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
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
```

In [2]:

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

```
In [3]:
```

```
print("\n\nNumber of data points in training data", training_data.shape)
print("\n\nThe attributes of data :", training_data.columns.values)
print('\n')
training_data.head(2)
```

```
Number of data points in training 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_ap
  proved']
```

Out[3]:

	Unnamed: 0	id	teacher_id	teacher_prefix	scl
0	160221	p253737	c90749f5d961ff158d4b4d1e7dc665fc	Mrs.	
1	140945	p258326	897464ce9ddc600bced1151f324dd63a	Mr.	

Preprocessing teacher_prefix

In [4]:

```
training_data['teacher_prefix'] = training_data['teacher_prefix'].replace(np.
training_data['teacher_prefix'] = training_data['teacher_prefix'].replace('Dr
training_data['teacher_prefix'] = training_data['teacher_prefix'].replace('Teacher_prefix'].replace('Teacher_prefix').
training_data['teacher_prefix'] = training_data['teacher_prefix'].replace('Mr
training_data['teacher_prefix'] = training_data['teacher_prefix'].replace('Ms
training_data['teacher_prefix'] = training_data['teacher_prefix'].replace('Mr
print(training_data['teacher_prefix'].head(5))
training data.head(2)
```

```
0
     MRS
1
      MR
2
      MS
3
     MRS
4
     MRS
Name: teacher prefix, dtype: object
```

Out[4]:

	Unnamed: 0	id	teacher_id	teacher_prefix	scl
0	160221	p253737	c90749f5d961ff158d4b4d1e7dc665fc	MRS	
1	140945	p258326	897464ce9ddc600bced1151f324dd63a	MR	

Summary

- Removing np.nan values from teacher prefix column and converting to MRS
- Removing Full stops from teacher prefix column and converting to Upper case

Preprocessing project_grade_category

In [5]:

```
training_data['project_grade_category'] = training_data['project_grade_category'] = training_data['project_grade_category'] = training_data['project_grade_category'] = training_data['project_grade_category'] = training_data['project_grade_category'] = training_data['project_grade_category'].head(5))
training_data.head(2)
```

```
0   Grades_PreK_2
1    Grades_6_8
2    Grades_6_8
3   Grades_PreK_2
4   Grades_PreK_2
Name: project_grade_category, dtype: object
```

Out[5]:

	Unnamed: 0	id	teacher_id	teacher_prefix	scl
0	160221	p253737	c90749f5d961ff158d4b4d1e7dc665fc	MRS	
1	140945	p258326	897464ce9ddc600bced1151f324dd63a	MR	

Summary

Replacing ' ' and '-' with '_' for all the values of project_grade_category column

In [6]:

```
print("\n\nNumber of data points in resource data", resource_data.shape)
print("\n\nThe attributes of data :",resource_data.columns.values)
resource_data.head(2)
```

Number of data points in resource data (1541272, 4)

The attributes of data : ['id' 'description' 'quantity' 'pric e']

Out[6]:

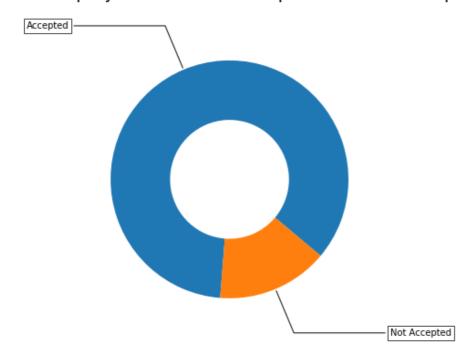
	id	description	quantity	price
0	p233245	LC652 - Lakeshore Double-Space Mobile Drying Rack	1	149.00
1	p069063	Bouncy Bands for Desks (Blue support pipes)	3	14.95

Data Analysis

(15.141695957820739 %)

```
# https://matplotlib.org/gallery/pie and polar charts/pie and donut labels.ht
y_value_counts = training_data['project_is_approved'].value_counts()
print("Number of projects that are approved for funding ", y_value_counts[1].
print("Number of projects that are not approved for funding ", y_value_counts
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("Number of projects that are Accepted and not accepted", fontsize
plt.show()
Number of projects that are approved for funding 92706, (8
4.85830404217927 %)
Number of projects that are not approved for funding 16542,
```

Number of projects that are Accepted and not accepted

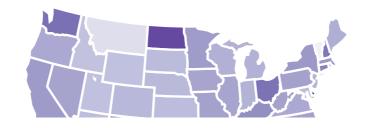


Summary

- 92706 projects were approved for funding out of 109248 applications i.e. 84.85 %
- 16542 projects were not approved for funding out of 109248 applications i.e. 15.14%
- Looks like an imbalanced dataset

