DONORS CHOOSE (EDA-TSNE)

Objective: To predict whether or not a DonorsChoose.org project proposal submitted by a teacher will be approved.

1.1 Importing the packages:

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 Reading the data:

```
In [2]:
```

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

Project data (train.csv)

```
In [3]:
#Printing shape and columns of project_data
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' 's
chool_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]:
#printing the first and last date, uptill which the data is in project_data
print(project_data.project_submitted_datetime.min())
print(project_data.project_submitted_datetime.max())
2016-04-27 00:27:36
2017-04-30 23:45:08
In [5]:
project_data.shape
Out[5]:
(109248, 17)
```

Note:

- 1. The project data contains 109248 entries, data from 2016-04-27 to 2017-04-30.
- 2. The project data has 17 attributes regarding project.

Resource Data (resource.csv)

In [6]:

```
#Printing the shape and columns of Resource data
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[6]:

	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

Note:

- 1. resource_data had contained 12344 duplicate entries.
- 2. Some projects has requested for multiple number of resource (number of project < number of resources)
- 3. The Resource data has 1541272 entries that contians the description, quantity and price of resource(s) required by the project.

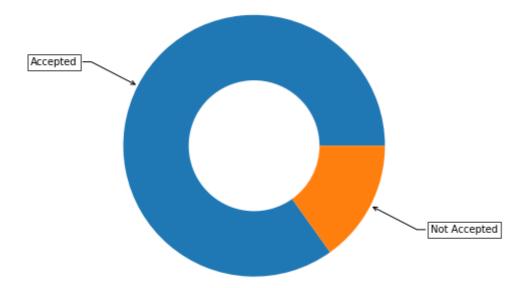
1.3 Data Anaysis

In [7]:

```
# https://matplotlib.org/gallery/pie and polar charts/pie and donut labels.html#sphx-qlr-ga
y_value_counts = project_data['project_is_approved'].value_counts()
print("Number of projects thar are approved for funding ", y_value_counts[1], ", (", (y_val
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=0)
bbox_props = dict(boxstyle="square,pad=0.3", fc="w", ec="k", lw=0.72)
kw = dict(xycoords='data', textcoords='data', arrowprops=dict(arrowstyle="->"),
          bbox=bbox props, zorder=0, va="center")
for i, p in enumerate(wedges):
    ang = (p.theta2 - p.theta1)/2. + p.theta1
    y = np.sin(np.deg2rad(ang))
    x = np.cos(np.deg2rad(ang))
    horizontalalignment = {-1: "right", 1: "left"}[int(np.sign(x))]
    connectionstyle = "angle,angleA=0,angleB={}".format(ang)
    kw["arrowprops"].update({"connectionstyle": connectionstyle})
    ax.annotate(recipe[i], xy=(x, y), xytext=(1.35*np.sign(x), 1.4*y),
                 horizontalalignment=horizontalalignment, **kw)
ax.set title("Nmber of projects that are Accepted and not accepted")
plt.show()
```

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

Nmber of projects that are Accepted and not accepted



Note:

- 1. project data has more data of the projects approved, hence it is an imbalance dataset.
- 2. Approximately 85% of the projects are approved for funding.
- 3. Nearly 15% of projects are NOT approved for funding.

1.4 Univariate Analysis

1.4.1 School State

In [8]:

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

temp = pd.DataFrame(project_data.groupby("school_state")["project_is_approved"].apply(np.metemp.columns = ['state_code', 'num_proposals']
```

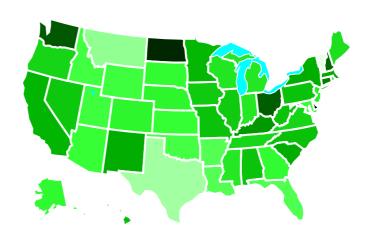
In [9]:

```
# How to plot US state heatmap: https://datascience.stackexchange.com/a/9620
scl = [[0.0, 'rgb(230, 255, 230)'], [0.2, 'rgb(128, 255, 128)'], [0.4, 'rgb(51, 255, 51)'], \
            [0.6, 'rgb(0, 179, 0)'], [0.8, 'rgb(0, 77, 0)'], [1.0, 'rgb(0, 0, 0)']]
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(0, 255, 255)',
        ),
    )
```

In [10]:

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

Project Proposals % of Acceptance Rate



In [11]:

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

```
state_code num_proposals
46
                 0.800000
         VT
         DC
                 0.802326
7
                 0.813142
43
         TX
26
         ΜT
                 0.816327
                 0.831245
18
         LA
_____
States with highest % approvals
  state_code num_proposals
30
         NH
                 0.873563
35
         OH
                 0.875152
47
         WΑ
                 0.876178
         ND
                 0.888112
28
```

0.897959

Note:

8

- 1. Vermont, District of Columbia, Texas, Montana and Louisiana are the states with least percentage of project approvals ranging 80%-83%.
- 2. New Hampshire, Ohio, Washington, North Dakota and Delaware are the stated with high percentage of project approvals ranging 87%-90% approx.
- 3. The average approval rate is 85%.

Function for Stack Plot

DE

In [12]:

```
#stacked bar plots matplotlib: https://matplotlib.org/gallery/lines_bars_and_markers/bar_st
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()
```

