI	243	285	0.852632
26	MT	200	245	0.816327
28	ND	127	143	0.888112
50	WY	82	98	0.836735
46	VT	64	80	0.800000

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

1.2.2 Univariate Analysis: teacher_prefix

In [13]: `univariate_barplots(project_data, 'teacher_prefix', 'project_is_approved' , top=F`



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

```
=====
```

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

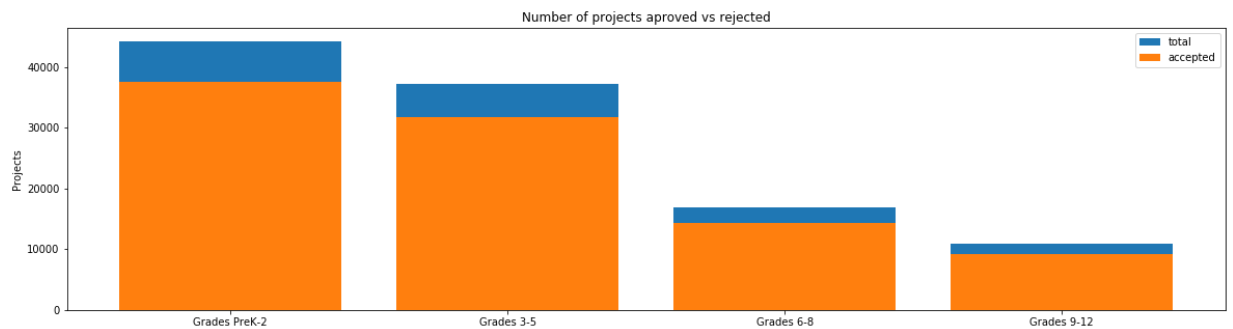
Summary:

The Project approval is high for titles Mrs, Ms, Mr almost 85%. The project approval is least for prefix having dr but for dr we don't have enough data to conclude anything as total number of submission is only 13 and out of 13, 9 got approved.

The teacher prefix Mrs has maximum number of submission and success rate.

1.2.3 Univariate Analysis: project_grade_category

In [14]: `univariate_barplots(project_data, 'project_grade_category', 'project_is_approved')`



	project_grade_category	project_is_approved	total	Avg
3	Grades PreK-2	37536	44225	0.848751
0	Grades 3-5	31727	37135	0.854369
1	Grades 6-8	14258	16923	0.842522
2	Grades 9-12	9182	10962	0.837621

	project_grade_category	project_is_approved	total	Avg
3	Grades PreK-2	37536	44225	0.848751
0	Grades 3-5	31727	37135	0.854369
1	Grades 6-8	14258	16923	0.842522
2	Grades 9-12	9182	10962	0.837621

Summary:

The Grades 3 to 5 has highest number of approval rate i.e. 85% and Grade 9 to 12 has lowest number of approval rate i.e. 83%. The project targeted towards kids are much likely to get approved.

1.2.4 Univariate Analysis: project_subject_categories

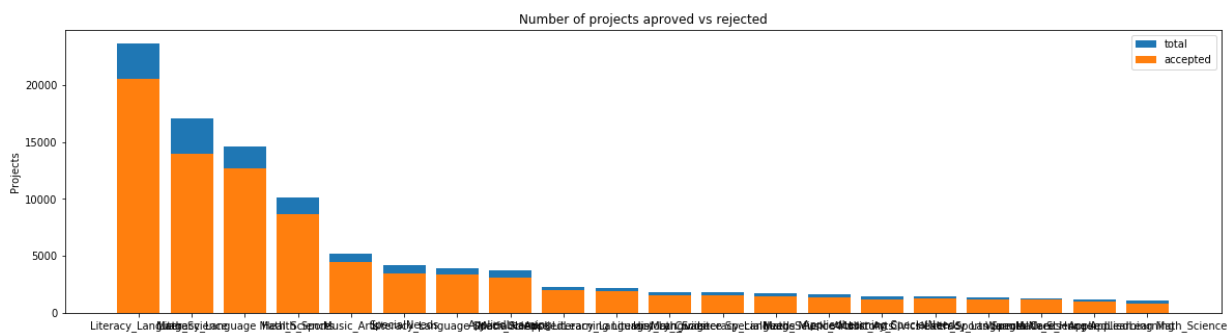
```
In [15]: categories = list(project_data['project_subject_categories'].values)
# remove special characters from list of strings python: https://stackoverflow.com/
# https://www.geeksforgeeks.org/removing-stop-words-nltk-python/
# https://stackoverflow.com/questions/23669024/how-to-strip-a-specific-word-from-a-string
# 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 remove them
        j = j.replace(' ', '') # we are replacing all the ' ' (space) with '' (empty)
        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 [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	school_state	project_sul
0	160221	p253737	c90749f5d961ff158d4b4d1e7dc665fc	Mrs.	IN	20
1	140945	p258326	897464ce9ddc600bcd1151f324dd63a	Mr.	FL	20

```
In [17]: univariate_barplots(project_data, 'clean_categories', 'project_is_approved', top=
```



	clean_categories	project_is_approved	total	Avg
24	Literacy_Language	20519	23654	0.867464
32	Math_Science	13991	17072	0.819529
28	Literacy_Language Math_Science	12723	14634	0.869414
8	Health_Sports	8640	10177	0.848973
40	Music_Arts	4429	5180	0.855019
=====				
	clean_categories	project_is_approved	total	Avg
19	History_Civics Literacy_Language	1271	1421	0.894441
14	Health_Sports SpecialNeeds	1215	1391	0.873472
50	Warmth Care_Hunger	1212	1309	0.925898
33	Math_Science AppliedLearning	1019	1220	0.835246
4	AppliedLearning Math_Science	855	1052	0.812738

Summary

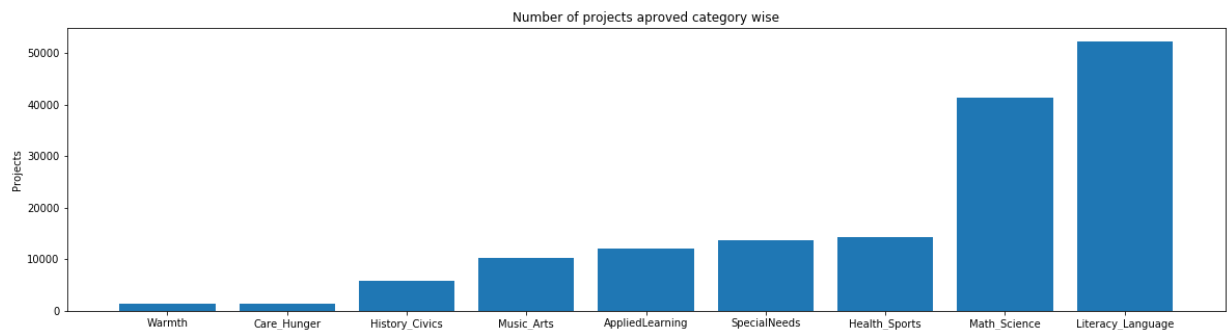
The non stem Subject involving Music & Arts, Literacy & Languages has approval rate of greater than 85%, the standalone stem subjects like Math & Science has relatively lower approval rate. Also, The Project's subject involving social services keywords like care,Hunger etc has much more approval rate i.e. >90%. It can be inferred that Donor's choose platform is more inclined towards arts,literature and Humanity.

```
In [18]: # count of all the words in corpus python: https://stackoverflow.com/a/22898595/4
from collections import Counter
my_counter = Counter()
for word in project_data['clean_categories'].values:
    my_counter.update(word.split())
```

```
In [19]: # 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('Number of projects aproved category wise') #Correction instead of % it
plt.xticks(ind, list(sorted_cat_dict.keys()))
plt.show()
```



Summary:

The most number of projects that has been submitted is from Literacy & Language and the least is from warmth category. The categories Music & Arts, Applied Learning, Special Needs and Health & sports belongs in the range of 10k to 15k sumbissions.

Clearly, we can see the below nummbers according to their project subject categories.

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

```
armth           :      1388
are_Hunger      :      1388
istory_Civics   :      5914
usic_Arts       :     10293
ppliedLearning  :     12135
pecialNeeds     :     13642
ealth_Sports    :     14223
ath_Science     :     41419
iteracy_Language :     52236
```

1.2.5 Univariate Analysis: project_subject_subcategories

```
In [21]: sub_categories = list(project_data['project_subject_subcategories'].values)
# remove special characters from list of strings python: https://stackoverflow.com/

# 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

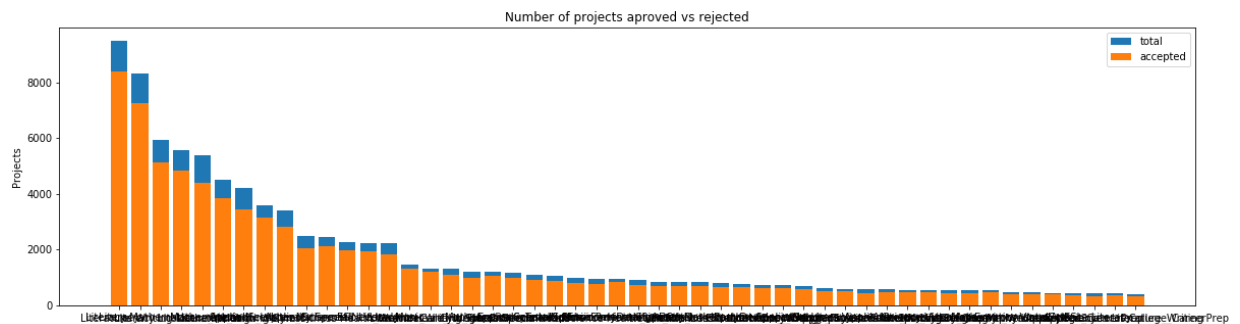
sub_cat_list = []
for i in sub_categories:
    temp = ""
    # consider we have text like this "Math & Science, Warmth, Care & Hunger"
    for j in i.split(','): # it will split it in three parts ["Math & Science", "Warmth", "Care & Hunger"]
        if 'The' in j.split(): # this will split each of the category based on space
            j=j.replace('The', '') # if we have the words "The" we are going to remove them
        j = j.replace(' ', '') # we are replacing all the ' ' (space) with '' (empty)
        temp +=j.strip()+" #" "abc ".strip() will return "abc", remove the trailing spaces
    temp = temp.replace('&', '_')
    sub_cat_list.append(temp.strip())
```

```
In [22]: project_data['clean_subcategories'] = sub_cat_list
project_data.drop(['project_subject_subcategories'], axis=1, inplace=True) #axis=1 means columns
project_data.head(2)
```

```
Out[22]:
```

	Unnamed: 0	id	teacher_id	teacher_prefix	school_state	project_status
0	160221	p253737	c90749f5d961ff158d4b4d1e7dc665fc	Mrs.	IN	2016
1	140945	p258326	897464ce9ddc600bced1151f324dd63a	Mr.	FL	2016

In [23]: `univariate_barplots(project_data, 'clean_subcategories', 'project_is_approved', t`



	clean_subcategories	project_is_approved	total	Avg
317	Literacy	8371	9486	0.882458
319	Literacy Mathematics	7259	8324	0.872057
331	Literature_Writing Mathematics	5139	5922	0.867781
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

Summary

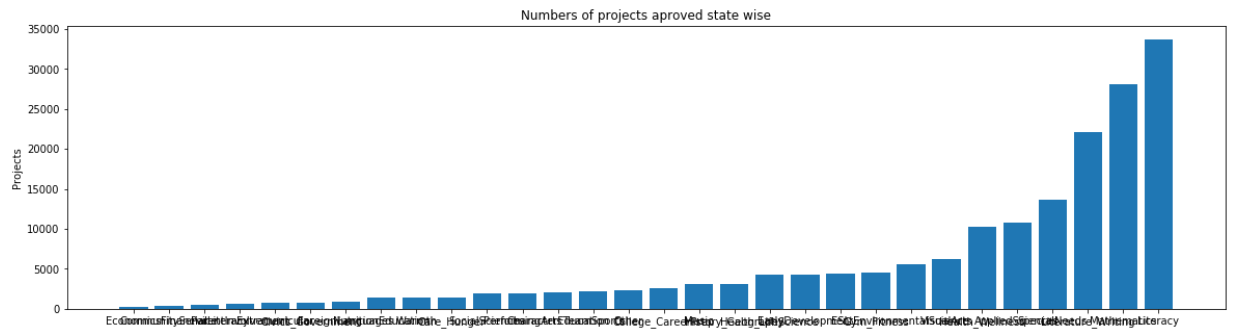
- The approval success rate is more than 80%.
- The highest approval rate is for Literacy. Also, If Literacy is tagged with another sub categories then it's approval rate is more than the standalone approval rate for sub categories. like for Literacy and Mathematics the approval rate is 87% but for mathematics it's 81%.

In [24]: `# count of all the words in corpus python: https://stackoverflow.com/a/22898595/4
from collections import Counter
my_counter = Counter()
for word in project_data['clean_subcategories'].values:
 my_counter.update(word.split())`

```
In [25]: # 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('Numbers of projects aproved state wise') #should be Number instead of
plt.xticks(ind, list(sorted_sub_cat_dict.keys()))
plt.show()
```



Summary

The number of project having subject sub category Literacy is maximum and their approval rate is also highest. The number of projects for economics is least. The Project's subject sub category involving the term Mathematics is second highest.