```
# Pandas dataframe groupby count, mean: https://stackoverflow.com/a/19385591/
temp = pd.DataFrame(training_data.groupby("school_state")["project_is_approve
# if you have data which contain only 0 and 1, then the mean = percentage (th
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
            [0.6, 'rgb(158,154,200)'], [0.8, 'rgb(117,107,177)'], [1.0, 'rgb(84
data = [ dict(
       type='choropleth',
        colorscale = scl,
       autocolorscale = False,
        locations = temp['state_code'],
        z = temp['num proposals'].astype(float),
        locationmode = 'USA-states',
       text = temp['state_code'],
       marker = dict(line = dict (color = 'rgb(255, 255, 255)', width = 2)),
        colorbar = dict(title = "% of pro")
   ) ]
layout = dict(
       title = 'Project Proposals % of Acceptance Rate by US States',
        geo = dict(
            scope='usa',
            projection=dict( type='albers usa' ),
            showlakes = True,
            lakecolor = 'rgb(255, 255, 255)',
        ),
fig = go.Figure(data=data, layout=layout)
offline.iplot(fig, filename='us-map-heat-map')
```

Project Proposals % of Acceptance Rate by US Sta





Summary

- The plot shows all the 51 states of United States of America with rate of acceptance
- The shades of purple represents the percentage of acceptance
- The darker the shade the higher the acceptance percentage
- Delaware (DE) is the state with Most Acceptance percentage i.e. 89.79 %
- North Dakota (ND) is the state with Second most Acceptance percentage i.e. 88.81 %
- Washington (WA) is the state with Third most Acceptance percentage i.e. 87.61 %

In [9]:

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

Summary

- Vermont state has the lowest percentage of acceptance i.e. 80.00 %
- Washington, D.C. has the second lowest percentage of acceptance i.e. 80.23 %
- Texas state has the third lowest percentage of acceptance i.e. 81.31 %
- Every state has greater than 80 % acceptance rate

In [10]:

```
#stacked bar plots matplotlib: https://matplotlib.org/gallery/lines_bars_and_
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/
    temp = pd.DataFrame(training_data.groupby(col1)[col2].agg(lambda x: x.eq

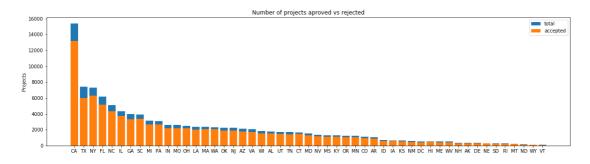
# Pandas dataframe grouby count: https://stackoverflow.com/a/19385591/40&
    temp['total'] = pd.DataFrame(training_data.groupby(col1)[col2].agg({'totatemp['Avg'] = pd.DataFrame(training_data.groupby(col1)[col2].agg({'Avg': temp.sort_values(by=['total'],inplace=True, ascending=False)

if top:
    temp = temp[0:top]

stack_plot(temp, xtick=col1, col2=col2, col3='total')
    print(temp.head(5))
    print('\n\n')
    print(temp.tail(5))
    print('\n')
```

Univariate Analysis: school_state

univariate_barplots(training_data, 'school_state', 'project_is_approved', Fal



	school_state	<pre>project_is_approved</pre>	total	Avg
4	CA	13205	15388	0.858136
43	TX	6014	7396	0.813142
34	NY	6291	7318	0.859661
9	FL	5144	6185	0.831690
27	NC	4353	5091	0.855038

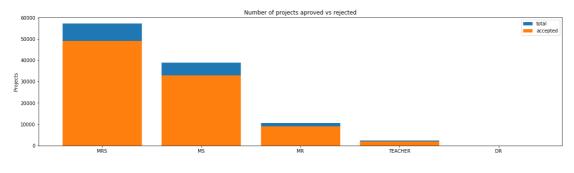
	school_state	<pre>project_is_approved</pre>	total	Avg
39	RI	243	285	0.852632
26	MT	200	245	0.816327
28	ND	127	143	0.888112
50	WY	82	98	0.836735
46	VT	64	80	0.800000

Summary

- California state has the most number of project submissions i.e. 15388
- Texas state has the second most number of project submissions i.e. 7396
- Wyoming state has the second least number of project submissions i.e. 98
- Vermont state has the least number of project submissions i.e. 80

Univariate Analysis: teacher_prefix

univariate_barplots(training_data, 'teacher_prefix', 'project_is_approved' ,



	teacher_prefix	<pre>project_is_approved</pre>	total	Avg
2	MRS	49000	57272	0.855566
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	<pre>project_is_approved</pre>	total	Avg
2	MRS	49000	57272	0.855566
3	MS	32860	38955	0.843537
1	MR	8960	10648	0.841473
4	TEACHER	1877	2360	0.795339
0	DR	9	13	0.692308

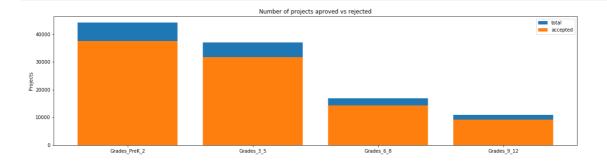
Summary

- MRS teacher_prefix has the most number of project submissions i.e. 57272 with 85.55 % acceptance rate
- MS teacher_prefix has the second most number of project submissions i.e. 38955 with 84.35 % acceptance rate
- Teacher teacher_prefix has the second least number of project submissions i.e. 2360 with 79.53 % acceptance rate
- DR teacher_prefix has the least number of project submissions i.e. 13 with 69.23 % acceptance rate

Univariate Analysis: project_grade_category

In [14]:

univariate_barplots(training_data, 'project_grade_category', 'project_is_appr



	<pre>project_grade_category</pre>	<pre>project_is_approved</pre>	total	Avg
3	Grades_PreK_2	37536	44225	0.848751
0	Grades_3_5	31729	37137	0.854377
1	Grades_6_8	14258	16923	0.842522
2	Grades 9 12	9183	10963	0.837636

	<pre>project_grade_category</pre>	<pre>project_is_approved</pre>	total	Avg
3	Grades_PreK_2	37536	44225	0.848751
0	Grades_3_5	31729	37137	0.854377
1	Grades_6_8	14258	16923	0.842522
2	Grades 9 12	9183	10963	0.837636

Summary

- Most number of projects are for Grades pre K 2 i.e. 44225 with 84.87 % acceptance rate
- Second Most number of projects are for Grades 3-5 i.e. 37137 with 85.43 % acceptance rate
- Third Most number of projects are for Grades 6 8 i.e. 16923 with 84.25 % acceptance rate
- Least number of projects are for Grades 9 12 i.e. 10963 with 83.76 % acceptance rate

Univariate Analysis: project_subject_categories

In [15]:

```
catogories = list(training_data['project_subject_categories'].values)
# remove special characters from list of strings python: https://stackoverflo
# https://www.geeksforgeeks.org/removing-stop-words-nltk-python/
# https://stackoverflow.com/questions/23669024/how-to-strip-a-specific-word-1
# https://stackoverflow.com/questions/8270092/remove-all-whitespace-in-a-stri
cat list = []
for i in catogories:
   temp = ""
   # consider we have text like this "Math & Science, Warmth, Care & Hunger'
   for j in i.split(','): # it will split it in three parts ["Math & Science
        if 'The' in j.split(): # this will split each of the catogory based of
            j=j.replace('The','') # if we have the words "The" we are going t
       j = j.replace(' ','') # we are placeing all the ' '(space) with ''(en
        temp+=j.strip()+" " #" abc ".strip() will return "abc", remove the tr
        temp = temp.replace('&','_') # we are replacing the & value into
   cat list.append(temp.strip())
```

In [16]:

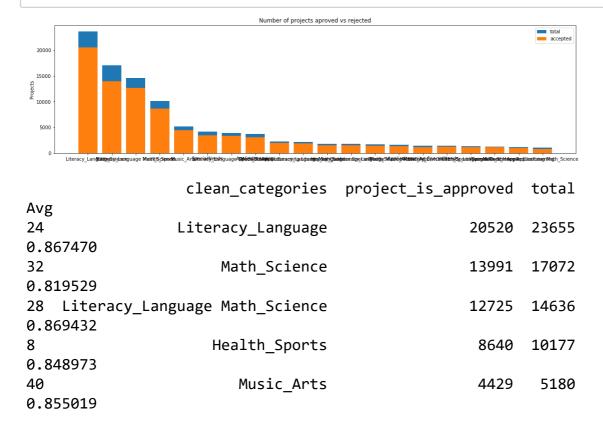
```
training_data['clean_categories'] = cat_list
training_data.drop(['project_subject_categories'], axis=1, inplace=True)
training_data.head(2)
```

Out[16]:

	Unnamed: 0	id	teacher_id	teacher_prefix	scl
0	160221	p253737	c90749f5d961ff158d4b4d1e7dc665fc	MRS	
1	140945	p258326	897464ce9ddc600bced1151f324dd63a	MR	

In [17]:

univariate_barplots(training_data, 'clean_categories', 'project_is_approved'



	clean_categories	<pre>project_is_approved</pre>	tot
al	Avg		
19	<pre>History_Civics Literacy_Language</pre>	1271	14
21	0.894441		
14	Health_Sports SpecialNeeds	1215	13
91	0.873472		
50	Warmth Care_Hunger	1212	13
09	0.925898		
33	Math_Science AppliedLearning	1019	12
20	0.835246		
4	AppliedLearning Math_Science	855	10
52	0.812738		

Summary

- Most number of projects are for Category Literacy & Language i.e. 23655 with 86.74 % acceptance rate
- Second Most number of projects are for Category Math & Science i.e. 17072 with 81.95
 % acceptance rate

- Third Most number of projects are for Category Literacy, Language, Math & Science i.e. 14636 with 0.86.94 % acceptance rate
- Least number of projects are for Category AppliedLearning, Math & Science i.e. 1052 with 81.27 % acceptance rate

In [18]:

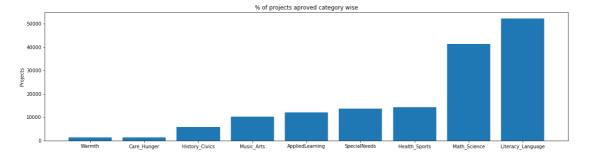
```
# count of all the words in corpus python: https://stackoverflow.com/a/228985
from collections import Counter
my_counter = Counter()
for word in training_data['clean_categories'].values:
    my_counter.update(word.split())
```

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



In [20]:

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

Warmth 1388 Care Hunger 1388 History Civics 5914 Music Arts 10293 AppliedLearning : 12135 SpecialNeeds 13642 Health Sports 14223 Math_Science 41421 Literacy Language 52239

1.2.5 Univariate Analysis: project_subject_subcategories

In [21]:

```
sub_catogories = list(training_data['project_subject_subcategories'].values)
# remove special characters from list of strings python: https://stackoverflo
# https://www.geeksforgeeks.org/removing-stop-words-nltk-python/
# https://stackoverflow.com/questions/23669024/how-to-strip-a-specific-word-j
# https://stackoverflow.com/questions/8270092/remove-all-whitespace-in-a-stri
sub cat list = []
for i in sub catogories:
   temp = ""
   # consider we have text like this "Math & Science, Warmth, Care & Hunger'
   for j in i.split(','): # it will split it in three parts ["Math & Science
        if 'The' in j.split(): # this will split each of the catogory based of
            j=j.replace('The','') # if we have the words "The" we are going t
        j = j.replace(' ','') # we are placeing all the ' '(space) with ''(en
        temp +=j.strip()+" "#" abc ".strip() will return "abc", remove the tr
        temp = temp.replace('&',' ')
   sub cat list.append(temp.strip())
```

In [22]:

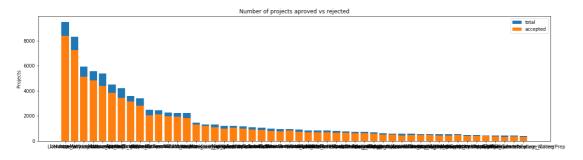
```
training_data['clean_subcategories'] = sub_cat_list
training_data.drop(['project_subject_subcategories'], axis=1, inplace=True)
training_data.head(2)
```

Out[22]:

	Unnamed: 0	id	teacher_id	teacher_prefix	scl
0	160221	p253737	c90749f5d961ff158d4b4d1e7dc665fc	MRS	
1	140945	p258326	897464ce9ddc600bced1151f324dd63a	MR	

In [23]:

univariate_barplots(training_data, 'clean_subcategories', 'project_is_approve



		clean_subcategories	<pre>project_is_approved</pre>	tota
1	Avg			
317	7	Literacy	8371	948
6	0.882458			
319)	Literacy Mathematics	7260	832
5	0.872072			
331	L Literatu	re_Writing Mathematics	5140	592
3	0.867803			
318	B Litera	acy Literature_Writing	4823	557
1	0.865733			
342	2	Mathematics	4385	537
9	0.815207			

		clean_subcategories	<pre>project_is_approved</pre>
tota	l Avg		
196	Environmer	talScience 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

- Most number of projects are for Sub-Category Literacy i.e. 9486 with 88.2458 % acceptance rate
- Secomd Most number of projects are for Sub-Category Literacy & Mathematics i.e. 8325 with 87.2072 % acceptance rate

 Least number of projects are for Sub-Category AppliedSciences & College_CareerPrep i.e. 405 with 81.4815 % acceptance rate

In [24]:

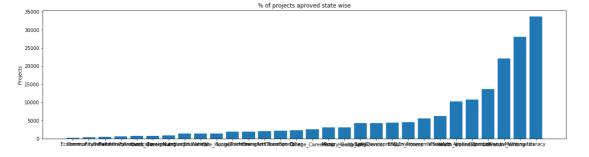
```
# count of all the words in corpus python: https://stackoverflow.com/a/228985
from collections import Counter
my_counter = Counter()
for word in training_data['clean_subcategories'].values:
    my_counter.update(word.split())
```

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



In [26]:

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

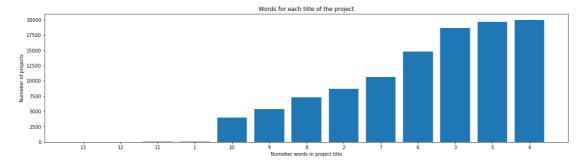
Economics 269 CommunityService 441 FinancialLiteracy 568 ParentInvolvement 677 Extracurricular 810 Civics Government 815 ForeignLanguages 890 NutritionEducation 1355 Warmth 1388 Care_Hunger 1388 SocialSciences : 1920 PerformingArts 1961 CharacterEducation 2065 TeamSports 2192 **Other** 2372 College CareerPrep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

In [27]:

```
#How to calculate number of words in a string in DataFrame: https://stackover
word_count = training_data['project_title'].str.split().apply(len).value_cour
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))
pl = 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()
```



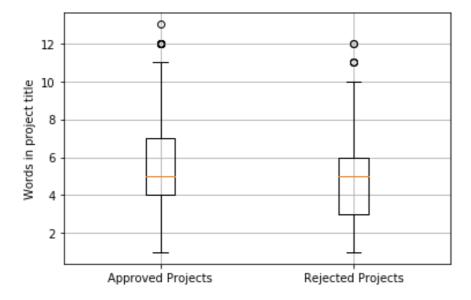
In [28]:

```
approved_title_word_count = training_data[training_data['project_is_approved
approved_title_word_count = approved_title_word_count.values

rejected_title_word_count = training_data[training_data['project_is_approved
rejected_title_word_count = rejected_title_word_count.values
```

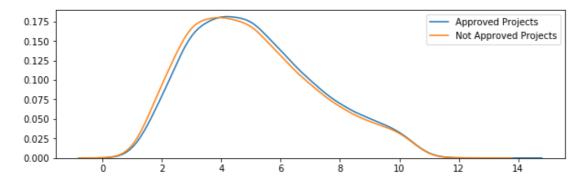
In [29]:

```
# 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()
```



In [30]:

```
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

- Most number of projects are accepted with 4 words in the title
- Second Most number of projects are accepted with 5 words in the title
- Third Most number of projects are accepted with 3 words in the title
- Least number of projects are accepted with 13 words in the title

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

In [31]:

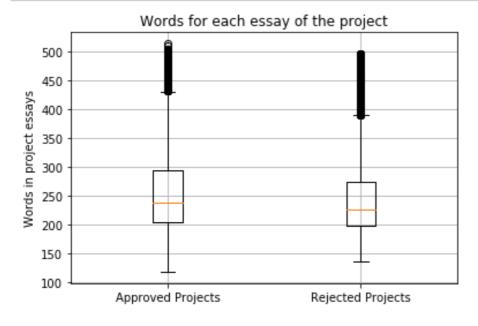
In [32]:

```
approved_word_count = training_data[training_data['project_is_approved']==1]|
approved_word_count = approved_word_count.values

rejected_word_count = training_data[training_data['project_is_approved']==0]|
rejected_word_count = rejected_word_count.values
```