Function for Univariate Bar plot

In [13]:

```
def univariate_barplots(data, col1, col2='project_is_approved', top=False):
    # Count number of zeros in dataframe python: https://stackoverflow.com/a/51540521/40840
    temp = pd.DataFrame(project_data.groupby(col1)[col2].agg(lambda x: x.eq(1).sum())).rese

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

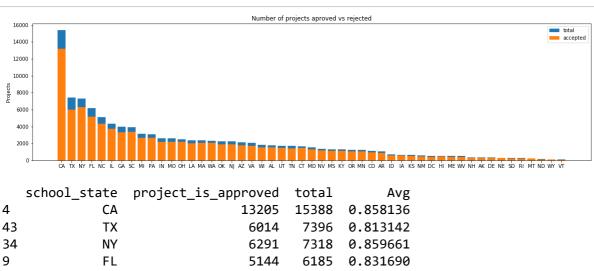
temp.sort_values(by=['total'],inplace=True, ascending=False)

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 [14]:

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



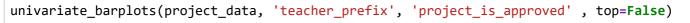
34	INY	6291	/318	0.859661
9	FL	5144	6185	0.831690
27	NC	4353	5091	0.855038
===		=======================================	======	=====
	school_state	<pre>project_is_approved</pre>	total	Avg
39	RI	243	285	0.852632
26	MT	200	245	0.816327
28	ND	127	143	0.888112
50	WY	82	98	0.836735
46	VT	64	80	0.800000

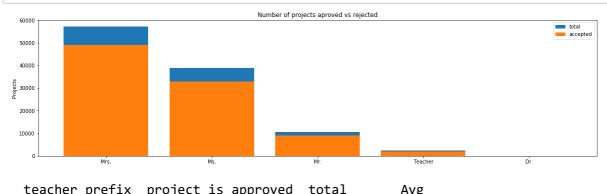
Note:

- 1. California has the highest total number of projects with the approval percentage of ~86%.
- 2. North Dakota has relatively very less total number of projects with the highest approval percentage of ~89%.
- 3. Every state has greater than 80% of project approval rate.

1.4.2 teacher_prefix

In [15]:





	teacher_prefix	<pre>project_is_approved</pre>	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	teacher_prefix Mrs.	project_is_approved 48997	total 57269	Avg 0.855559
2 3	- -			J
_	Mrs.	48997	57269	0.855559
_	Mrs. Ms.	48997 32860	57269 38955	0.855559 0.843537

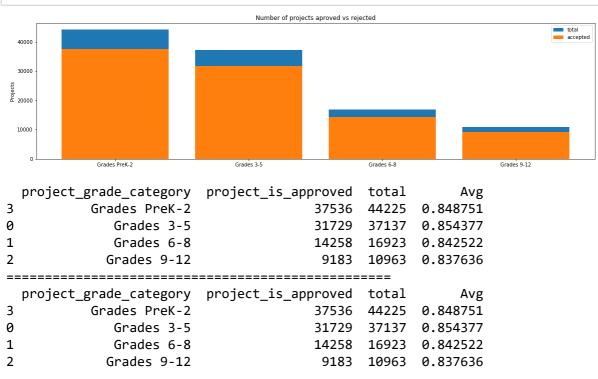
Note:

- 1. The total number of projects proposed by Teacher_prefixes (Mrs. and Ms.) are reltively much higher than others.
- 2. The teacher_prefix(Mrs) has the highest success rate of 85.55%.
- 3. The total number of projects proposed and the success rate(i.e 69%) is least with the teacher_prefix(Dr.).

1.4.3 project_grade_category

In [16]:

#univariate bar plots matplotlib: https://matplotlib.org/gallery/lines_bars_and_markers/bar
univariate_barplots(project_data, 'project_grade_category', 'project_is_approved', top=Fals



Note:

- 1. Project approval rate for grades 3-5 is highest(~85%) and least for grades 9-12(84%).
- 2. There is no siginficant difference between the success rates of different grade catgories, the success rate range from 83.7% to 85.4%..
- 3. The total number of projects for grades 9-12 is relativey much lesser than others.

1.4.4 project_subject_categories

In [17]:

In [18]:

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

Out[18]:

	Unnamed: 0	id	teacher_id	teacher_prefix	school_state	project
0	160221	p253737	c90749f5d961ff158d4b4d1e7dc665fc	Mrs.	IN	
1	140945	p258326	897464ce9ddc600bced1151f324dd63a	Mr.	FL	
4						

In [19]:

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

Number of projects aproved vs rejected

```
total accepted
 20000
 15000
 5000
                   clean_categories
                                      project_is_approved
                                                             total
                                                                          Avg
24
                  Literacy_Language
                                                     20520
                                                             23655
                                                                    0.867470
32
                       Math_Science
                                                     13991
                                                             17072
                                                                    0.819529
    Literacy_Language Math_Science
28
                                                             14636
                                                                    0.869432
                                                     12725
8
                      Health_Sports
                                                      8640
                                                             10177
                                                                    0.848973
40
                         Music Arts
                                                      4429
                                                              5180
                                                                    0.855019
                     clean_categories project_is_approved total
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
```

In [20]:

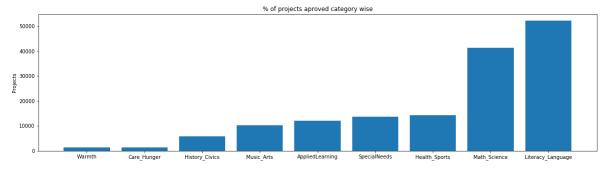
```
# count of all the words in corpus python: https://stackoverflow.com/a/22898595/4084039
from collections import Counter
my_counter = Counter()
for word in project_data['clean_categories'].values:
    my_counter.update(word.split())
```

In [21]:

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

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

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



In [22]:

```
#categories in incresing order of approval rate
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 52239 Literacy_Language

Note:

- 1. The projects of categories 'Literacy and Language' and 'Maths and Science' are higher in number.
- 2. The project of category 'Warmth Care and Hunger' and 'History and civis' has the least approval rate.
- 3. Each project category has greater than 80% of acceptance rate.

1.4.5 project_subject_subcategories

In [23]:

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	school_state	project <u></u>
0	160221	p253737	c90749f5d961ff158d4b4d1e7dc665fc	Mrs.	IN	
1	140945	p258326	897464ce9ddc600bced1151f324dd63a	Mr.	FL	
4						

In [25]:

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

```
Number of projects aproved vs rejected
                clean_subcategories
                                     project_is_approved
                                                          total
                                                                      Avg
317
                                                           9486
                           Literacy
                                                    8371
                                                                 0.882458
319
               Literacy Mathematics
                                                           8325
                                                    7260
                                                                 0.872072
331
    Literature_Writing Mathematics
                                                    5140
                                                           5923
                                                                 0.867803
        Literacy Literature_Writing
318
                                                    4823
                                                           5571
                                                                 0.865733
                        Mathematics
                                                           5379
                                                                 0.815207
342
                                                    4385
clean_subcategories project_is_approved
                                                                          Αv
g
          EnvironmentalScience Literacy
196
                                                         389
                                                                444
                                                                     0.87612
6
127
                                    ESL
                                                         349
                                                                421
                                                                     0.82897
9
79
                     College_CareerPrep
                                                         343
                                                                421
                                                                     0.81472
7
17
    AppliedSciences Literature_Writing
                                                         361
                                                                420
                                                                     0.85952
4
    AppliedSciences College_CareerPrep
3
                                                         330
                                                                405
                                                                     0.81481
5
```

In [26]:

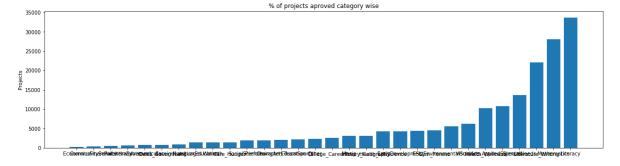
```
# count of all the words in corpus python: https://stackoverflow.com/a/22898595/4084039
from collections import Counter
my_counter = Counter()
for word in project_data['clean_subcategories'].values:
    my_counter.update(word.split())
```

In [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 aproved category wise')
plt.xticks(ind, list(sorted_sub_cat_dict.keys()))
plt.show()
```



In [28]:

```
# sub_categories in increasing order of approval rate
for i,j in sorted_sub_cat_dict.items():
    print("{:20} :{:10}".format(i,j))
```

Economics 269 CommunityService 441 FinancialLiteracy 568 ParentInvolvement 677 Extracurricular 810 Civics_Government : 815 ForeignLanguages 890 NutritionEducation 1355 Warmth 1388 Care_Hunger 1388 SocialSciences 1920 PerformingArts 1961 CharacterEducation 2065 TeamSports 2192 0ther 2372 College_CareerPrep 2568 Music 3145 History_Geography 3171 Health LifeScience : 4235 EarlyDevelopment 4254 **ESL** 4367 Gym_Fitness 4509 EnvironmentalScience : 5591 VisualArts 6278 Health_Wellness 10234 AppliedSciences 10816 SpecialNeeds 13642 Literature_Writing : 22179 Mathematics 28074 Literacy 33700

Note:

1. The project of Sub-category, Literacy'has accepted projects higher in number.

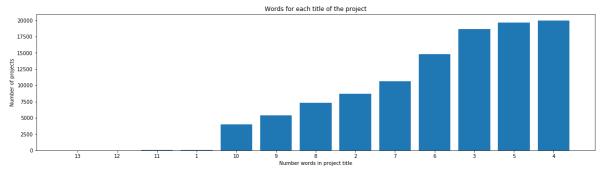
1.4.6 Text features (Title)

In [29]:

```
#How to calculate number of words in a string in DataFrame: https://stackoverflow.com/a/374
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()
```



Note:

- 1. Most of the project titles are of 3-6 words that means most of the titles justified project motives in 3-6 words
- 2. Very few of the project titles are of 1 word or 11 words.

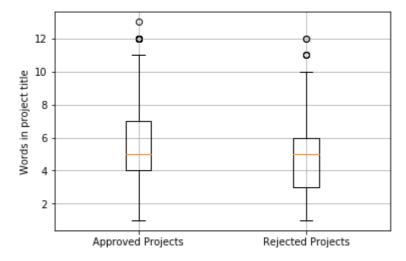
In [30]:

```
approved_title_word_count = project_data[project_data['project_is_approved']==1]['project_t
approved_title_word_count = approved_title_word_count.values

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

In [31]:

```
# 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()
```

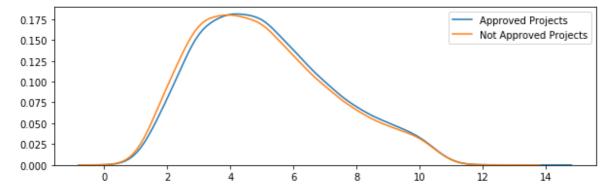


Note:

- 1. Most of the projects that got approved has 6 or more word in its title.
- 2. Both approved and non-approved projects has equal average number of words in their titles.

In [32]:

```
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()
```



Note:

- 1. This plot is not much informative, we cannot draw any strong conclusion from this plot.
- 2. approved title word count is slightly greater than rejected title word count.

1.4.7 Text features (Project Essay's)

In [33]:

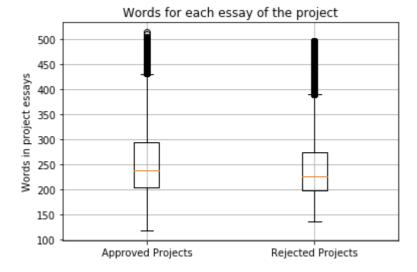
In [34]:

```
approved_word_count = project_data[project_data['project_is_approved']==1]['essay'].str.spl
approved_word_count = approved_word_count.values

rejected_word_count = project_data[project_data['project_is_approved']==0]['essay'].str.spl
rejected_word_count = rejected_word_count.values
```

In [35]:

```
# 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()
```

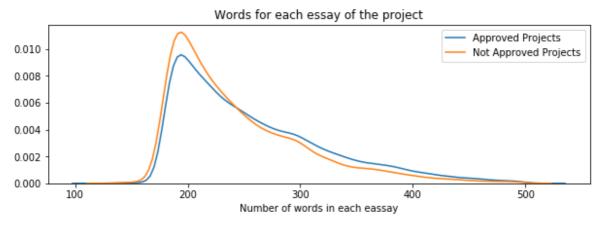


Note:

- 1. No strong conclusion can be drawn from the above plot.
- 2. Approved word count is slightly higher than rejected word count.

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 eassay')
plt.legend()
plt.show()
```



Note:

- 1. Approved word count is bit higher than rejected word count.
- 2. Projects with higher number words in its essay are more likely to get approved.

1.4.8 Cost per project

In [37]:

```
# we get the cost of the project using resource.csv file
resource_data.head(5)
```

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
2	p069063	Cory Stories: A Kid's Book About Living With Adhd	1	8.45
3	p069063	Dixon Ticonderoga Wood-Cased #2 HB Pencils, Bo	2	13.59
4	p069063	EDUCATIONAL INSIGHTS FLUORESCENT LIGHT FILTERS	3	24.95

In [38]:

```
# https://stackoverflow.com/questions/22407798/how-to-reset-a-dataframes-indexes-for-all-gr
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]:

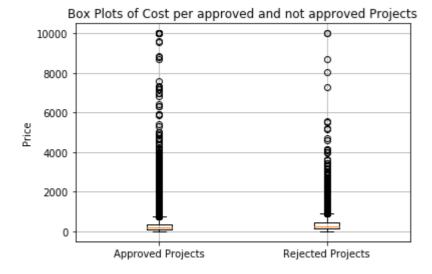
```
# join two dataframes in python: https://www.datacamp.com/community/tutorials/joining-dataf
project_data = pd.merge(project_data, price_data, on='id', how='left')
```

In [40]:

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

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()
```



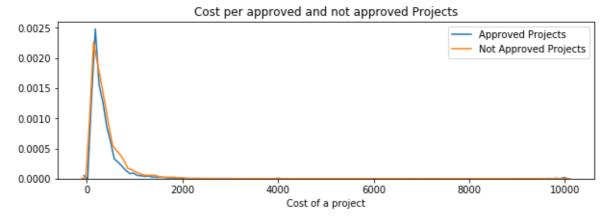
Note:

- 1. No strong conclusion can be drawn from the above plot.
- 2. Number of otliers are very high.

3. cost of project is not a good dimension to estimate the approval rate.

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()
```



Note:

- 1. No strong conclusion can be drawn from the above plot.
- 2. Cheaper projects are showing slightly higher approval rate.