```
In [26]: for i, j in sorted_sub_cat_dict.items():
         print("{:20} {:10}".format(i,j))
```

```
conomics          :      269
communityService  :      441
financialLiteracy  :      568
parentInvolvement :      677
extracurricular   :      810
 civics_Government :      815
foreignLanguages  :      890
nutritionEducation :     1355
armth             :     1388
are_Hunger        :     1388
ocialSciences     :     1920
performingArts    :     1961
characterEducation :     2065
teamSports        :     2192
ther              :     2372
college_CareerPrep :     2568
usic              :     3145
istory_Geography  :     3171
ealth_LifeScience :     4235
arlyDevelopment   :     4254
SL                :     4367
ym_Fitness        :     4509
nvironmentalScience :     5591
isualArts         :     6278
ealth_Wellness    :    10234
ppliedSciences    :    10816
pecialNeeds       :    13642
iterature_Writing :    22177
athematics        :    28072
iteracy           :   33699
```

Since, Economics has least number of submission i.e. 269, But let's find out how many them have been approved

```
In [27]: #https://stackoverflow.com/questions/8364674/how-to-count-the-number-of-true-elements
#https://stackoverflow.com/questions/19377969/combine-two-columns-of-text-in-dataframe
np.sum(project_data[project_data['project_is_approved']==1]\
       ['clean_subcategories'].str.split().apply(lambda x : re.search(r"\bEconomi
```

Out[27]: 226

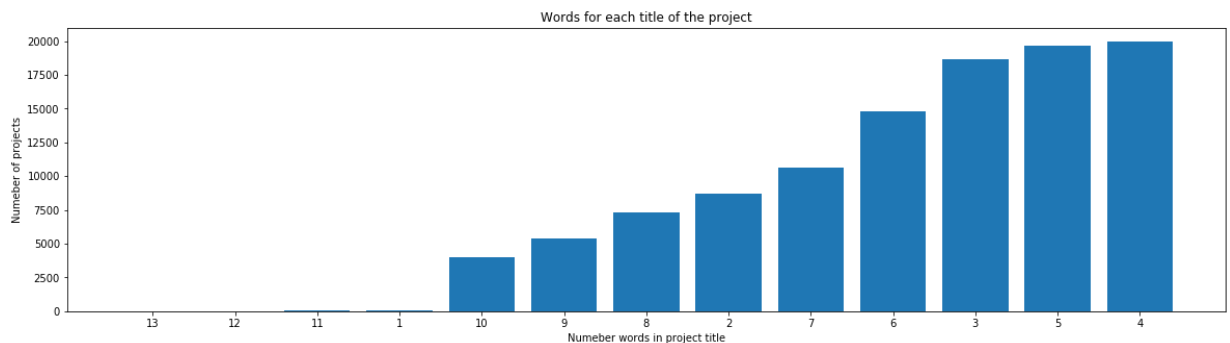
As we clearly see that the term involving Economics has 84% approval rate.

1.2.6 Univariate Analysis: Text features (Title)

```
In [28]: #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('Numeber of projects')
plt.xlabel('Numeber words in project title')
plt.title('Words for each title of the project')
plt.xticks(ind, list(word_dict.keys()))
plt.show()
```



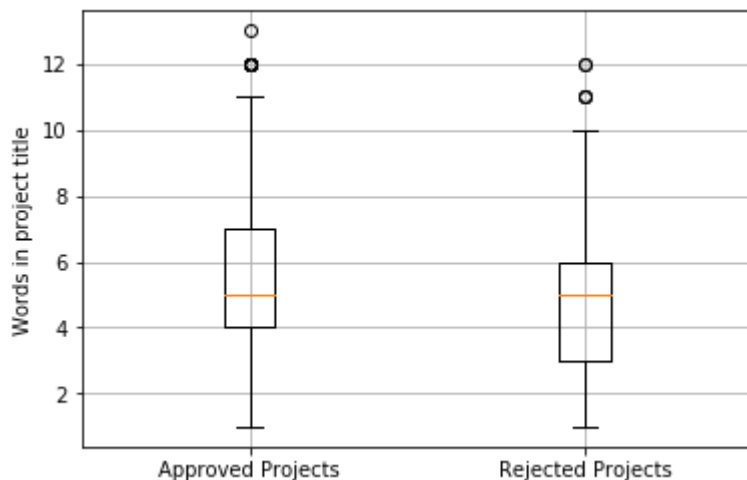
Summary

The most common length for project titles that has been submitted is 4 and least common is 10.
The title having length 5 is 2nd most common among the others.

```
In [29]: approved_title_word_count = project_data[project_data['project_is_approved']==1][
approved_title_word_count = approved_title_word_count.values

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

```
In [30]: # https://glowingpython.blogspot.com/2012/09/boxplot-with-matplotlib.html
plt.boxplot([approved_title_word_count, rejected_title_word_count])
plt.xticks([1,2],('Approved Projects','Rejected Projects'))
plt.ylabel('Words in project title')
plt.grid()
plt.show()
```

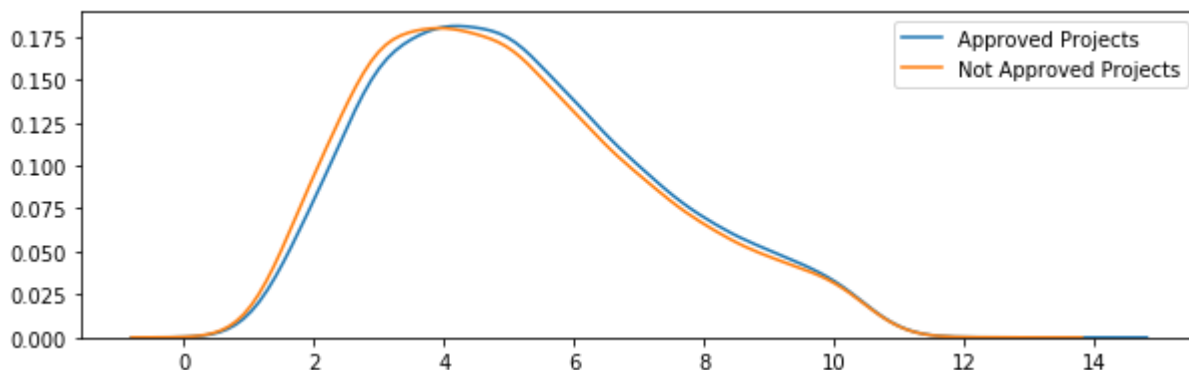


Summary

As we can from box plot that, 50th percentile line (known as median) for both decisions either approval or rejection is almost same i.e. 5.

- For Approved projects, the median (i.e. 5) is closer towards the 25th percentile that means there are few titles which has 5 words or less.
- For Approved projects, there are more titles that has words lie between lengths 5 to 7.
- For Rejected Projects, there are more titles that has words lie between lengths 3 to 5.
- For Rejected Projects, the median (i.e. 5) is closer towards the 75th percentile that means there are few titles which has 5 words or more.

```
In [31]: plt.figure(figsize=(10,3))
sns.kdeplot(approved_title_word_count,label="Approved Projects", bw=0.6)
sns.kdeplot(rejected_title_word_count,label="Not Approved Projects", bw=0.6)
plt.legend()
plt.show()
```



Summary

The KDE plot for both classes is right tailed and they have same distribution. It's very hard to distinguish the class attribute with title features

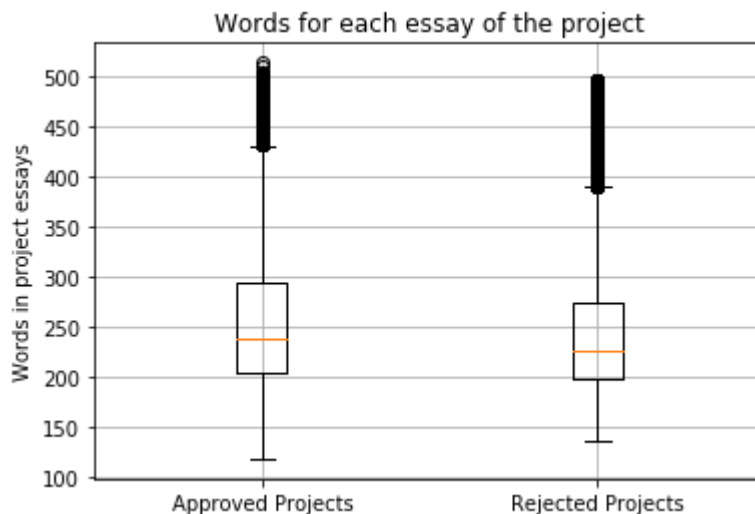
1.2.7 Univariate Analysis: Text features (Project Essay's)

```
In [32]: # 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 [33]: approved_word_count = project_data[project_data['project_is_approved']==1]['essay']
approved_word_count = approved_word_count.values

rejected_word_count = project_data[project_data['project_is_approved']==0]['essay']
rejected_word_count = rejected_word_count.values

In [34]: # https://glowingpython.blogspot.com/2012/09/boxplot-with-matplotlib.html
plt.boxplot([approved_word_count, rejected_word_count])
plt.title('Words for each essay of the project')
plt.xticks([1,2],('Approved Projects','Rejected Projects'))
plt.ylabel('Words in project essays')
plt.grid()
plt.show()
```



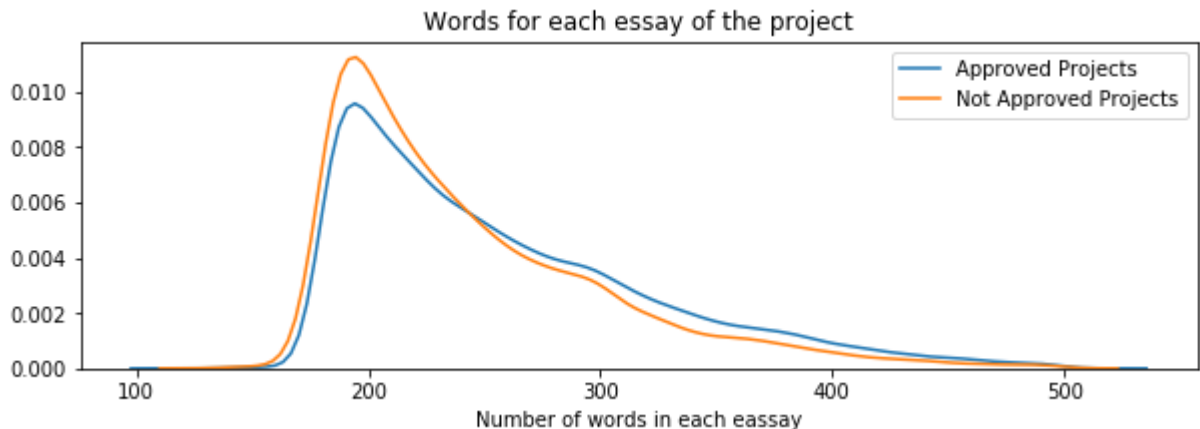
```
In [35]: print("for approved ", np.median(approved_word_count))
print("for rejected", np.median(rejected_word_count))

for approved 239.0
for rejected 226.0
```

Summary

The median for approved project is 239 and for rejected project is 226. Also, We can see that there are maximum number of data points lie between 226 to near 300 for approved projects and for rejected project it's 239 to approx 275. This IQR is also overlapping but not as of previous one and we can see this by plotting density plot as shown below.

```
In [36]: 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()
```



It's also right tailed and they are almost overlapping each other except for non approved project whose peak is taller at around 180-190 words count.