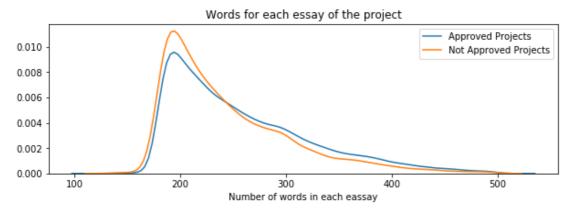
In [33]:

```
# 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 [34]:

```
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()
```



Summary

- More than 50 % of accepted projects have less than 250 words in their essays
- More than 50 % of rejected projects have less than 250 words in their essays
- More than 75 % of accepted projects have less than 300 words in their essays
- More than 75 % of rejected projects have less than 300 words in their essays

1.2.8 Univariate Analysis: Cost per project

In [35]:

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

Out[35]:

	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 [36]:

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

Out[36]:

	id	price	quantity
0	p000001	459.56	7
1	p000002	515.89	21

In [37]:

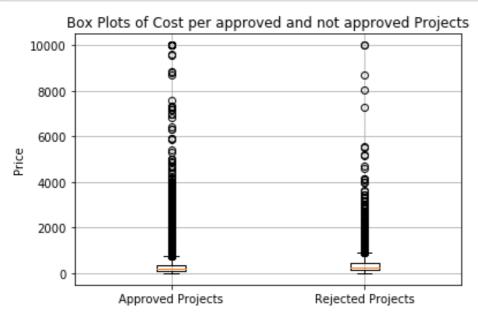
```
# join two dataframes in python:
training_data = pd.merge(training_data, price_data, on='id', how='left')
```

In [38]:

```
approved_price = training_data[training_data['project_is_approved']==1]['pric
rejected_price = training_data[training_data['project_is_approved']==0]['pric
```

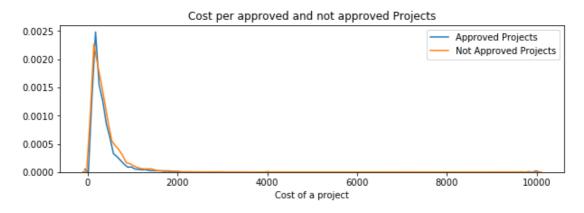
In [39]:

```
# 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 [40]:

```
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()
```



In [41]:

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

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

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.peprint(x))
```

+ Percentile	+ Approved Projects	Not Approved Projects
+	+	
0	0.66	1.97
5	13.59	41.9
10	33.88	73.67
15	58.0	99.109
20	77.38	118.56
25	99.95	140.892
30	116.68	162.23
35	137.232	184.014
40	157.0	208.632
45	178.265	235.106
50	198.99	263.145
55	223.99	292.61
60	255.63	325.144
65	285.412	362.39
70	321.225	399.99
75	366.075	449.945
80	411.67	519.282
85	479.0	618.276
90	593.11	739.356
95	801.598	992.486
100	9999.0	9999.0
+	+	

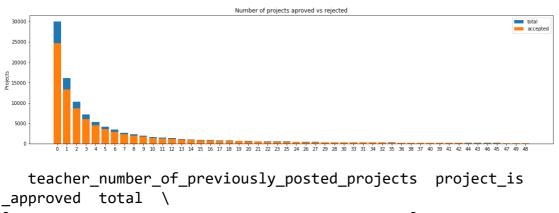
Summary

- All the projects prices are between 0.66 to 9999.0
- 50 % of accepted project prices are under 198.99
- 50 % of rejected project prices are under 263.145
- Accepted project prices are less when compared to rejected projects prices

Univariate Analysis: teacher_number_of_previously_posted_project:

In [42]:

univariate_barplots(training_data, 'teacher_number_of_previously_posted_proje



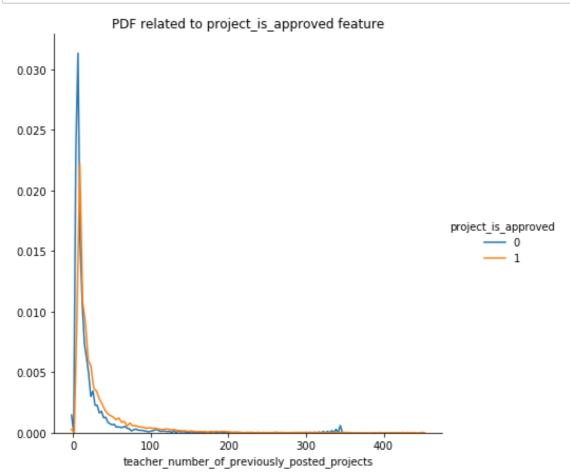
```
0
                                                     0
24652
        30014
                                                     1
13329
       16058
                                                     2
2
8705
      10350
3
                                                     3
5997
        7110
                                                     4
4
4452
        5266
```

Avg 0 0.821350 1 0.830054 2 0.841063 3 0.843460 4 0.845423

teacher_number_of_previously_posted_projects project_i s_approved total \

Avg 46 0.908537 45 0.921569 47 0.895833 49 0.895105

In [43]:



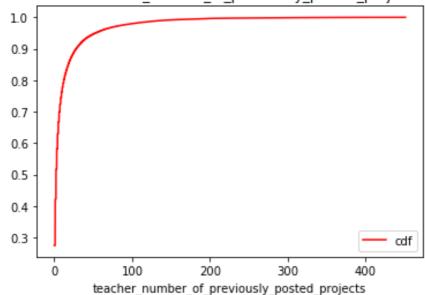
In [44]:

```
number_of_Points, bin_edges = np.histogram(training_data['teacher_number_of_points = 109248, density='True')

pdf = number_of_Points/sum(number_of_Points)
cdf = np.cumsum(pdf)
plt.xlabel('teacher_number_of_previously_posted_projects')
plt.plot(bin_edges[1:], cdf, 'r-', label = 'cdf')

plt.title('CDF related to teacher_number_of_previously_posted_projects featur
plt.legend()
plt.show()
training_data['teacher_number_of_previously_posted_projects'].max()
```

CDF related to teacher_number_of_previously_posted_projects feature



Out[44]:

451

Summary

- 27.46 % of teachers have posted the projects for the first time
- The first timers have posted 30014 projects with 82.13 % acceptance rate
- The maximum number of projects posted by any teacher is 451

project_resource_summary

In [45]:

```
training_data.project_resource_summary.head(20).get
```

Out[45]:

```
My students need opportunit
<bound method NDFrame.get of 0</pre>
ies to practice beg...
      My students need a projector to help with view...
      My students need shine guards, athletic socks,...
2
      My students need to engage in Reading and Math...
3
4
      My students need hands on practice in mathemat...
5
      My students need movement to be successful. Be...
      My students need some dependable laptops for d...
6
7
      My students need ipads to help them access a w...
      My students need three devices and three manag...
8
      My students need great books to use during Ind...
9
      My students need books by their favorite autho...
10
11
      My students need paper, three chromebooks, and...
      My students need 3D and 4D life science activi...
12
      My students need access to technology that wil...
13
      My students need 5 tablets for our classroom t...
14
      My students need activities to play during rec...
15
      My students need 2 LeapPad that will engage th...
16
      My students need Chromebooks to publish writte...
17
18
      My students need privacy partitions to use whi...
19
      My students need 7 Hokki stools to encourage a...
Name: project_resource_summary, dtype: object>
```

In [46]:

```
summary_list = list(training_data['project_resource_summary'].values)

project_resource_summary_list = []
for i in summary_list:
   if re.search('\d', i):
        project_resource_summary_list.append(1)
   else:
        project_resource_summary_list.append(0)
```

In [47]:

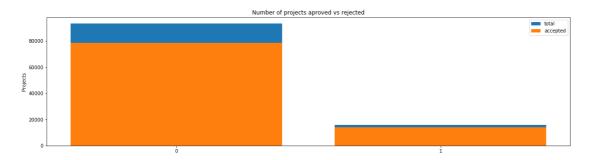
training_data['project_resource_summary_with_digits'] = project_resource_summary_with_digits'] = project_resource_summary_with_digits'

Out[47]:

	Unnamed: 0	id	teacher_id	teacher_prefix	school_state
0	160221	p253737	c90749f5d961ff158d4b4d1e7dc665fc	MRS	IN
1	140945	p258326	897464ce9ddc600bced1151f324dd63a	MR	FL

In [48]:

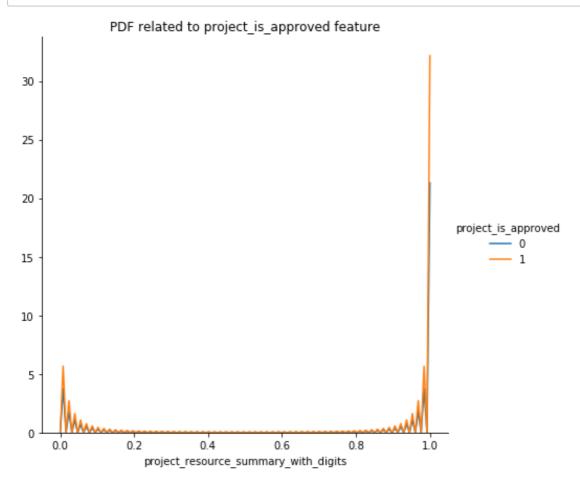
univariate_barplots(training_data, 'project_resource_summary_with_digits', 'project_resource_summary_summary_with_digits', 'project_resource_summary_with_digits', 'project_resource_summary_with_digits', 'project_resource_summary_summary_with_digits', 'project_resource_summary_with_digits', 'project_resource_summary_with_digits', 'project_resource_summary_summary_with_digits', 'project_resource_summary_with_digits', 'project_resource_summary_with_digits', 'project_resource_summary_s



project_resource_summary_with_digits project_is_approved
total Avg
0 0 78616
93492 0.840885
1 1 14090
15756 0.894263

<pre>project_is_approved</pre>	<pre>rce_summary_with_digits</pre>	ject_resou	pro
		Avg	total
78616	0		0
		0.840885	93492
14090	1		1
		0.894263	15756

In [49]:



In [50]:

training_data.head(2)

Out[50]:

	Unnamed: 0	id	teacher_id	teacher_prefix	scl
0	160221	p253737	c90749f5d961ff158d4b4d1e7dc665fc	MRS	
1	140945	p258326	897464ce9ddc600bced1151f324dd63a	MR	

2 rows × 21 columns

Summary

- 93492 projects were submitted without any digits in their essays with 84.0885 % acceptance ratio
- 15756 projects were submitted without any digits in their essays with 89.4263 % acceptance ratio
- Projects with digits in their essays have 5.33 % more acceptance ratio than projects without digits