In [43]:

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

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

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

Percentile	Approved Projects	Not Approved Projects
,	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

1.4.9 teacher_number_of_previously_posted_projects (Assignment Part)

In [44]:

univariate_barplots(project_data, 'teacher_number_of_previously_posted_projects', 'project_

```
Number of projects aproved vs rejected
 30000
                                                                              total accepted
 2000
E 15000
 10000
  5000
   teacher_number_of_previously_posted_projects project_is_approved
0
                                                  0
                                                                     24652
                                                                             30014
1
                                                  1
                                                                     13329
                                                                            16058
2
                                                  2
                                                                      8705
                                                                            10350
3
                                                  3
                                                                      5997
                                                                              7110
4
                                                  4
                                                                      4452
                                                                              5266
        Avg
   0.821350
0
   0.830054
1
2
   0.841063
3
   0.843460
   0.845423
4
_____
    teacher_number_of_previously_posted_projects project_is_approved
15
                                                  15
                                                                        818
                                                                                942
16
                                                  16
                                                                        769
                                                                                894
17
                                                  17
                                                                        712
                                                                                803
18
                                                  18
                                                                                772
                                                                        666
19
                                                  19
                                                                        632
                                                                                710
         Avg
15
    0.868365
    0.860179
16
17
    0.886675
18
    0.862694
19
    0.890141
```

Note:

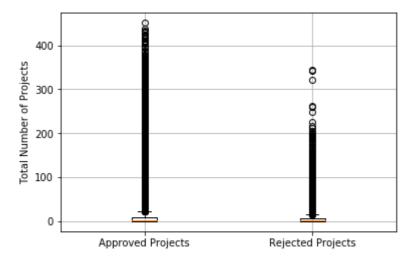
- 1. The approval rate is more that 80% for each number of previously posted projects by the teachers.
- 2. We can't say that the teacher with more number of previously posted projects has higher approval rate although the teacher which has previously posted 48 project has the approval rate of 96%.

In [45]:

```
approval = project_data[project_data['project_is_approved']==1]['teacher_number_of_previous rejection = project_data[project_data['project_is_approved']==0]['teacher_number_of_previous reject_is_approved']==0]['teacher_number_of_previous reject_is_approved']=0]['teacher_number_of_previous reject_is_approved']=0]['teacher_number_of_previous reject_is_approved']=0]['teacher_number_of_previous reject_is_approved']=0]['teacher_number_of_previous reject_is_approved']=0]['teacher_number_of_previous reject_is_approved']=0]['teacher_number_of_previous reject_is_ap
```

In [46]:

```
plt.boxplot([approval, rejection])
plt.xticks([1,2],('Approved Projects','Rejected Projects'))
plt.ylabel('Total Number of Projects')
plt.grid()
plt.show()
```

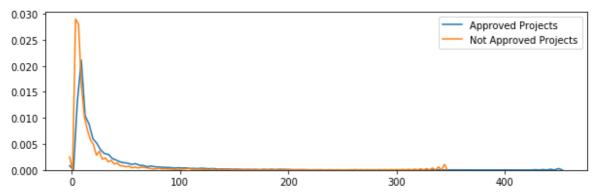


Note:

1. No strong conclusion can be drawn from the above plot.

In [47]:

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



Note:

1. At some extent there is a gradual increase in approval rate with increase of number of previously posted projects.

2. No strong conclusion can be drawn from the above plot.

1.4.10 project_resource_summary (Assignment Part)

In [48]:

```
#How to check if a string contains a number, https://stackoverflow.com/questions/19859282/c
import re
def hasNumbers(inputString):
    return bool(re.search(r'\d',inputString))

resource_summary=list(project_data['project_resource_summary'].values)
has_digits = []
for i in resource_summary:
    if (hasNumbers(i)==True):
        has_digits.append(1)
    else:
        has_digits.append(0)

project_data['summary_digits']=has_digits
```

In [49]:

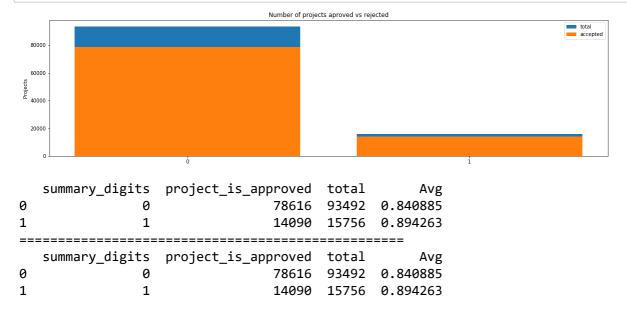
project_data.head(5)

Out[49]:

	Unnamed: 0	id	teacher_id	teacher_prefix	school_state	project _.
0	160221	p253737	c90749f5d961ff158d4b4d1e7dc665fc	Mrs.	IN	
1	140945	p258326	897464ce9ddc600bced1151f324dd63a	Mr.	FL	
2	21895	p182444	3465aaf82da834c0582ebd0ef8040ca0	Ms.	AZ	
3	45	p246581	f3cb9bffbba169bef1a77b243e620b60	Mrs.	KY	
4	172407	p104768	be1f7507a41f8479dc06f047086a39ec	Mrs.	TX	
5 rc	ows × 21 cc	olumns				

In [50]:

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



Note:

- 1. Very few (14% approx) projects has digits in its resource summary.
- 2. The number of projects that contains digits in its resource summary has reatively higher approval rate than those doesn't have digits in its resource summary.
- 3. approval rates: (3.a) project resource that has digits: 89.4%, (3.b) project resource that doesn't have digits: 84%