1.2.8 Univariate Analysis: Cost per project

```
In [37]: # we get the cost of the project using resource.csv file
resource_data.head(2)
```

```
Out[37]:
```

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

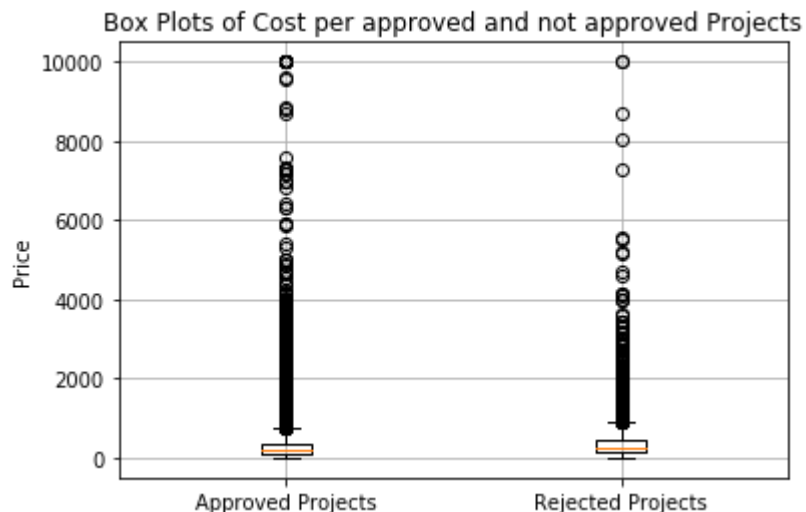
```
Out[38]:
```

	id	price	quantity
0	p000001	459.56	7
1	p000002	515.89	21

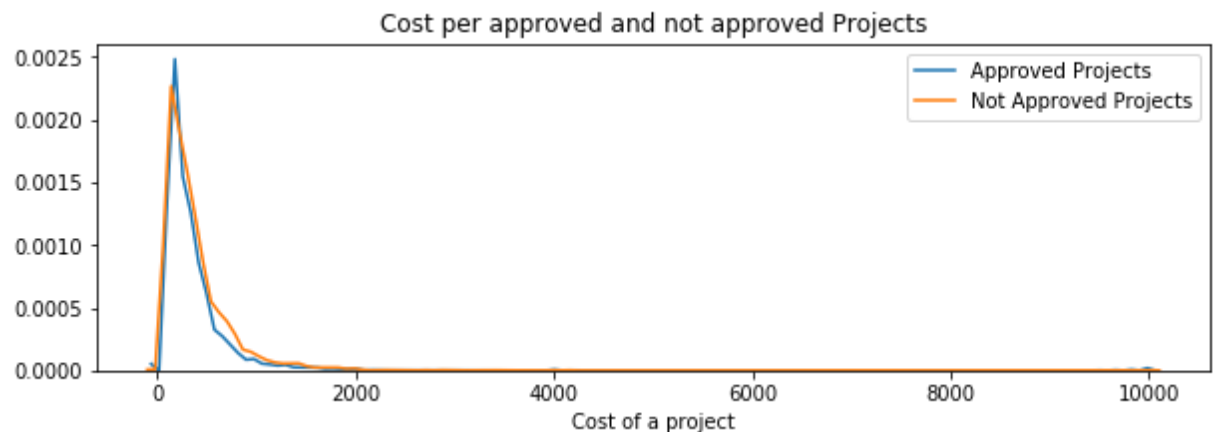
```
In [39]: # join two dataframes in python:
project_data = pd.merge(project_data, price_data, on='id', how='left')
```

```
In [40]: approved_price = project_data[project_data['project_is_approved']==1]['price'].va
rejected_price = project_data[project_data['project_is_approved']==0]['price'].va
```

```
In [41]: # https://glowingpython.blogspot.com/2012/09/boxplot-with-matplotlib.html
plt.boxplot([approved_price, rejected_price])
plt.title('Box Plots of Cost per approved and not approved Projects')
plt.xticks([1,2],('Approved Projects','Rejected Projects'))
plt.ylabel('Price')
plt.grid()
plt.show()
```



```
In [42]: plt.figure(figsize=(10,3))
sns.distplot(approved_price, hist=False, label="Approved Projects")
sns.distplot(rejected_price, hist=False, label="Not Approved Projects")
plt.title('Cost per approved and not approved Projects')
plt.xlabel('Cost of a project')
plt.legend()
plt.show()
```



Summary

From box plot, we can't conclude anything it's all messed up. However I can see the PDF above and can infer that blue line is almost vanishing when it's starts hitting the 10K mark for cost. It means that the higher cost is not supporting the project approval process.

```
In [43]: # http://zetcode.com/python/prettytable/
from prettytable import PrettyTable

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

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	77.374	118.56
25	99.95	140.892
30	116.672	162.23
35	137.207	184.014
40	157.0	208.632
45	178.259	235.106
50	198.99	263.145
55	223.99	292.61
60	255.598	325.144
65	285.41	362.39
70	321.222	399.99
75	366.07	449.945
80	411.666	519.282
85	479.0	618.276
90	593.082	739.356
95	801.494	992.486
100	9999.0	9999.0

Summary

Now the percentile can easily uncover the secrets of cost vs project approval. If we are seeing the IQR range (where the majority of data lies) -

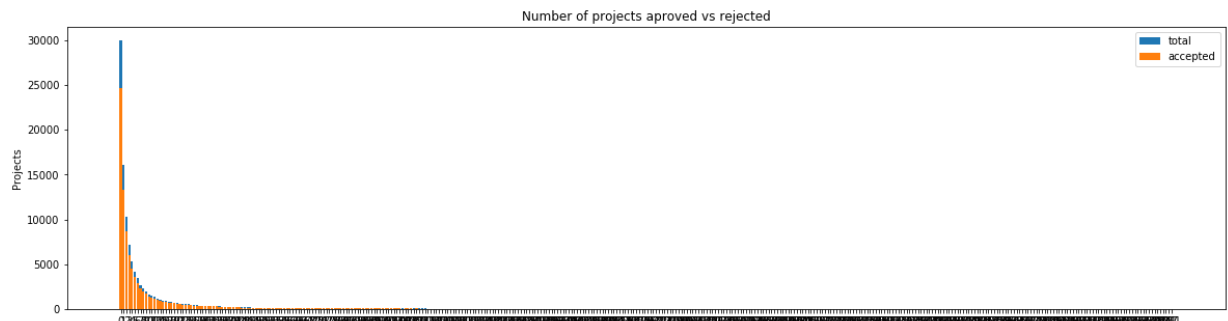
- At 25 %ile, Approved one has almost 40 units lesser than non approved projects.
- At 50 %ile, Approved one has almost 64 units lesser than non approved projects.
- At 75 %ile, Approved one has almost 83 units lesser than non approved projects.

The above table can clearly indicate that for non approved projects costs are higher than approved projects.

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 [44]: *# Plotting the bar plot to to check for approval rates for this variable*
univariate_barplots(project_data, 'teacher_number_of_previously_posted_projects', 'project_is_approved')



teacher_number_of_previously_posted_projects	project_is_approved	total	\
0	24650	30012	
1	13328	16057	
2	8705	10350	
3	5997	7110	
4	4452	5266	

Avg
0.821338
0.830043
0.841063
0.843460
0.845423

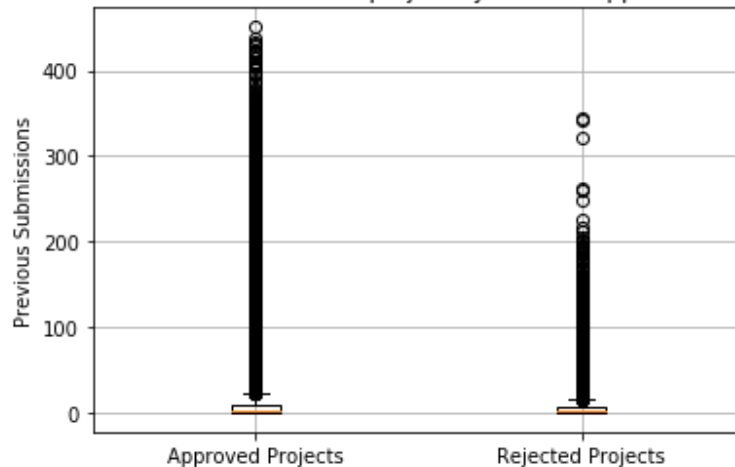
teacher_number_of_previously_posted_projects	project_is_approved	total
42	242	1
68	270	1
34	234	1
35	347	1
73	451	1

Avg
42 1.0
68 1.0
34 1.0
35 1.0
73 1.0

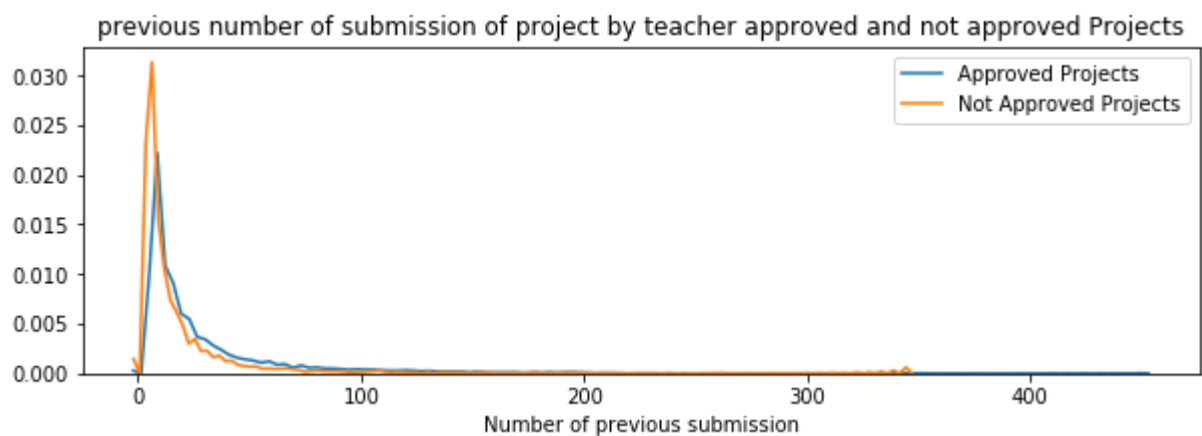
In [45]: approved_numbers = project_data[project_data['project_is_approved']==1]['teacher_number_of_previously_posted_projects']
rejected_numbers = project_data[project_data['project_is_approved']==0]['teacher_number_of_previously_posted_projects']

```
In [46]: # https://glowingpython.blogspot.com/2012/09/boxplot-with-matplotlib.html
plt.boxplot([approved_numbers, rejected_numbers])
plt.title('Box Plots of previous number of submission of project by teacher approved and not approved Projects')
plt.xticks([1,2],('Approved Projects','Rejected Projects'))
plt.ylabel('Previous Submissions')
plt.grid()
plt.show()
```

Box Plots of previous number of submission of project by teacher approved and not approved Projects



```
In [47]: plt.figure(figsize=(10,3))
sns.distplot(approved_numbers, hist=False, label="Approved Projects")
sns.distplot(rejected_numbers, hist=False, label="Not Approved Projects")
plt.title('previous number of submission of project by teacher approved and not approved Projects')
plt.xlabel('Number of previous submission')
plt.legend()
plt.show()
```



```
In [48]: # http://zetcode.com/python/prettytable/
from prettytable import PrettyTable

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

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_numbers,i), 3), np.round(np.percentile(not_approved_numbers,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	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

Summary

The most frequent number of submission is the new submission i.e. approx 30K and 82% of them got approved. 450 is the maximum number of previous submission done on this platform by one teacher. The submission by same teacher for projects count above say 100 is rare. The box plot is not useful here as the data distribution is unreadable as well as unseperable for this features. As From, Percentile Table we can see that both are having same median and almost same distribution. The number of submission is not contributing anything towards the decision of approval of projects on Donors Choose platform. However, this platform looks more encouraging initially (upto 100 submissions) to teachers to post more projects.

1.2.10 Univariate Analysis: project_resource_summary

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

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

```
In [49]: print(project_data['project_resource_summary'].values[0])
print("="*50)
print(project_data['project_resource_summary'].values[100])
print("="*50)
print(project_data['project_resource_summary'].values[150])
print("="*50)
```

y students need opportunities to practice beginning reading skills in English
t home.

=====

y students need laptops that have printing abilities. I would like my students
o have the ability to work on their projects and to print their works, researc
and writings.

=====

y students need 5 Hokki stools to increase their movement even while sitting.

=====

```
In [50]: #https://stackoverflow.com/questions/34962104/pandas-how-can-i-use-the-apply-func
#https://stackoverflow.com/questions/19859282/check-if-a-string-contains-a-number
# considering digits to be non-negative
project_data["digits_in_resource_summary"] = project_data['project_resource_summa
project_data.iloc[150,]
```

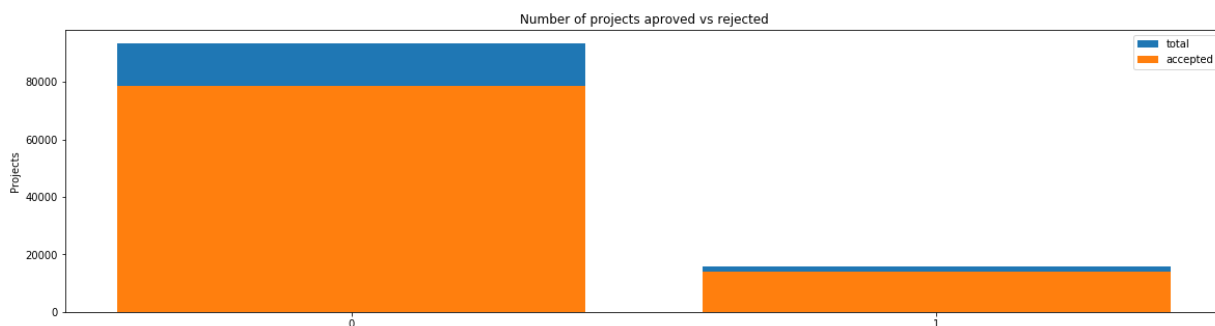
```
Out[50]: nnamed: 0
731
d
142819
each_id eafd3233848365
b7130b83e100434e7
each_prefix
s.
chool_state
0
roject_submitted_datetime 2
16-09-27 23:30:32
roject_grade_category
rades 3-5
roject_title More Movement
ith Hokki Stools
roject_essay_1 The 51 fifth grade students tha
will cycle th...
roject_essay_2 My students will use these five
rightly color...
roject_essay_3
aN
roject_essay_4
aN
roject_resource_summary My students need 5 Hokki stools
o increase th...
each_number_of_previously_posted_projects
6
roject_is_approved

lean_categories
ealth_Sports
lean_subcategories
ealth_Wellness
ssay The 51 fifth grade students tha
will cycle th...
rice
35.3
uantity

igits_in_resource_summary

ame: 150, dtype: object
```

```
In [51]: univariate_barplots(project_data, 'digits_in_resource_summary', 'project_is_approved')
```



digits_in_resource_summary	project_is_approved	total	Avg	
0	0	78614	93490	0.840881
1	1	14089	15755	0.894256

=====

digits_in_resource_summary	project_is_approved	total	Avg	
0	0	78614	93490	0.840881
1	1	14089	15755	0.894256

Summary

We can see that the presence of digits is merely affecting chances of approval of projects by 5%. Also, The volume of projects which doesn't have digits in their summary resource has got fair approval of almost 84%.

1.3 Text preprocessing

1.3.1 Essay Text

```
In [52]: project_data.head(2)
```

Out[52]:

	Unnamed: 0	id	teacher_id	teacher_prefix	school_state	project_sul
0	160221	p253737	c90749f5d961ff158d4b4d1e7dc665fc	Mrs.	IN	20
1	140945	p258326	897464ce9ddc600bced1151f324dd63a	Mr.	FL	20

2 rows × 21 columns


```
In [53]: # 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)
```

y students are English learners that are working on English as their second or hird languages. We are a melting pot of refugees, immigrants, and native-born mericans bringing the gift of language to our school. \r\n\r\n We have over 24 anguaues represented in our English Learner program with students at every lev l of mastery. We also have over 40 countries represented with the families wi hin our school. Each student brings a wealth of knowledge and experiences to s that open our eyes to new cultures, beliefs, and respect.\"The limits of you language are the limits of your world.\"-Ludwig Wittgenstein Our English lea ner's have a strong support system at home that begs for more resources. Many imes our parents are learning to read and speak English along side of their ch ldren. Sometimes this creates barriers for parents to be able to help their c ild learn phonetics, letter recognition, and other reading skills.\r\n\r\nBy p oviding these dvd's and players, students are able to continue their mastery o the English language even if no one at home is able to assist. All families ith students within the Level 1 proficiency status, will be a offered to be a art of this program. These educational videos will be specially chosen by the nglish Learner Teacher and will be sent home regularly to watch. The videos a e to help the child develop early reading skills.\r\n\r\nParents that do not h ve access to a dvd player will have the opportunity to check out a dvd player o use for the year. The plan is to use these videos and educational dvd's for he years to come for other EL students.\r\nnnannan

=====

he 51 fifth grade students that will cycle through my classroom this year all ove learning, at least most of the time. At our school, 97.3% of the students eceive free or reduced price lunch. Of the 560 students, 97.3% are minority st dents. \r\nThe school has a vibrant community that loves to get together and c lebrate. Around Halloween there is a whole school parade to show off the beaut ful costumes that students wear. On Cinco de Mayo we put on a big festival wit crafts made by the students, dances, and games. At the end of the year the sc ool hosts a carnival to celebrate the hard work put in during the school year, ith a dunk tank being the most popular activity. My students will use these fiv brightly colored Hokki stools in place of regular, stationary, 4-legged chair . As I will only have a total of ten in the classroom and not enough for each tudent to have an individual one, they will be used in a variety of ways. Duri g independent reading time they will be used as special chairs students will e ch use on occasion. I will utilize them in place of chairs at my small group t bles during math and reading times. The rest of the day they will be used by t e students who need the highest amount of movement in their life in order to s ay focused on school.\r\n\r\nWhenever asked what the classroom is missing, my tudents always say more Hokki Stools. They can't get their fill of the 5 stool we already have. When the students are sitting in group with me on the Hokki tools, they are always moving, but at the same time doing their work. Anytime he students get to pick where they can sit, the Hokki Stools are the first to e taken. There are always students who head over to the kidney table to get on

of the stools who are disappointed as there are not enough of them. \r\n\r\nWask 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

=====

How do you remember your days of school? Was it in a sterile environment with plain walls, rows of desks, and a teacher in front of the room? A typical day in your room is nothing like that. I work hard to create a warm inviting themed room for my students look forward to coming to each day.\r\n\r\nMy class is made up of 28 wonderfully unique boys and girls of mixed races in Arkansas.\r\nThey attend a Title I school, which means there is a high enough percentage of free and reduced-price lunch to qualify. Our school is an "open classroom" concept, which is very unique as there are no walls separating the classrooms. These 9 and 10 year-old students are very eager learners; they are like sponges, absorbing all the information and experiences and keep on wanting more. With these resources such as the comfy red throw pillows and the whimsical nautical hanging decor and the blue fish nets, I will be able to help create the mood in our classroom setting to be one of a themed nautical environment. Creating a classroom environment is very important in the success in each and every child's education. The nautical photo props will be used with each child as they step foot into your classroom for the first time on Meet the Teacher evening. I'll take pictures of each child with them, have them developed, and then hung in our classroom ready for their first day of 4th grade. This kind gesture will set the tone before even the first day of school! The nautical thank you cards will be used throughout the year by the students as they create thank you cards to their team groups.\r\n\r\nYour generous donations will help me to help make our classroom fun, inviting, learning environment from day one.\r\n\r\nIt costs a lot of money out of my own pocket on resources to get our classroom ready. Please consider helping with this project to make our new school year a very successful one. Thank you!nannan

=====

My wonderful students are 3, 4, and 5 years old. We are located in a small town outside of Charlotte, NC. All of my 22 students are children of school district employees.\r\nMy students are bright, energetic, and they love to learn! They love hands-on activities that get them moving. Like most preschoolers, they enjoy music and creating different things. \r\nAll of my students come from wonderful families that are very supportive of our classroom. Our parents enjoy watching their children's growth as much as we do! These materials will help me teach my students all about the life cycle of a butterfly. We will watch as the Painted Lady caterpillars grow bigger and build their chrysalis. After a few weeks they will emerge from the chrysalis as beautiful butterflies! We already have a net for the chrysalises, but we still need the caterpillars and feeding station.\r\nThis will be an unforgettable experience for my students. My students absolutely love hands-on materials. They learn so much from getting to touch and manipulate different things. The supporting materials I have selected will help my students understand the life cycle through exploration.nannan

=====

The students in my classroom are learners, readers, writers, explorers, scientists, and mathematicians! The potential in these first graders is endless! Each day they come in grinning from ear-to-ear and ready to learn more. \r\nI choose curriculum that is real and relevant to the students, but it will also prepare them for their futures. These kids are encouraged to investigate concepts that are exciting for them and I hope we can keep this momentum going! These kids deserve the best, please help me give that to them! Thank you! :) These kits include

e a wide variety of science, technology, engineering, and mechanics for my students to dive into at the beginning of the year. I want them to hit the ground running this upcoming year and these kits always encourage high interest.\r\nWho wouldn't want to build their own roller coaster, design a car, or even think critically to make a bean bag bounce as far as it can go?? These kits will also show students potential careers that they may have never heard of before!\r\nAny donations would be greatly appreciated and my students will know exactly who to thank for them!\nannan

=====

In [54]: `# 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 [55]: `sent = decontracted(project_data['essay'].values[20000])
print(sent)
print("="*50)`

My wonderful students are 3, 4, and 5 years old. We are located in a small town outside of Charlotte, NC. All of my 22 students are children of school district employees.\r\nMy students are bright, energetic, and they love to learn! They love hands-on activities that get them moving. Like most preschoolers, they enjoy music and creating different things. \r\nAll of my students come from wonderful families that are very supportive of our classroom. Our parents enjoy watching their children's growth as much as we do! These materials will help me teach my students all about the life cycle of a butterfly. We will watch as the Painted Lady caterpillars grow bigger and build their chrysalis. After a few weeks they will emerge from the chrysalis as beautiful butterflies! We already have a net for the chrysalises, but we still need the caterpillars and feeding station.\r\nThis will be an unforgettable experience for my students. My students absolutely love hands-on materials. They learn so much from getting to touch and manipulate different things. The supporting materials I have selected will help my students understand the life cycle through exploration.\nannan

=====

```
In [56]: # \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)
```

y wonderful students are 3, 4, and 5 years old. We are located in a small town outside of Charlotte, NC. All of my 22 students are children of school district employees. My students are bright, energetic, and they love to learn! They love hands-on activities that get them moving. Like most preschoolers, they enjoy music and creating different things. All of my students come from wonderful families that are very supportive of our classroom. Our parents enjoy watching their children's growth as much as we do! These materials will help me teach my students all about the life cycle of a butterfly. We will watch as the Painted Lady caterpillars grow bigger and build their chrysalis. After a few weeks they will emerge from the chrysalis as beautiful butterflies! We already have a net for the chrysalises, but we still need the caterpillars and feeding station. This will be an unforgettable experience for my students. My student absolutely love hands-on materials. They learn so much from getting to touch and manipulate different things. The supporting materials I have selected will help my students understand the life cycle through exploration.nannan

```
In [57]: #remove spacial character: https://stackoverflow.com/a/5843547/4084039
sent = re.sub('[^A-Za-z0-9]+', ' ', sent)
print(sent)
```

y wonderful students are 3 4 and 5 years old We are located in a small town outside of Charlotte NC All of my 22 students are children of school district employees My students are bright energetic and they love to learn They love hands on activities that get them moving Like most preschoolers they enjoy music and reating different things All of my students come from wonderful families that re very supportive of our classroom Our parents enjoy watching their children s growth as much as we do These materials will help me teach my students all a out the life cycle of a butterfly We will watch as the Painted Lady caterpillars grow bigger and build their chrysalis After a few weeks they will emerge from the chrysalis as beautiful butterflies We already have a net for the chrysalises but we still need the caterpillars and feeding station This will be an unforgettable experience for my students My student absolutely love hands on materials They learn so much from getting to touch and manipulate different things Th supporting materials I have selected will help my students understand the lif cycle through exploration nannan

```
In [58]: # 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'll", "you'd", 'your', 'yours', 'yourself', 'yourselves', 'he', 'she', "she's", 'her', 'hers', 'herself', 'it', "it's", 'its', 'itself', 'theirs', 'themselves', 'what', 'which', 'who', 'whom', 'this', 'that', 'am', 'is', 'are', 'was', 'were', 'be', 'been', 'being', 'have', 'has', 'did', 'doing', 'a', 'an', 'the', 'and', 'but', 'if', 'or', 'because', 'at', 'by', 'for', 'with', 'about', 'against', 'between', 'into', 'through', 'above', 'below', 'to', 'from', 'up', 'down', 'in', 'out', 'on', 'off', 'then', 'once', 'here', 'there', 'when', 'where', 'why', 'how', 'all', 'most', 'other', 'some', 'such', 'only', 'own', 'same', 'so', 'than', 's', 't', 'can', 'will', 'just', 'don', "don't", 'should', "should've", 've', 'y', 'ain', 'aren', "aren't", 'couldn', "couldn't", 'didn', "didn't", 'hadn't', 'hasn', "hasn't", 'haven', "haven't", 'isn', "isn't", 'ma', "mustn't", 'needn', "needn't", 'shan', "shan't", 'shouldn', "shouldn't", 'won', "won't", 'wouldn', "wouldn't"]
```

```
In [59]: # Combining all the above statements
# https://stackoverflow.com/questions/42212810/tqdm-in-jupyter-notebook
from tqdm import tqdm_notebook as tqdm
preprocessed_essays = []
# tqdm is for printing the status bar
for sentence in tqdm(project_data['essay'].values):
    sent = decontracted(sentence)
    sent = sent.replace('\\n', ' ')
    sent = sent.replace('\\\"', ' ')
    sent = sent.replace('\\n', ' ')
    sent = re.sub('[^A-Za-z0-9]+', ' ', sent)
    # https://gist.github.com/sebleier/554280
    sent = ' '.join(e for e in sent.split() if e not in stopwords)
    preprocessed_essays.append(sent.lower().strip())
```