1.5 Text Preprocessing

In [51]:

```
#Random Sampling of data, considering 5k random entries
# https: https://pandas.pydata.org/pandas-docs/stable/reference/api/pandas.Dataframe.smaple
approved_project = project_data[project_data['project_is_approved']==1].sample(n=5000, rand rejected_project = project_data[project_data['project_is_approved']==0].sample(n=5000, rand project_data = pd.concat([approved_project, rejected_project])
```

1.5.1 Essay Text

In [52]:

project_data.head(2)

Out[52]:

	Unnamed: 0	id	teacher_id	teacher_prefix	school_state	pr
35239	36406	p164226	18b42d3d4237f28ebe0145704b571ad3	Mrs.	СТ	
66180	13120	p230144	83690eedd9ec0b0aef1d247a13fc9385	Ms.	NJ	

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])
```

There is always anticipation and excitement on the first day of the new scho ol year. Everything is new and exciting for my very young students. \r\n\r\n My kindergarten students attend school in a very large urban district in Tex as where resources are very limited. \r\n\r\nMy very young students come to school very eager and excited to learn reading, writing, math and how to ge t along with others. It is easy to remember that wonderful feeling of starti ng school with new friends. \r\n\r\n\r\nThe ink and paper requested will be used to create customized homework packets for my kindergarten students. \r\n Each school year I make an effort to provide customized and differentia ted homework for each of my students. For example, when school starts, I pr int their full names where they trace the dotted lines. Once they have mast ered this, then I hand write each of my students' full names on special writ ing paper and use it as a \"master\" copy. Each week I make copies from thi s master copy as part of their homework. As time passes, I am able to see m ajor improvement in their printing skills. In addition, I use the ink and p aper to print words then sentences as part of their daily homework. Math ho mework sheets are also printed with these resources. \r\n Creating a customi zed homework package takes a lot of planning, time, ink and paper! The stic ky wall pads will be used to model daily homework expectations.nannan

Students enter class eager to learn. Technology engages young minds to make learning exciting. "We need technology in every classroom and in every stude nt and teacher's hand, because it is the pen and paper of our time, and it i s the lens through which we experience much of our world." -D. Warlick\r\nTh e students I teach come from different socioeconomic and ethnic background s.\r\nI teach second graders at a Title One school. I am fortunate to have a diverse group of second grade students. At our school, we have a high povert y level. All of our students receive free or reduced-price lunch. Several st udents receive resource and ELL services. One thing all my students have in common is that they enjoy learning with technology. My students will use the Chromebooks regularly. They will use them for math, language arts, science a nd social studies. Students will work research topics with the Chromebooks a nd use them to enhance their learning. With the Chromebooks, the students w ould be able to utilize more technology. These Chromebooks will help students The Chromebooks will be used by my second grade students. They will use them as they work in collaborative groups. Students will use them acros

s subject areas. For example with social studies, students will use them to research important historical figures like Dr. Martin Luther King Jr. and Ab raham Lincoln. Students will work in their groups to create digital present ations on their historical figure to share with others. In science, student s will use the Chromebooks in collaborative groups to study weather pattern s. In reading, students will use the Chromebooks to listen to leveled books and create digital retells. In math, students will use the Chromebooks to f ocus addition/subtraction practice. These are only a few ways that the Chromebooks will be used in my classroom.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"\'re", " are", phrase)
    phrase = re.sub(r"\'s", " is", phrase)
    phrase = re.sub(r"\'d", " would", phrase)
    phrase = re.sub(r"\'d", " will", phrase)
    phrase = re.sub(r"\'t", " not", phrase)
    phrase = re.sub(r"\'t", " have", phrase)
    phrase = re.sub(r"\'ve", " have", phrase)
    phrase = re.sub(r"\'ve", " am", phrase)
    return phrase
```

```
In [55]:
```

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

Students enter class eager to learn. Technology engages young minds to make learning exciting. "We need technology in every classroom and in every stude nt and teacher's hand, because it is the pen and paper of our time, and it i s the lens through which we experience much of our world." -D. Warlick\r\nTh e students I teach come from different socioeconomic and ethnic background s.\r\nI teach second graders at a Title One school. I am fortunate to have a diverse group of second grade students. At our school, we have a high povert y level. All of our students receive free or reduced-price lunch. Several st udents receive resource and ELL services. One thing all my students have in common is that they enjoy learning with technology. My students will use the Chromebooks regularly. They will use them for math, language arts, science a nd social studies. Students will work research topics with the Chromebooks a nd use them to enhance their learning. With the Chromebooks, the students w ould be able to utilize more technology. These Chromebooks will help students learn. The Chromebooks will be used by my second grade students. They will use them as they work in collaborative groups. Students will use them acros s subject areas. For example with social studies, students will use them to research important historical figures like Dr. Martin Luther King Jr. and Ab raham Lincoln. Students will work in their groups to create digital present ations on their historical figure to share with others. In science, student s will use the Chromebooks in collaborative groups to study weather pattern In reading, students will use the Chromebooks to listen to leveled books and create digital retells. In math, students will use the Chromebooks to f ocus addition/subtraction practice. These are only a few ways that the Chro mebooks will be used in my classroom.nannan

In [56]:

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

Students enter class eager to learn. Technology engages young minds to make learning exciting. "We need technology in every classroom and in every stude nt and teacher's hand, because it is the pen and paper of our time, and it i s the lens through which we experience much of our world." -D. Warlick The students I teach come from different socioeconomic and ethnic backgrounds. I teach second graders at a Title One school. I am fortunate to have a diver se group of second grade students. At our school, we have a high poverty lev el. All of our students receive free or reduced-price lunch. Several student s receive resource and ELL services. One thing all my students have in commo n is that they enjoy learning with technology. My students will use the Chro mebooks regularly. They will use them for math, language arts, science and s ocial studies. Students will work research topics with the Chromebooks and u se them to enhance their learning. With the Chromebooks, the students would be able to utilize more technology. These Chromebooks will help students lear The Chromebooks will be used by my second grade students. They will use them as they work in collaborative groups. Students will use them across su bject areas. For example with social studies, students will use them to res earch important historical figures like Dr. Martin Luther King Jr. and Abrah am Lincoln. Students will work in their groups to create digital presentati ons on their historical figure to share with others. In science, students w ill use the Chromebooks in collaborative groups to study weather patterns. In reading, students will use the Chromebooks to listen to leveled books and create digital retells. In math, students will use the Chromebooks to focus addition/subtraction practice. These are only a few ways that the Chromeboo ks will be used in my classroom.nannan

In [57]:

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

Students enter class eager to learn Technology engages young minds to make 1 earning exciting We need technology in every classroom and in every student and teacher s hand because it is the pen and paper of our time and it is the lens through which we experience much of our world D Warlick The students I teach come from different socioeconomic and ethnic backgrounds I teach secon d graders at a Title One school I am fortunate to have a diverse group of se cond grade students At our school we have a high poverty level All of our st udents receive free or reduced price lunch Several students receive resource and ELL services One thing all my students have in common is that they enjoy learning with technology My students will use the Chromebooks regularly They will use them for math language arts science and social studies Students wil 1 work research topics with the Chromebooks and use them to enhance their le arning With the Chromebooks the students would be able to utilize more techn ology These Chromebooks will help students learn The Chromebooks will be use d by my second grade students They will use them as they work in collaborati ve groups Students will use them across subject areas For example with socia 1 studies students will use them to research important historical figures li ke Dr Martin Luther King Jr and Abraham Lincoln Students will work in their groups to create digital presentations on their historical figure to share w ith others In science students will use the Chromebooks in collaborative gro ups to study weather patterns In reading students will use the Chromebooks t o listen to leveled books and create digital retells In math students will u se the Chromebooks to focus addition subtraction practice These are only a f ew ways that the Chromebooks will be used in my classroom nannan

In [58]:

In [59]:

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

100%| 100%| 10000/10000 [00:12<00:00, 785.86it/s]

In [60]:

```
# after preprocesing
preprocessed_essays[1000]
```

Out[60]:

'students enter class eager learn technology engages young minds make learni ng exciting we need technology every classroom every student teacher hand pe n paper time lens experience much world d warlick the students i teach come different socioeconomic ethnic backgrounds i teach second graders title one school i fortunate diverse group second grade students at school high povert y level all students receive free reduced price lunch several students recei ve resource ell services one thing students common enjoy learning technology my students use chromebooks regularly they use math language arts science so cial studies students work research topics chromebooks use enhance learning with chromebooks students would able utilize technology these chromebooks he lp students learn the chromebooks used second grade students they use work c ollaborative groups students use across subject areas for example social stu dies students use research important historical figures like dr martin luthe r king jr abraham lincoln students work groups create digital presentations historical figure share others in science students use chromebooks collabora tive groups study weather patterns in reading students use chromebooks liste n leveled books create digital retells in math students use chromebooks focu s addition subtraction practice these ways chromebooks used classroom nanna

1.5.2 Project title Text (Assignment Part)

In [61]:

```
# printing some random essays.
print(project_data['project_title'].values[0])
print("="*50)
print(project_data['project_title'].values[150])
print("="*50)
print(project_data['project_title'].values[1000])
```

In [62]:

```
# Combining all the above statemennts
from tqdm import tqdm
preprocessed_titles = []
# tqdm is for printing the status bar
for sentance in tqdm(project_data['project_title'].values):
    sent = decontracted(sentance)
    sent = sent.replace('\\r', ' ')
    sent = sent.replace('\\", ' ')
    sent = sent.replace('\\", ' ')
    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())
```

```
100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 100% | 10
```

In [63]:

```
# after preprocesing
print("{{}}".format(preprocessed_titles[150]))
```

ink paper results

1.6 Preparing data for models

In [64]:

```
project_data.columns
```

```
Out[64]:
```

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

1.6.1 Vectorizing Categorical data

1.6.1.1 Clean Categories

In [65]:

```
# 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, bina
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',
```

1.6.1.2 Sub-Clean-categories

Shape of matrix after one hot encoding (10000, 9)

'SpecialNeeds', 'Health_Sports', 'Math_Science', 'Literacy_Language']

In [66]:

```
# 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', '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 (10000, 30)
```

1.6.1.3 School State (Assignment Part)

In [67]:

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

state_dict = dict(my_counter)

sorted_state_dict = dict(sorted(state_dict.items(),key=lambda kv: kv[1]))
vectorizer = CountVectorizer(vocabulary=list(sorted_state_dict.keys()), lowercase=False, bi
vectorizer.fit(project_data['school_state'].values)
print(vectorizer.get_feature_names())

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

```
['VT', 'WY', 'ND', 'DE', 'NE', 'MT', 'NH', 'RI', 'SD', 'WV', 'AK', 'NM', 'H
I', 'ME', 'DC', 'KS', 'IA', 'ID', 'CO', 'MN', 'KY', 'AR', 'NV', 'OR', 'MS',
'MD', 'CT', 'TN', 'AL', 'UT', 'WI', 'VA', 'OH', 'NJ', 'WA', 'MA', 'AZ', 'M
O', 'IN', 'OK', 'LA', 'MI', 'PA', 'SC', 'GA', 'IL', 'NC', 'FL', 'NY', 'TX',
'CA']
Shape of matrix after one hot encoding (10000, 51)
```

1.6.1.4 Teacher Prefixes (Assignment Part)

In [68]:

```
#https://stackoverflow.com/questions/39303912/tfidfvectorizer-in-scikit-learn-valueerror-np
prefix = list(set(project_data['teacher_prefix'].values))
vectorizer = CountVectorizer(vocabulary=prefix,lowercase=False,binary=True)

vectorizer.fit(project_data['teacher_prefix'].values.astype('U'))
print(vectorizer.get_feature_names())

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

```
['Dr.', 'Teacher', 'Mr.', 'Mrs.', 'Ms.'] Shape of matrix after one hot encoding (10000, 5)
```

1.6.1.5 project_grade_category (Assignment Part)

In [69]:

```
prefix = list(set(project_data['project_grade_category'].values))
vectorizer = CountVectorizer(vocabulary=prefix,lowercase=False,binary=True)
vectorizer.fit(project_data['project_grade_category'].values.astype('U'))
print(vectorizer.get_feature_names())
grade_one_hot = vectorizer.transform(project_data['project_grade_category'].values.astype('print("Shape of matrix after one hot encoding ",grade_one_hot.shape)
```

```
['Grades 3-5', 'Grades 6-8', 'Grades 9-12', 'Grades PreK-2'] Shape of matrix after one hot encoding (10000, 4)
```

1.7 Vectorizing Text data

1.7.1 Bow:Essay

In [70]:

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

Shape of matrix after one hot encoding (10000, 6108)

1.7.2 Bow:Project_title (Assignment Part)

In [71]:

```
vectorizer = CountVectorizer(min_df=10)
title_bow = vectorizer.fit_transform(preprocessed_titles)
print("Shape of matrix after one hot encoding ",title_bow.shape)
```

Shape of matrix after one hot encoding (10000, 664)

1.8 TFIDF vectorizer

1.8.1 TFIDF on essays

In [72]:

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

Shape of matrix after one hot encoding (10000, 6108)

1.8.2 TFIDF on project title (Assignment Part)

In [73]:

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

Shape of matrix after one hot encoding (10000, 664)

1.9 Using Pretrained Models: Avg W2V

In [74]:

```
import pickle
import itertools
from gensim.models.word2vec import Text8Corpus
from glove import Corpus, Glove
```

In [75]:

```
# 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')
words = []
for i in preprocessed_essays:
    words.extend(i.split(' '))
for i in preprocessed_titles:
    words.extend(i.split(' '))
print("all the words in the coupus", len(words))
words = set(words)
print("the unique words in the coupus", len(words))
inter_words = set(model.keys()).intersection(words)
print("The number of words that are present in both glove vectors and our coupus", \
      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-pickl
import pickle
with open('glove_vectors', 'wb') as f:
    pickle.dump(words courpus, f)
Loading Glove Model
1917495it [09:54, 3225.40it/s]
Done. 1917495 words loaded!
all the words in the coupus 1526924
the unique words in the coupus 23068
The number of words that are present in both glove vectors and our coupus 22
089 (95.756 %)
word 2 vec length 22089
```

In [76]:

```
# stronging variables into pickle files python: http://www.jessicayung.com/how-to-use-pickl
with open('glove_vectors', 'rb') as f:
    model = pickle.load(f)
    glove_words = set(model.keys())
```

1.9.2 AVG W2V on project_essay

In [77]:

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

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

```
100%| 100%| 10000/10000 [00:06<00:00, 1555.67it/s]
10000
300
```

1.9.2 AVG W2V on project_title (Assignment Part)

In [78]:

```
avg_w2v_vectors_titles = []; # 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_titles.append(vector)

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

```
100%| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 10000| 100
```

1.10 Using Pretrained Models: TFIDF weighted W2V

1.10.1 TFIDF Weighted W2V on project_essay

```
In [79]:
```

```
# 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())
print(len(tfidf_words))
```

22348

In [80]:

```
tfidf_w2v_vectors = [];
for sentence in tqdm(preprocessed_essays):
    vector = np.zeros(300)
    tf idf weight=0;
    for word in sentence.split():
        try:
            if(word in glove_words) and (word in tfidf_words):
                vec = model[word]
                tf_idf = dictionary[word]*(sentence.count(word)/len(sentence.split()))
                vector += (vec * tf idf)
                tf_idf_weight += tf_idf
        except:
            pass
    if tf_idf_weight !=0:
        vector /=tf_idf_weight
    tfidf_w2v_vectors.append(vector)
print(len(tfidf_w2v_vectors))
print(len(tfidf_w2v_vectors[0]))
```

100%| 10000 | 10000 | 10000 | 10000 | 224.88it/s | 10000 | 300

1.10.1 TFIDF Weighted W2V on project_title

In [81]:

```
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())
print(len(tfidf_words))  #To get the length of tfidf vector
```

5366

In [82]:

```
tfidf w2v vectors titles = [];
for sentence in tqdm(preprocessed_titles):
    vector = np.zeros(300)
    tf idf weight=0;
    for word in sentence.split():
        try:
            if(word in glove_words) and (word in tfidf_words):
                vec = model[word]
                tf_idf = dictionary[word]*(sentence.count(word)/len(sentence.split()))
                vector += (vec * tf idf)
                tf_idf_weight += tf_idf
        except:
            pass
    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]))
```

```
100%| 100%| 10000/10000 [00:00<00:00, 14775.67it/s]
10000
300
```

1.11 Vectorizing Numerical features

Standardizing price

```
In [83]:
```

```
# check this one: https://www.youtube.com/watch?v=0HOqOcln3Z4&t=530s
# standardization sklearn: https://scikit-learn.org/stable/modules/generated/sklearn.prepro
from sklearn.preprocessing import StandardScaler

# price_standardized = standardScalar.fit(project_data['price'].values)
# this will rise the error
# ValueError: Expected 2D array, got 1D array instead: array=[725.05 213.03 329. ... 399.
# 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 standar
print(f"Mean : {price_scalar.mean_[0]}, Standard deviation : {np.sqrt(price_scalar.var_[0])
# Now standardize the data with above maen and variance.
price_standardized = price_scalar.transform(project_data['price'].values.reshape(-1, 1))
```

Mean : 323.177644, Standard deviation : 352.44291350615816

Standardizing Number of previously posted projects by teachers

```
In [84]:
```

C:\Users\VANSHIKA\Anaconda3\lib\site-packages\sklearn\utils\validation.py:59

5: DataConversionWarning:

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

```
Mean: 9.2928, Standard deviation: 24.92901659031098
```

C:\Users\VANSHIKA\Anaconda3\lib\site-packages\sklearn\utils\validation.py:59

5: DataConversionWarning:

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

In [85]:

```
price_standardized
```

```
Out[85]:
```

1.12 Merging all the above features

```
In [86]:
```

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

(10000, 9)
(10000, 30)
(10000, 6108)
(10000, 1)
```

Assignment 2: Apply TSNE

2.1 t-SNE BoW encoding of project_title

```
In [87]:
```

```
Out[87]:
```

(10000, 765)

In [88]:

```
X_dense = X.todense()
type(X_dense)
```

Out[88]:

numpy.matrix

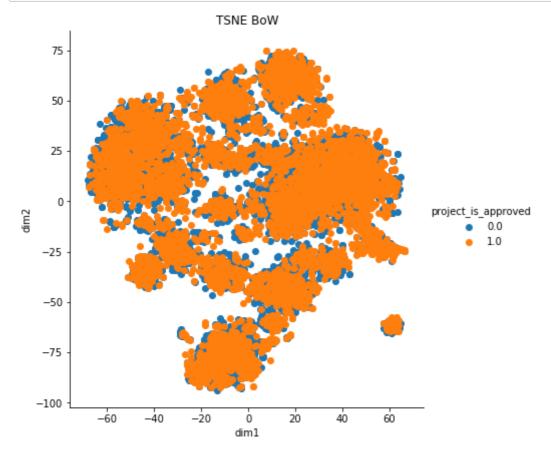
In [89]:

```
from sklearn.manifold import TSNE
model = TSNE(n_components=2, random_state=0, perplexity = 50, n_iter = 5000)

tsne_data = model.fit_transform(X_dense)
y = project_data['project_is_approved']

tsne_data = np.vstack((tsne_data.T, y)).T
tsne_df = pd.DataFrame(data=tsne_data, columns=("dim1", "dim2", "project_is_approved"))

# Ploting the result of tsne
sns.FacetGrid(tsne_df, hue="project_is_approved", height=6).map(plt.scatter, 'dim1', 'dim2'
plt.title("TSNE BoW")
plt.show()
```



Note:

- 1. Not much improvement on plot has been obeserved on changing preplexity and n_iter.
- 2. The approved and rejected projects are highly overlapped.

2.1 t-SNE for TFIDF

In [90]:

Out[90]:

numpy.matrix

In [91]:

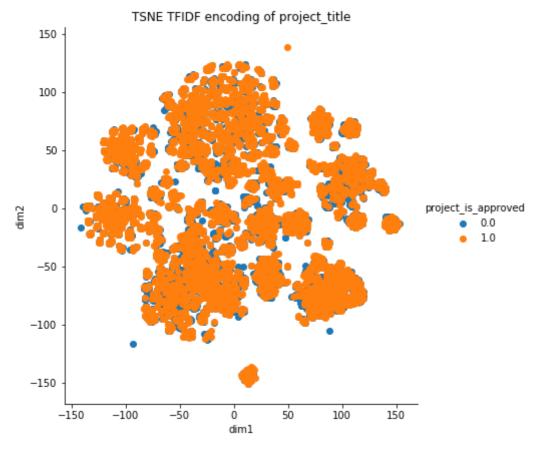
```
from sklearn.manifold import TSNE
model = TSNE(n_components=2, random_state=0, perplexity = 30, n_iter = 5000)

tsne_data = model.fit_transform(X_dense)

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

tsne_df = pd.DataFrame(data=tsne_data, columns=("dim1", "dim2", "project_is_approved"))

# Ploting the result of tsne
sns.FacetGrid(tsne_df, hue="project_is_approved", height=6).map(plt.scatter, 'dim1', 'dim2'
plt.title("TSNE TFIDF encoding of project_title")
plt.show()
```



Note:

- 1. Not much improvement on plot has been obeserved on changing preplexity and n iter.
- 2. The approved and rejected projects are highly overlapped.

2.3 TSNE with AVG W2V encoding of project_title

In [94]:

Out[94]:

numpy.matrix

In [95]:

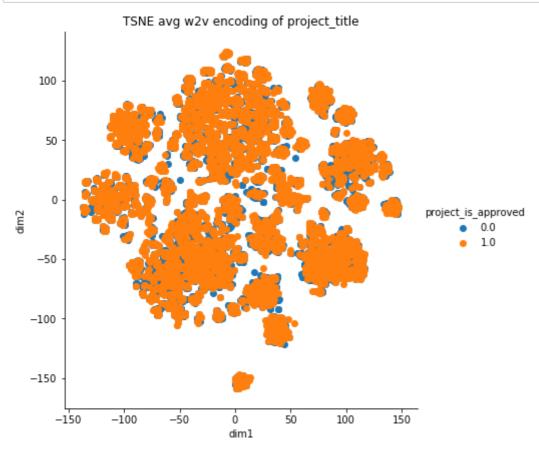
```
from sklearn.manifold import TSNE
model = TSNE(n_components=2, random_state=0, perplexity = 30, n_iter = 5000)

tsne_data = model.fit_transform(X_dense)

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

tsne_df = pd.DataFrame(data=tsne_data, columns=("dim1", "dim2", "project_is_approved"))

# Ploting the result of tsne
sns.FacetGrid(tsne_df, hue="project_is_approved", height=6).map(plt.scatter, 'dim1', 'dim2'
plt.title("TSNE avg w2v encoding of project_title")
plt.show()
```



Note:

- 1. Not much improvement on plot has been obeserved on changing preplexity and n_iter.
- 2. The approved and rejected projects are highly overlapped.

2.4 TSNE with TFIDF Weighted W2V encoding of project_title

In [96]:

Out[96]:

numpy.matrix

In [97]:

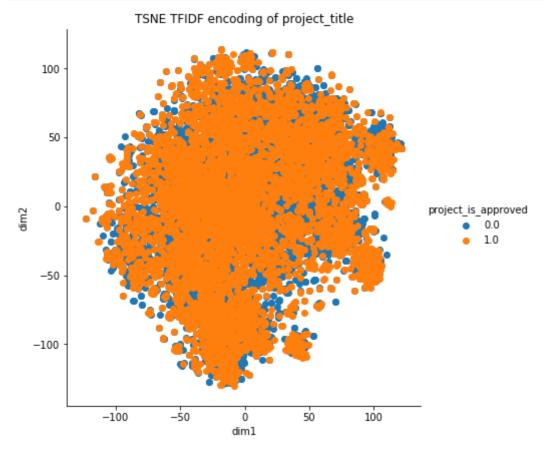
```
from sklearn.manifold import TSNE
model = TSNE(n_components=2, random_state=0, perplexity = 30, n_iter = 5000)

tsne_data = model.fit_transform(X_dense)

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

tsne_df = pd.DataFrame(data=tsne_data, columns=("dim1", "dim2", "project_is_approved"))

# Ploting the result of tsne
sns.FacetGrid(tsne_df, hue="project_is_approved", height=6).map(plt.scatter, 'dim1', 'dim2'
plt.title("TSNE TFIDF encoding of project_title")
plt.show()
```



Note:

- 1. Not much improvement on plot has been obeserved on changing preplexity and n iter.
- 2. The approved and rejected projects are highly overlapped.

Conclusion:

- 1. Vermont, District of Columbia, Texas, Montana and Louisiana are the states with least percentage of project approvals ranging 80%-83%.
- 2. New Hampshire, Ohio, Washington, North Dakota and Delaware are the stated with high percentage of project approvals ranging 87%-90% approx.
- 3. More number of projects in a particular category does not ensure higher approval rate.
- 4. The project based on "Literacy" and education" has higher approval rate.
- 5. Projects by Mrs., MS., and Mr. has relatively higher approval rate.
- 6. Price is not a good feature to see check whether the project will be approved or not.
- 7. Project summary with numerical values mentioned has higher approval rate.
- 8. The number of projects that contains digits in its resource summary has reatively higher approval rate than those doesn't have digits in its resource summary.
- 9. Projects with higher number words in its essay are more likely to get approved.
- 10. TSNE didn't work well to separate the two categories.