HBox(children=(IntProgress(value=0, max=109245), HTML(value='')))

```
In [60]: # after preprocessing
preprocessed_essays[20000]
```

```
Out[60]: 'my wonderful students 3 4 5 years old we located small town outside charlotte
c all 22 students children school district employees my students bright energy
ic love learn they love hands activities get moving like preschoolers enjoy mu
ic creating different things all students come wonderful families supportive c
assroom our parents enjoy watching children growth much these materials help t
ach students life cycle butterfly we watch painted lady caterpillars grow bigg
r build chrysalis after weeks emerge chrysalis beautiful butterflies we alread
net chrysalises still need caterpillars feeding station this unforgettable ex
perience students my student absolutely love hands materials they learn much ge
ting touch manipulate different things the supporting materials i selected hel
students understand life cycle exploration nannan'
```

1.3.2 Project title Text

```
In [61]: # similarly you can preprocess the titles also
print(project_data['project_title'].values[0])
print(project_data['project_title'].values[50])
print(project_data['project_title'].values[100])
print(project_data['project_title'].values[150])
```

Educational Support for English Learners at Home
 Be Active! Be Energized!
 21st Century learners, 21st century technology!
 More Movement with Hokki Stools

```
In [62]: # preprocessing the projec title
preprocessed_title = []
for sentence in tqdm(project_data['project_title'].values):
    sent = decontracted(sentence)
    sent = sent.replace('\\n', ' ')
    sent = sent.replace('\\\"', ' ')
    sent = sent.replace('\\n', ' ')
    sent = re.sub('[^A-Za-z0-9]+', ' ', sent)
    # https://gist.github.com/sebleier/554280
    sent = ' '.join(e for e in sent.split() if e not in stopwords)
    preprocessed_title.append(sent.lower().strip())
```

HBox(children=(IntProgress(value=0, max=109245), HTML(value='')))

```
In [63]: preprocessed_title[1000]
```

```
Out[63]: 'sailing into super 4th grade year'
```

```
In [64]: project_data['project_title'].values[1000]
```

```
Out[64]: 'Sailing Into a Super 4th Grade Year'
```

1. 4 Preparing data for models

```
In [65]: project_data.columns
```

```
Out[65]: 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',
               'digits_in_resource_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/>)

```
In [66]: # 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', 'Health_Sports', 'Math_Science', 'Literacy_Language']
Shape of matrix after one hot encoding (109245, 9)
```

```
In [67]: # 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'])
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 (109245, 30)
```

```
In [68]: # Unique values in school state column i.e. total number of states
project_data.school_state.unique()
```

```
Out[68]: array(['IN', 'FL', 'AZ', 'KY', 'TX', 'CT', 'GA', 'SC', 'NC', 'CA', 'NY', 'OK', 'MA', 'NV', 'OH', 'PA', 'AL', 'LA', 'VA', 'AR', 'WA', 'WV', 'ID', 'TN', 'MS', 'CO', 'UT', 'IL', 'MI', 'HI', 'IA', 'RI', 'NJ', 'MO', 'DE', 'MN', 'ME', 'WY', 'ND', 'OR', 'AK', 'MD', 'WI', 'SD', 'NE', 'NM', 'DC', 'KS', 'MT', 'NH', 'VT'], dtype=object)
```

```
In [69]: # one hot encoding feature encoding for states
# no need to pass the vocabulary as distinguish values are there for every cell
state_vec = CountVectorizer(lowercase=False, binary=True)
state_vec.fit(project_data['school_state'])
print(state_vec.get_feature_names())
state_one_hot = state_vec.transform(project_data['school_state'].values)
print("Shape of matrix after one hot encoding ", 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 encoding (109245, 51)
```

```
In [70]: # one hot encoding for teacher_prefix
# no need to pass the vocabulary as distinguish values are there for every cell
t_prefix_vec = CountVectorizer(lowercase=False, binary=True)
t_prefix_vec.fit(project_data['teacher_prefix'])
print(t_prefix_vec.get_feature_names())
t_prefix_one = t_prefix_vec.transform(project_data['teacher_prefix'].values)
print("Shape of matrix after one hot encoding is ", t_prefix_one.shape)
```

```
['Dr', 'Mr', 'Mrs', 'Ms', 'Teacher']
Shape of matrix after one hot encoding is (109245, 5)
```



```
In [71]: # One Hot encoding for Project Grade category
# Vocabulary need to passed otherwise It will be splitted into Grades, prek-2,num
# Hence vecotrizer needs to know what tokens to choose in vectorization
p_grade_vec = CountVectorizer(vocabulary=list(project_data.project_grade_category
p_grade_vec.fit(project_data.project_grade_category)
print(p_grade_vec.get_feature_names())
p_grade_one = p_grade_vec.transform(project_data.project_grade_category.values)
print("The shape of matrix after one hot encoding is ",p_grade_one.shape)
```

```
['Grades PreK-2', 'Grades 6-8', 'Grades 3-5', 'Grades 9-12']
The shape of matrix after one hot encoding is (109245, 4)
```

1.4.2 Vectorizing Text data

1.4.2.1 Bag of words

```
In [72]: # We are considering only the words which appeared in at least 10 documents(rows
vectorizer = CountVectorizer(min_df=10)
text_bow = vectorizer.fit_transform(preprocessed_essays)
print("Shape of matrix after one hot encodig ",text_bow.shape)
```

```
Shape of matrix after one hot encodig (109245, 16623)
```

1.4.2.2 Bag of Words on project_title

```
In [73]: # Vectorization of title
# We are considering only the words which appeared in at least 10 documents(rows
vectorizer = CountVectorizer(min_df=10) #min_df minimum document frequency
title_bow = vectorizer.fit_transform(preprocessed_title)
print("Shape of matrix after applying BOW on Project title ",title_bow.shape)
```

```
Shape of matrix after applying BOW on Project title (109245, 3329)
```

1.4.2.3 TFIDF vectorizer

```
In [74]: # # We are considering only the words which appeared in at least 10 documents(row
from sklearn.feature_extraction.text import TfidfVectorizer
vectorizer = TfidfVectorizer(min_df=10)
text_tfidf = vectorizer.fit_transform(preprocessed_essays)
print("Shape of matrix after one hot encodig ",text_tfidf.shape)
```

```
Shape of matrix after one hot encodig (109245, 16623)
```

1.4.2.4 TFIDF Vectorizer on project_title

```
In [75]: # Similarly you can vectorize for title also
# We are considering only the words which appeared in at least 10 documents(rows)
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 (109245, 3329)

1.4.2.5 Using Pretrained Models: Avg W2V

```
In [76]: # 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): #1st letter is word and rest of them are vectors values
        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!
```

```
# '''
```

Loading Glove Model

```
HBox(children=(IntProgress(value=1, bar_style='info', max=1), HTML(value='')))
```

Done. 1917495 words loaded!

```
Out[76]: '# =====\nOutput:\n\nLoading Glove Model\n1917495it\n[06:32, 4879.69it/s]\nDone. 1917495 words loaded!\n\n# '
```

```
In [77]: words = []
for i in preprocessed_essays:
    words.extend(i.split(' '))

for i in preprocessed_title:
    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

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

```
all the words in the coupus 17013963
the unique words in the coupus 58966
The number of words that are present in both glove vectors and our coupus 51501
( 87.34 %)
word 2 vec length 51501
```

```
In [78]: # stronging variables into pickle files python: http://www.jessicayung.com/how-to
# 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 [79]: # average Word2Vec
# compute average word2vec for each document in corpus.
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]))
```

HBox(children=(IntProgress(value=0, max=109245), HTML(value='')))

109245

300

1.4.2.6 Using Pretrained Models: AVG W2V on project_title

```
In [80]: # Similarly you can vectorize for title also
# average Word2Vec
# compute average word2vec for each document in corpus.
avg_w2v_vectors_title = []; # 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_vectors_title.append(vector)

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

HBox(children=(IntProgress(value=0, max=109245), HTML(value='')))

109245

300

1.4.2.7 Using Pretrained Models: TFIDF weighted W2V

```
In [81]: # 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 [82]: # average Word2Vec
# compute average word2vec for each review.
tfidf_w2v_vectors = []; # 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
    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
            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.append(vector)

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

HBox(children=(IntProgress(value=0, max=109245), HTML(value='')))

109245
300

1.4.2.9 Using Pretrained Models: TFIDF weighted W2V on project_title

```
In [83]: # tfidf weighted word 2 vec for project_title column
tfidf_model_title = TfidfVectorizer()
tfidf_model_title.fit(preprocessed_title)
# we are converting a dictionary with word as a key, and the idf as a value
dictionary_title = dict(zip(tfidf_model_title.get_feature_names(), list(tfidf_model_title.idf_)))
tfidf_words_title = set(tfidf_model_title.get_feature_names())
```

```
In [84]: # average Word2Vec
# compute average word2vec for each review.
tfidf_w2v_vectors_title = []; # the avg-w2v for each sentence/review is stored in
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_title):
            vec = model[word] # getting the vector for each word
            # here we are multiplying idf value(dictionary[word]) and the tf value
            tf_idf = dictionary_title[word]*(sentence.count(word)/len(sentence.split()))
            vector += (vec * tf_idf) # calculating tfidf weighted w2v
            tf_idf_weight += tf_idf
    if tf_idf_weight != 0:
        vector /= tf_idf_weight
    tfidf_w2v_vectors_title.append(vector)

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

HBox(children=(IntProgress(value=0, max=109245), HTML(value='')))

109245

300

1.4.3 Vectorizing Numerical features

```
In [85]: # check this one: https://www.youtube.com/watch?v=0H0q0cLn3Z4&t=530s
# standardization sklearn: https://scikit-learn.org/stable/modules/generated/sklearn.preprocessing.StandardScaler
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.
# 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 variance
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 : 298.1152448166964, Standard deviation : 367.49642545627506

In [86]: price_standardized

Out[86]: array([[-0.39052147],
[0.00240752],
[0.5952024],
...,
[-0.1582471],
[-0.61242839],
[-0.51215531]])

In [87]: *# we will be doing the standardization of teacher_number_of_previously_posted_projects*
teacher_pp_count = StandardScaler()
teacher_pp_count.fit(project_data.teacher_number_of_previously_posted_projects.values)
print(f"Mean : {teacher_pp_count.mean_[0]}, Standard deviation : {np.sqrt(teacher_pp_count.var_[0])}")
teacher_pp_count_std = teacher_pp_count.transform(project_data.teacher_number_of_previously_posted_projects.values)

:\\Anaconda\\lib\\site-packages\\sklearn\\utils\\validation.py:475: DataConversionWarning:
ata with input dtype int64 was converted to float64 by StandardScaler.

Mean : 11.153462401025218, Standard deviation : 27.77734982798095

:\\Anaconda\\lib\\site-packages\\sklearn\\utils\\validation.py:475: DataConversionWarning:
ata with input dtype int64 was converted to float64 by StandardScaler.

In [88]: teacher_pp_count_std

Out[88]: array([[-0.40153083],
[-0.14952695],
[-0.36553028],
...,
[-0.29352917],
[-0.40153083],
[-0.40153083]])

1.4.4 Merging all the above features

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

In [89]: *# shape of some encoded variables and featured vectors*

```
print(categories_one_hot.shape)
print(sub_categories_one_hot.shape)
print(state_one_hot.shape)
print(t_prefix_one.shape)
print(text_bow.shape)
print(title_bow.shape)
print(price_standardized.shape)
print(teacher_pp_count_std.shape)
```

```
(109245, 9)
(109245, 30)
(109245, 51)
(109245, 5)
(109245, 16623)
(109245, 3329)
(109245, 1)
(109245, 1)
```

In [90]: *# 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((state_one_hot, categories_one_hot, sub_categories_one_hot, t_prefix_one_hot, title_bow, price_standardized, teacher_pp_count_std))
```

In [91]: *# https://stackoverflow.com/questions/11953111/numpy-how-to-filter-matrix-lines*
https://stackoverflow.com/questions/35646908/numpy-shuffle-multidimensional-array
https://stackoverflow.com/questions/21887754/concatenate-two-numpy-arrays-vertically

```
def extract_data(text_vec, n):
    x = hstack((X, text_vec))
    x = x.todense()
    x = np.c_[x, project_data.project_is_approved]
    x_pos = x[project_data.project_is_approved==1, :]
    x_neg = x[project_data.project_is_approved==0, :]
    x_pos = x_pos[:n, :] #filtering 3000 positive reviews
    x_neg = x_neg[:n, :] #filtering out 3000 negative reviews
    print("Shape of text vectors ", x.shape)
    print("Shape of positive reviews matrix ", x_pos.shape)
    print("Shape of negative reviews matrix ", x_neg.shape)
    concat_mat = np.vstack((x_pos, x_neg)) # concatenating positive and negative reviews
    print("Shape of concatenated matrix ", concat_mat.shape)
    # to shuffle the positive and negative reviews
    np.random.shuffle(concat_mat) #inplace operation return none
    return concat_mat
```



```
In [92]: # for bag of words
# 6000 reviews
X_bow_title = extract_data(text_vec=title_bow,n=3000)
X_bow_title.shape
```

```
Shape of text vectors (109245, 3427)
Shape of positive reviews matrix (3000, 3427)
Shape of negative reviews matrix (3000, 3427)
Shape of concatenated matrix (6000, 3427)
```

Out[92]: (6000, 3427)

```
In [93]: # for tfidf
# 6000 reviews
X_tfidf_title = extract_data(text_vec=title_tfidf,n=3000)
X_tfidf_title.shape
```

```
Shape of text vectors (109245, 3427)
Shape of positive reviews matrix (3000, 3427)
Shape of negative reviews matrix (3000, 3427)
Shape of concatenated matrix (6000, 3427)
```

Out[93]: (6000, 3427)

```
In [94]: # for average word 2 vec
# 6000 reviews
X_avg_w2v_title = extract_data(text_vec=np.array(avg_w2v_vectors_title),n=3000)
X_avg_w2v_title.shape
```

```
Shape of text vectors (109245, 398)
Shape of positive reviews matrix (3000, 398)
Shape of negative reviews matrix (3000, 398)
Shape of concatenated matrix (6000, 398)
```

Out[94]: (6000, 398)

```
In [95]: # Need to convert to matrix first for Word 2 vec as we stored in 300 dimension list
# for tfidf word 2 vec
# 6000 reviews
X_tfidf_w2v_title = extract_data(text_vec=np.array(tfidf_w2v_vectors_title),n=3000)
X_tfidf_w2v_title.shape
```

```
Shape of text vectors (109245, 398)
Shape of positive reviews matrix (3000, 398)
Shape of negative reviews matrix (3000, 398)
Shape of concatenated matrix (6000, 398)
```

Out[95]: (6000, 398)

Assignment 2: Apply TSNE

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

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

2.1 TSNE with BOW encoding of project_title feature

```

In [96]: #from MulticoreTSNE import MulticoreTSNE as TSNE
from sklearn.manifold import TSNE
# https://github.com/DmitryUlyanov/Multicore-TSNE

def plot_tsne(param,tsne_data,tsne_label):
    """
    <Doc String>
    param will be perplexity and n_iter passed as list of tuple
    tsne_data is the input data matrix
    tsne_label is the class attribute

    """
    for p,i in param:
        if p is not None and i is not None:
            model = TSNE(n_components=2, random_state =0, perplexity =p, n_iter=i
                #random state is 0 before tsne is an probabilistic Algorithm

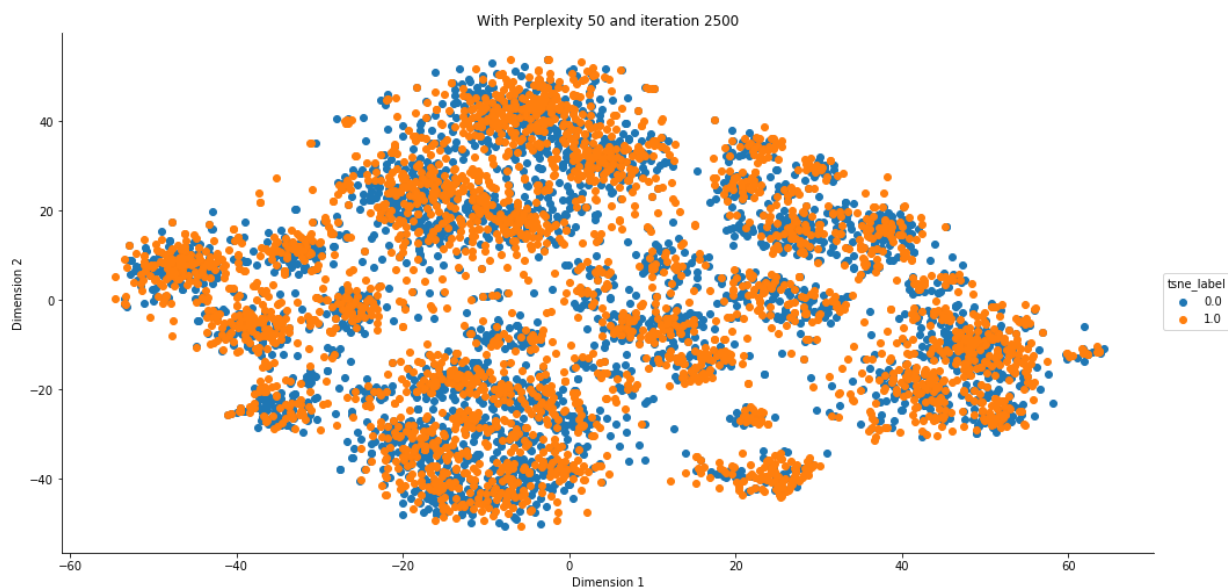
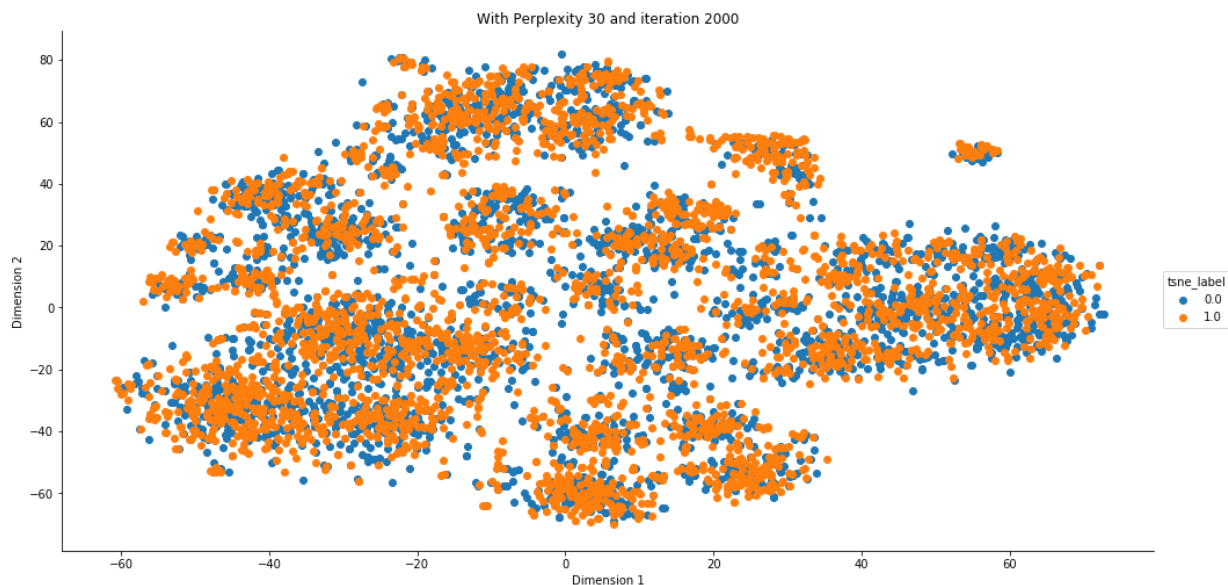
            # n_components is no of dimension it should be reduced to.
            # configuring the parameteres
            # the number of components = 2
            # default perplexity = 30
            # default learning rate = 200
            # default Maximum number of iterations for the optimization = 1000

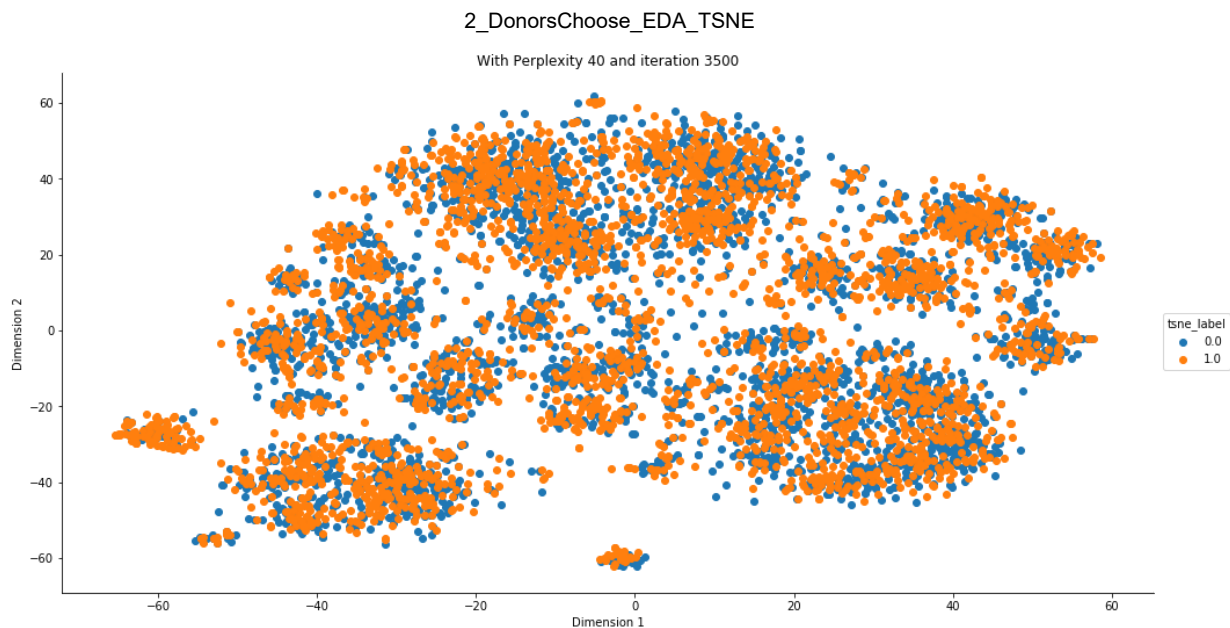
            tsne_fitted_data = model.fit_transform(tsne_data)
            # Visualization
            #concat the fitted data with their corresponding labels
            tsne_fitted_data = np.c_[tsne_fitted_data,tsne_label]
            tsne_df = pd.DataFrame(data = tsne_fitted_data,columns =["Dimension 1",
                "Dimension 2",tsne_label])
            sns.FacetGrid(data = tsne_df,hue='tsne_label',size = 7,aspect=2)\
                .map(plt.scatter,"Dimension 1","Dimension 2").add_legend()
            plt.title("With Perplexity {0} and iteration {1}".format(p,i))
            plt.show()

```

```
In [97]: # Implementation of TSNE with BOW encoding of project_title feature

ip_data = X_bow_title[:,X_bow_title.shape[1]-1]
label = X_bow_title[:, -1]
trial_tsne_values = [(30,2000),(50,2500),(40,3500)] #(perplexity,n_iter)
# executing TSNE and plotting with trial tsne values
plot_tsne(param=trial_tsne_values,tsne_data=ip_data,tsne_label=label)
```

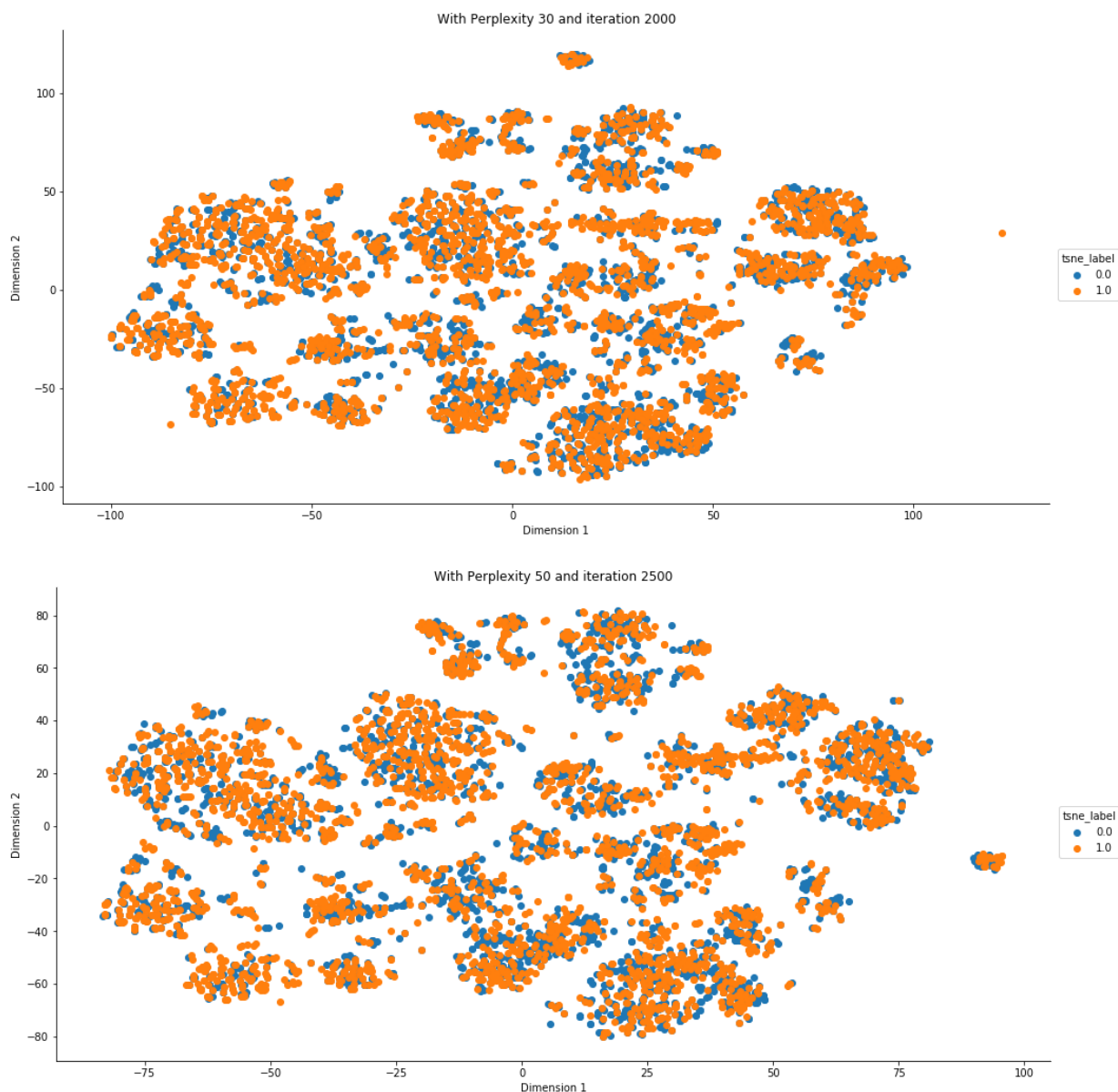


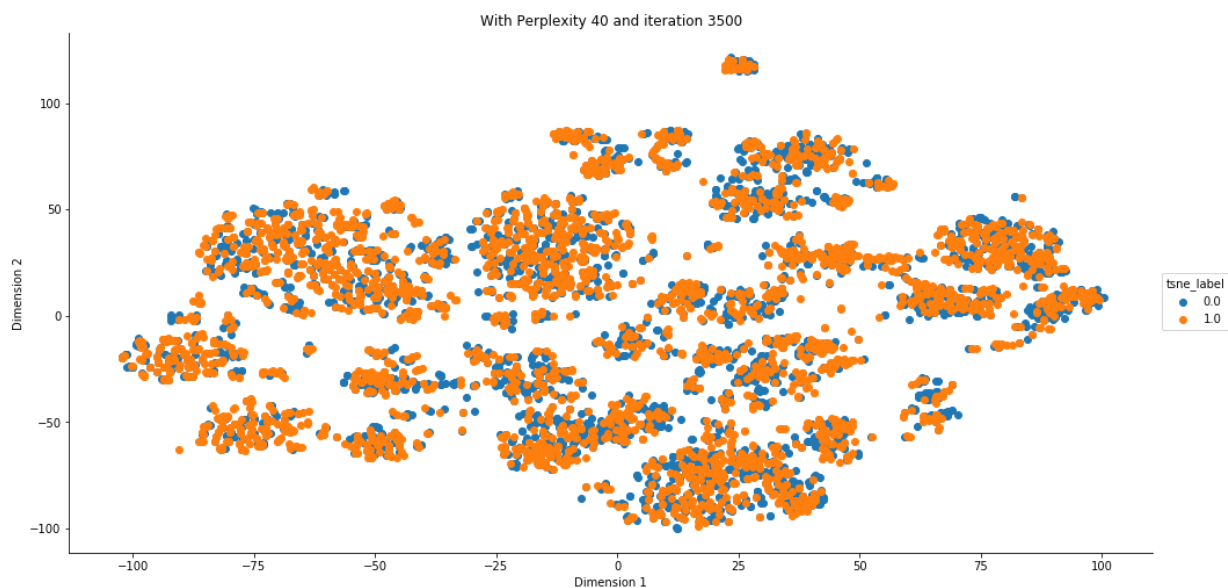


2.2 TSNE with TFIDF encoding of project_title feature

```
In [98]: # Implementation of TSNE with TFIDF encoding of project_title feature

ip_data = X_tfidf_title[:,X_tfidf_title.shape[1]-1]
label = X_tfidf_title[:, -1]
trial_tsne_values = [(30,2000),(50,2500),(40,3500)] #(perplexity,n_iter)
# executing TSNE and plotting with trial tsne values
plot_tsne(param=trial_tsne_values,tsne_data=ip_data,tsne_label=label)
```

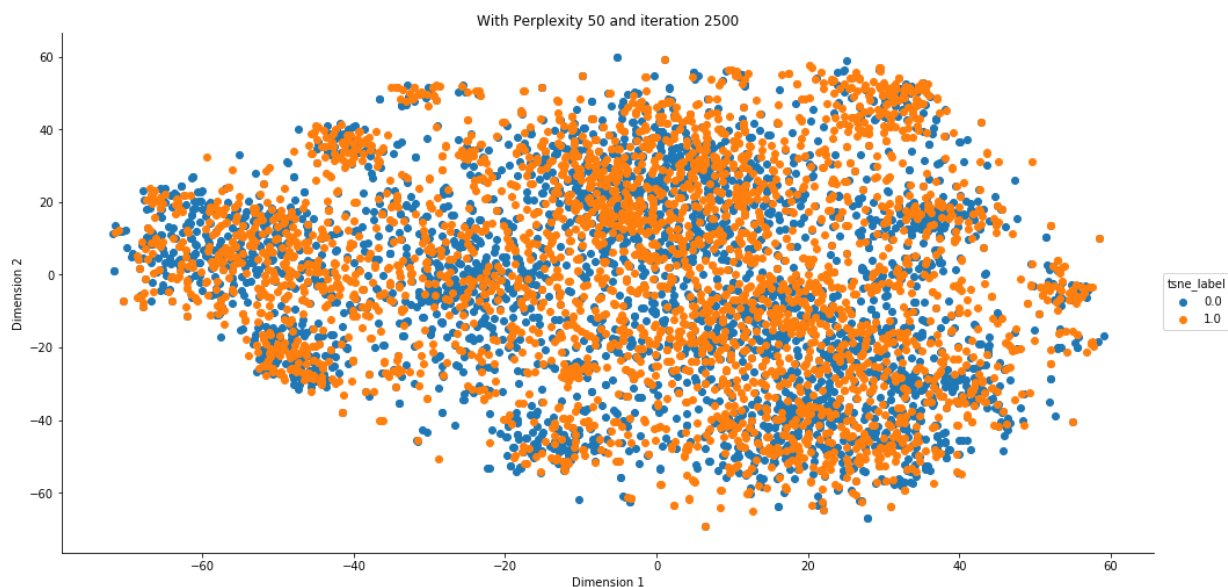
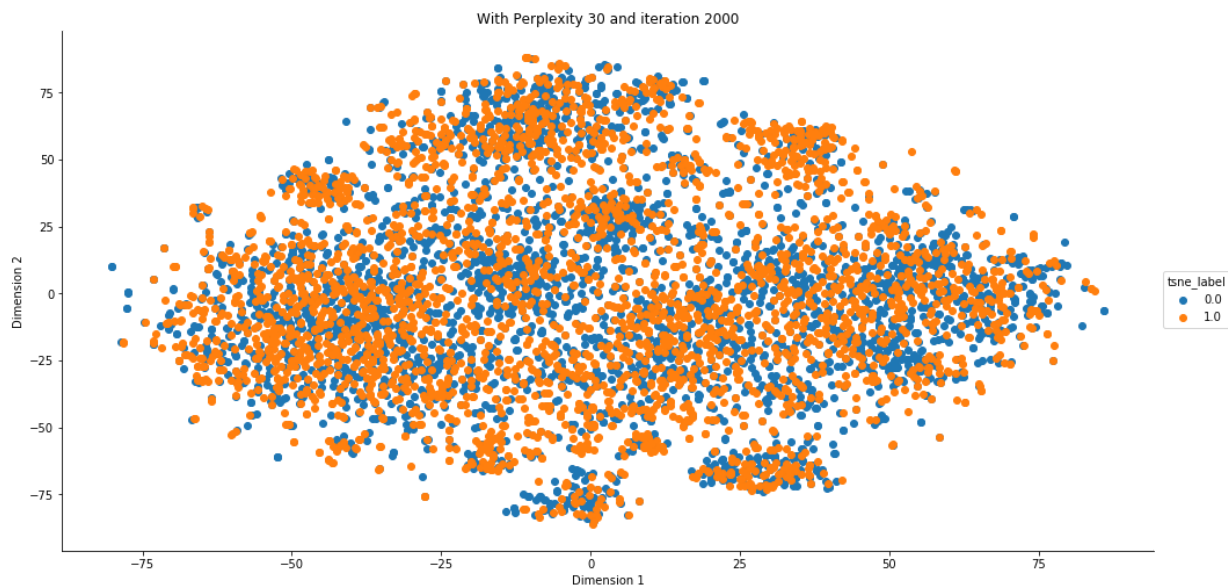


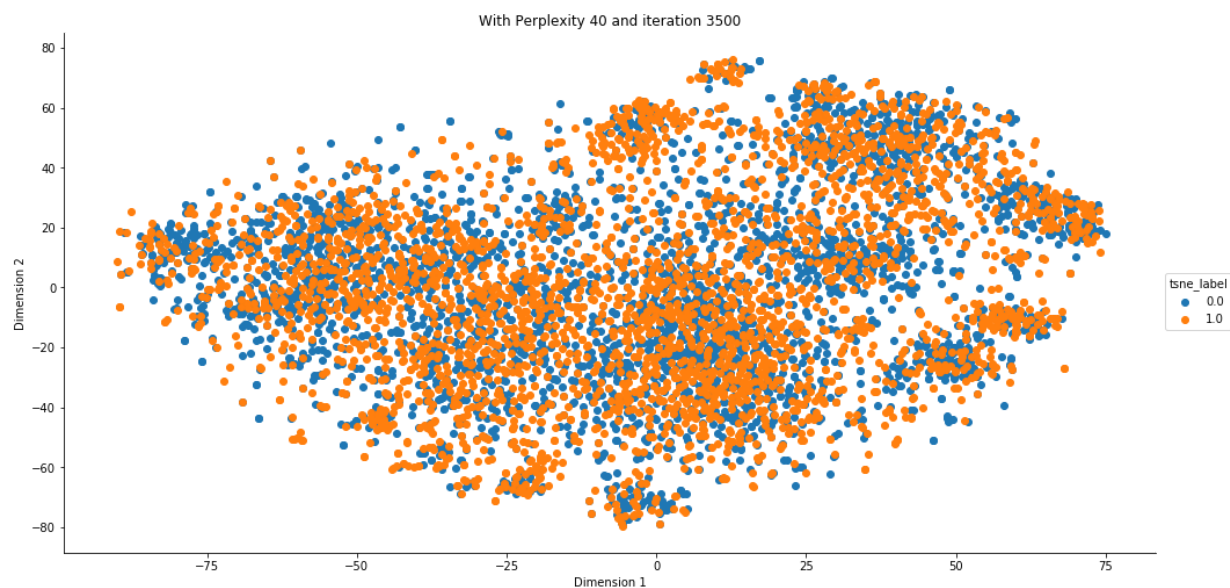


2.3 TSNE with AVG W2V encoding of project_title feature

In [99]: *# Implementation of TSNE with Avg W2V encoding of project_title feature*

```
ip_data = X_avg_w2v_title[:,X_avg_w2v_title.shape[1]-1]
label = X_avg_w2v_title[:, -1]
trial_tsne_values = [(30,2000),(50,2500),(40,3500)] #(perplexity,n_iter)
# executing TSNE and plotting with trial tsne values
plot_tsne(param=trial_tsne_values,tsne_data=ip_data,tsne_label=label)
```

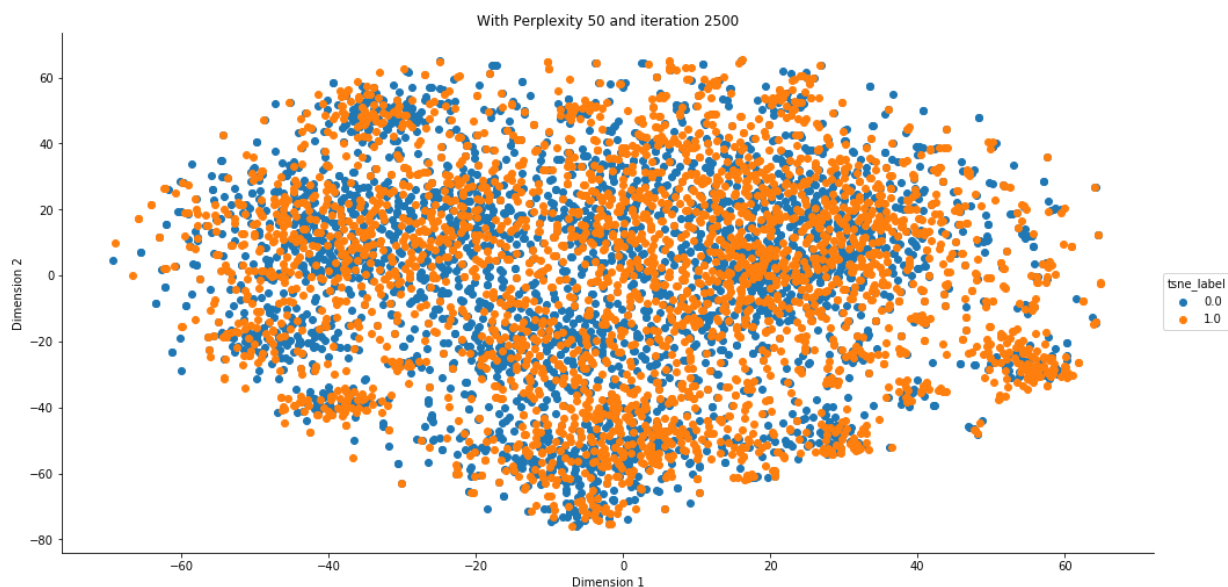
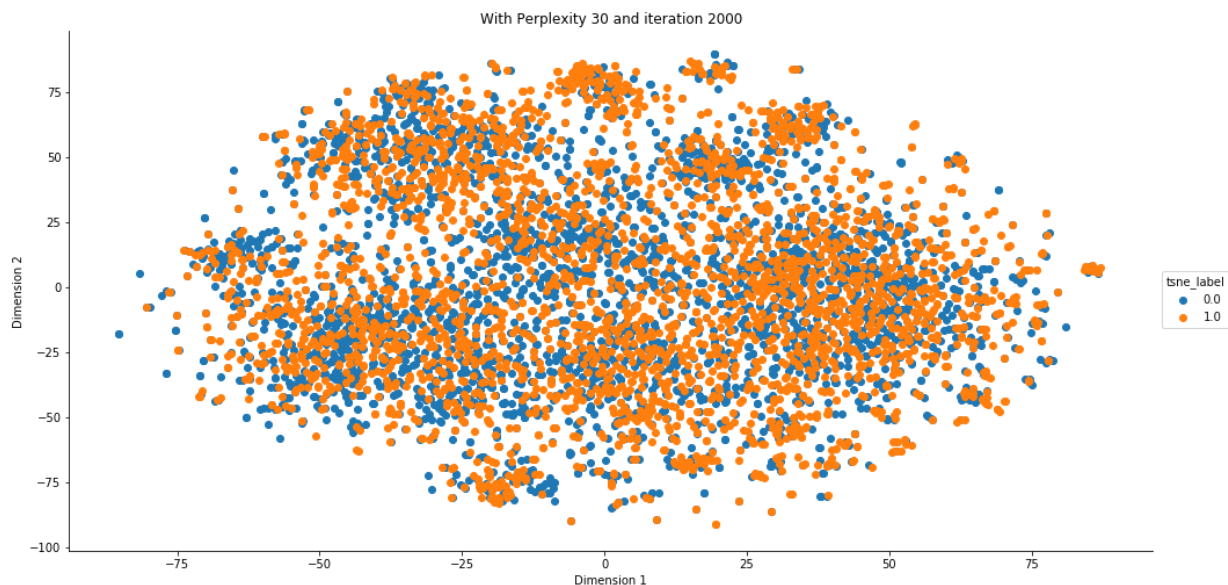


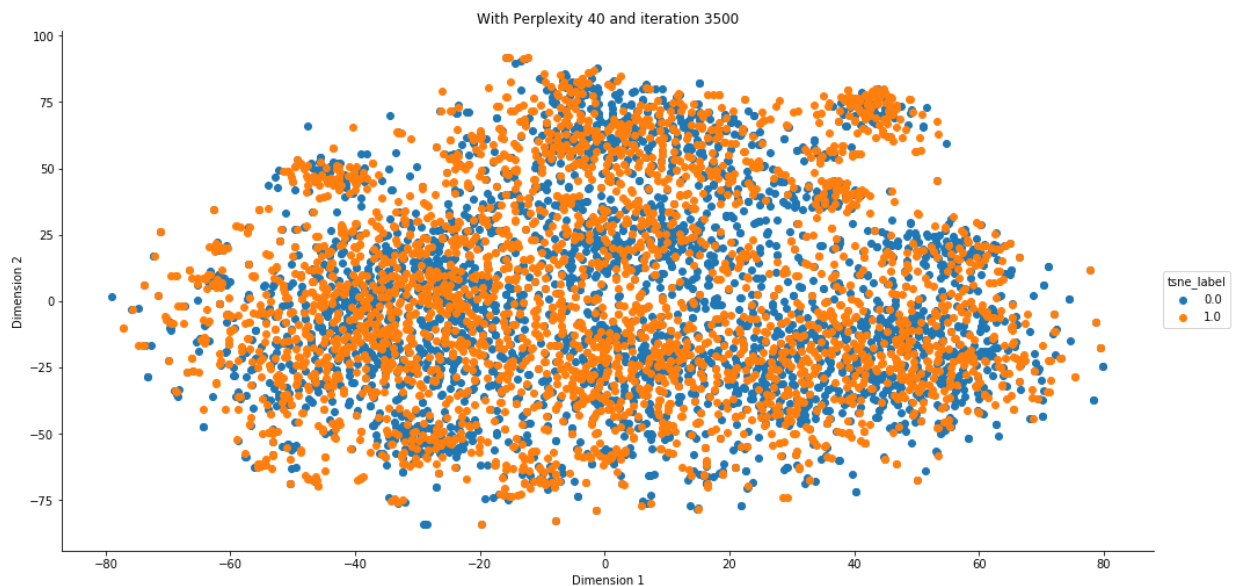


2.4 TSNE with TFIDF Weighted W2V encoding of project_title feature

In [100]: *# Implementation of TSNE with TFIDF Weighted W2V encoding of project_title feature*

```
ip_data = X_tfidf_w2v_title[:,X_tfidf_w2v_title.shape[1]-1]
label = X_tfidf_w2v_title[:, -1]
trial_tsne_values = [(30,2000),(50,2500),(40,3500)] #(perplexity,n_iter)
# executing TSNE and plotting with trial tsne values
plot_tsne(param=trial_tsne_values,tsne_data=ip_data,tsne_label=label)
```





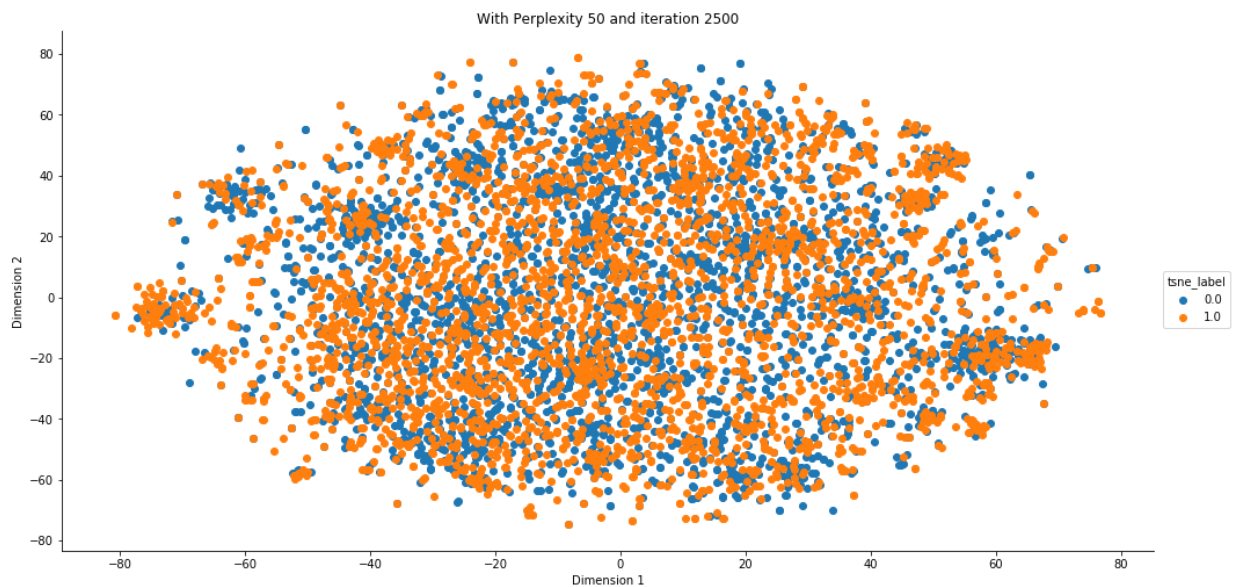
TSNE with all feature combined encoding of project_title feature

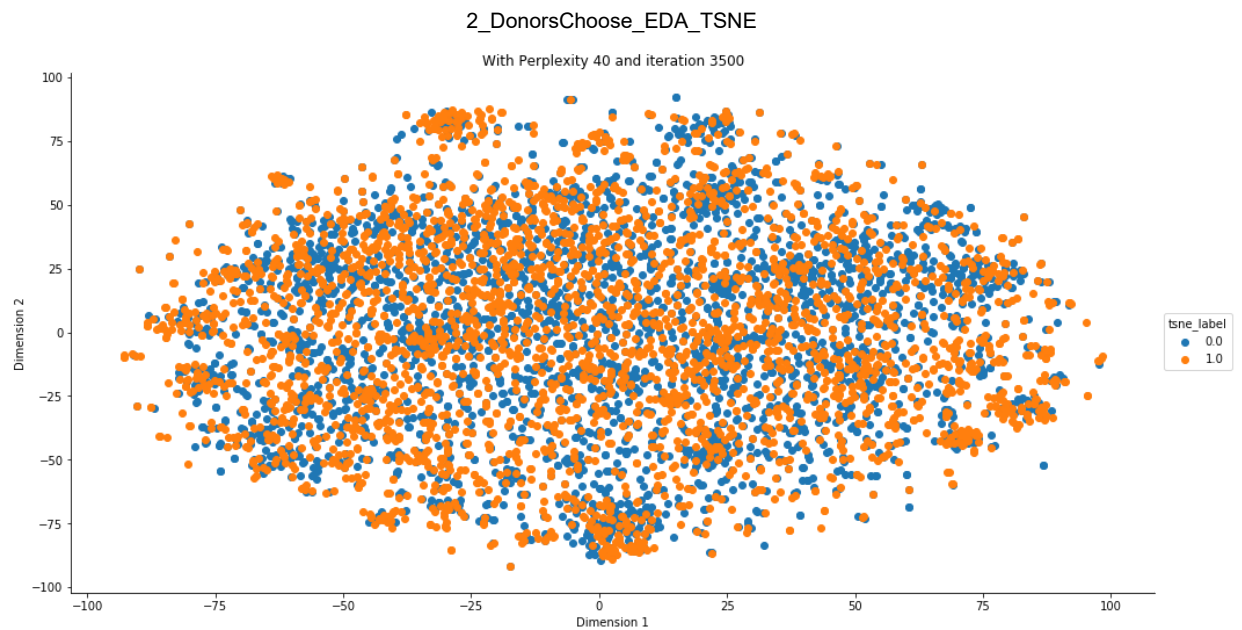
```
In [101]: count=3000 #no of data points
x_all = hstack((X,title_bow,title_tfidf,np.array(avg_w2v_vectors_title),np.array(
x_all = x_all.todense()
x_all = np.c_[x_all,project_data.project_is_approved]
x_pos = x_all[project_data.project_is_approved==1,:]
x_neg = x_all[project_data.project_is_approved==0,:]
x_pos = x_pos[:count,:]
x_neg = x_neg[:count,:]
print("Shape of text vectors ",x_all.shape)
print("Shape of positive reviews matrix ",x_pos.shape)
print("Shape of negative reviews matrix ",x_neg.shape)
x_all_mat = np.vstack((x_pos,x_neg))
print("Shape of concatenated matrix ",x_all_mat.shape)
np.random.shuffle(x_all_mat) #inplace operation return none
```

```
Shape of text vectors (109245, 7356)
Shape of positive reviews matrix (3000, 7356)
Shape of negative reviews matrix (3000, 7356)
Shape of concatenated matrix (6000, 7356)
```

In [103]: *# Implementation of TSNE with all features(BOW,TFIDF,Avg W2V, TFIDF W2V) encoding*

```
ip_data = x_all_mat[:,x_all.shape[1]-1]
label = x_all_mat[:, -1]
trial_tsne_values = [(30,2000),(50,2500),(40,3500)] #(perplexity,n_iter)
# executing TSNE and plotting with trial tsne values
plot_tsne(param=trial_tsne_values,tsne_data=ip_data,tsne_label=label)
```





2.5 Summary

- The features are not separable in hyper dimension also. As they are inseparable in 2D embedded space.
- There are almost no variation of shape and separation at these hyper parameters setting.