```
# printing some random essays.
print(training_data['essay'].values[0])
print("\n\n")
print(training_data['essay'].values[150])
print("\n\n")
print(training_data['essay'].values[1000])
print("\n\n")
print(training_data['essay'].values[20000])
print("\n\n")
print(training_data['essay'].values[99999])
print(training_data['essay'].values[99999])
```

My students are English learners that are working on English a s their second or third languages. We are a melting pot of ref ugees, immigrants, and native-born Americans bringing the gift of language to our school. \r\n\r\n We have over 24 languages represented in our English Learner program with students at ev ery level of mastery. We also have over 40 countries represen ted with the families within our school. Each student brings a wealth of knowledge and experiences to us that open our eyes to new cultures, beliefs, and respect.\"The limits of your lan guage are the limits of your world.\"-Ludwig Wittgenstein Our English learner's have a strong support system at home that be gs for more resources. Many times our parents are learning to read and speak English along side of their children. Sometime s this creates barriers for parents to be able to help their c hild learn phonetics, letter recognition, and other reading sk ills.\r\n\r\nBy providing these dvd's and players, students ar e able to continue their mastery of the English language even if no one at home is able to assist. All families with studen ts within the Level 1 proficiency status, will be a offered to be a part of this program. These educational videos will be s pecially chosen by the English Learner Teacher and will be sen t home regularly to watch. The videos are to help the child d evelop early reading skills.\r\n\r\nParents that do not have a ccess to a dvd player will have the opportunity to check out a dvd player to use for the year. The plan is to use these vide os and educational dvd's for the years to come for other EL st udents.\r\nnannan

The 51 fifth grade students that will cycle through my classro om this year all love learning, at least most of the time. At our school, 97.3% of the students receive free or reduced pric e lunch. Of the 560 students, 97.3% are minority students. \r\nThe school has a vibrant community that loves to get togethe r and celebrate. Around Halloween there is a whole school para de to show off the beautiful costumes that students wear. On C

inco de Mayo we put on a big festival with crafts made by the students, dances, and games. At the end of the year the school hosts a carnival to celebrate the hard work put in during the school year, with a dunk tank being the most popular activity. My students will use these five brightly colored Hokki stools in place of regular, stationary, 4-legged chairs. As I will on ly have a total of ten in the classroom and not enough for eac h student to have an individual one, they will be used in a va riety of ways. During independent reading time they will be us ed as special chairs students will each use on occasion. I wil l utilize them in place of chairs at my small group tables dur ing math and reading times. The rest of the day they will be u sed by the students who need the highest amount of movement in their life in order to stay focused on school.\r\n\r\nWhenever asked what the classroom is missing, my students always say mo re Hokki Stools. They can't get their fill of the 5 stools we already have. When the students are sitting in group with me o n the Hokki Stools, they are always moving, but at the same ti me doing their work. Anytime the students get to pick where th ey can sit, the Hokki Stools are the first to be taken. There are always students who head over to the kidney table to get o ne of the stools who are disappointed as there are not enough of them. $\r\n\$ ask a lot of students to sit for 7 hours a day. The Hokki stools will be a compromise that allow my stude nts to do desk work and move at the same time. These stools wi 11 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 s till.nannan

How do you remember your days of school? Was it in a sterile e nvironment with plain walls, rows of desks, and a teacher in f ront of the room? A typical day in our room is nothing like th at. I work hard to create a warm inviting themed room for my s tudents 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 ther e 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 classroom s. These 9 and 10 year-old students are very eager learners; t hey are like sponges, absorbing all the information and experi ences and keep on wanting more. With these resources such as th e comfy red throw pillows and the whimsical nautical hanging d ecor 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 e nvironment. Creating a classroom environment is very important in the success in each and every child's education. The nautic al photo props will be used with each child as they step foot into our classroom for the first time on Meet the Teacher even

ing. I'll take pictures of each child with them, have them dev eloped, and then hung in our classroom ready for their first d ay of 4th grade. This kind gesture will set the tone before e ven 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 a fun, inviting, learning environment from day one.\r\n\r\nIt costs lost of mone y out of my own pocket on resources to get our classroom read y. Please consider helping with this project to make our new s chool year a very successful one. Thank you!nannan

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

The mediocre teacher tells. The good teacher explains. The sup erior teacher demonstrates. The great teacher inspires. -Willi am A. Ward\r\n\r\nMy school has 803 students which is makeup i s 97.6% African-American, making up the largest segment of the student body. A typical school in Dallas is made up of 23.2% A frican-American students. Most of the students are on free or reduced lunch. We aren't receiving doctors, lawyers, or engine ers children from rich backgrounds or neighborhoods. As an edu cator I am inspiring minds of young children and we focus not only on academics but one smart, effective, efficient, and dis ciplined students with good character. In our classroom we can utilize the Bluetooth for swift transitions during class. I us e a speaker which doesn't amplify the sound enough to receive the message. Due to the volume of my speaker my students can't hear videos or books clearly and it isn't making the lessons a s meaningful. But with the bluetooth speaker my students will

be able to hear and I can stop, pause and replay it at any tim e.\r\nThe cart will allow me to have more room for storage of things that are needed for the day and has an extra part to it I can use. The table top chart has all of the letter, words a nd pictures for students to learn about different letters and it is more accessible.nannan

In [52]:

```
# 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"\'re", " is", phrase)
    phrase = re.sub(r"\'d", " would", phrase)
    phrase = re.sub(r"\'d", " would", 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"\'re", " am", phrase)
    return phrase
```

In [53]:

formatted_essay = decontracted(training_data['essay'].values[20000])
print(formatted essay)

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

In [54]:

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

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

In [55]:

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

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

In [56]:

```
In [57]:
# Combining all the above statemennts
from tqdm import tqdm
preprocessed_essays = []
# tqdm is for printing the status bar
for sentance in tqdm(training_data['essay'].values):
    sent = decontracted(sentance)
   sent = sent.replace('\\r', ' ')
   sent = sent.replace('\\"',
   sent = sent.replace('\\n', ' ')
   sent = re.sub('[^A-Za-z0-9]+', ' ', sent)
    sent = ' '.join(e for e in sent.split() if e not in stopwords)
    preprocessed_essays.append(sent.lower().strip())
100%
               | 109248/109248 [00:53<00:00, 2055.54it/s]
In [58]:
# after preprocesing
print(len(preprocessed_essays))
109248
In [59]:
training data.head(30)
Out[59]:
```

	Unnamed: 0	id	teacher_id	teacher_prefix	school_state
0	160221	p253737	c90749f5d961ff158d4b4d1e7dc665fc	MRS	IN
1	140945	p258326	897464ce9ddc600bced1151f324dd63a	MR	FL

In [60]:

```
# Combining all the above statemennts
from tqdm import tqdm
preprocessed_titles = []
# tqdm is for printing the status bar
for sentance in tqdm(training_data['project_title'].values):
    sent = decontracted(sentance)
    sent = sent.replace('\\r', ' ')
    sent = sent.replace('\\r', ' ')
    sent = sent.replace('\\n', ' ')
    sent = re.sub('[^A-Za-z0-9]+', ' ', sent)
    sent = ' '.join(e for e in sent.split() if e not in stopwords)
    preprocessed_titles.append(sent.lower().strip())
```

100%| 109248/109248 [00:02<00:00, 44584.12it/s]

```
In [61]:
```

```
print(preprocessed_titles[20000],'\n\n')
print(preprocessed_essays[20000])
training_data.values[20000]
```

we need to move it while we input it

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

Out[61]:

```
array([65303, 'p115814', 'ffa0035d53b0c5379720954131051f60', 'MRS', 'NJ', '2016-08-11 09:06:14', 'Grades_PreK_2', 'We Need To Move It While We Input It!',
```

'My kindergarten students have varied disabilities rang ing from speech and language delays, cognitive delays, gross/f ine motor delays, to autism. They are eager beavers and always strive to work their hardest working past their limitations. \\r\\n\\r\\nThe materials we have are the ones I seek out for my students. I teach in a Title I school where most of the stu dents receive free or reduced price lunch. Despite their disa bilities and limitations, my students love coming to school and come eager to learn and explore.',

"Have you ever felt like you had ants in your pants and you needed to groove and move as you were in a meeting? This is how my kids feel all the time. The want to be able to move as they learn or so they say. Wobble chairs are the answer and I love then because they develop their core, which enhances gross motor and in Turn fine motor skills. \\r\\nThey also want to learn through games, my kids don't want to sit and do workshee ts. They want to learn to count by jumping and playing. Physic al engagement is the key to our success. The number toss and color and shape mats can make that happen. My students will for get they are doing work and just have the fun a 6 year old deserves.",

nan, nan,

'My students need wobble chairs, number toss games and colors and shapes mats to make our learning fun, hands on and physically engaging!',

21, 1, 'Health_Sports SpecialNeeds',
'Health Wellness SpecialNeeds',

"My kindergarten students have varied disabilities rang ing from speech and language delays, cognitive delays, gross/f ine motor delays, to autism. They are eager beavers and always strive to work their hardest working past their limitations. my students. I teach in a Title I school where most of the stu dents receive free or reduced price lunch. Despite their disa bilities and limitations, my students love coming to school an d come eager to learn and explore. Have you ever felt like you had ants in your pants and you needed to groove and move as yo u were in a meeting? This is how my kids feel all the time. Th e want to be able to move as they learn or so they say. Wobble chairs are the answer and I love then because they develop the ir core, which enhances gross motor and in Turn fine motor ski lls. \\r\\nThey also want to learn through games, my kids do n't want to sit and do worksheets. They want to learn to count by jumping and playing. Physical engagement is the key to our success. The number toss and color and shape mats can make tha t happen. My students will forget they are doing work and just have the fun a 6 year old deserves.nannan",

171.94, 12, 0], dtype=object)

Preparing data for models

In [62]:

training_data.columns

Out[62]:

```
- 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
```

Vectorizing Categorical data

In [63]:

```
# 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()), lowercate vectorizer.fit(training_data['clean_categories'].values)
print(vectorizer.get_feature_names())

categories_one_hot = vectorizer.transform(training_data['clean_categories'].values)
print("Shape of matrix after one hot encodig ",categories_one_hot.shape)
```

```
['Warmth', 'Care_Hunger', 'History_Civics', 'Music_Arts', 'App liedLearning', 'SpecialNeeds', 'Health_Sports', 'Math_Scienc e', 'Literacy_Language']
Shape of matrix after one hot encodig (109248, 9)
```

In [64]:

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

sub_categories_one_hot = vectorizer.transform(training_data['clean_subcategories'])
print("Shape of matrix after one hot encodig ",sub_categories_one_hot.shape)
```

```
['Economics', 'CommunityService', 'FinancialLiteracy', 'Parent Involvement', 'Extracurricular', 'Civics_Government', 'Foreign Languages', 'NutritionEducation', 'Warmth', 'Care_Hunger', 'So cialSciences', 'PerformingArts', 'CharacterEducation', 'TeamSp orts', 'Other', 'College_CareerPrep', 'Music', 'History_Geogra phy', 'Health_LifeScience', 'EarlyDevelopment', 'ESL', 'Gym_Fitness', 'EnvironmentalScience', 'VisualArts', 'Health_Wellness', 'AppliedSciences', 'SpecialNeeds', 'Literature_Writing', 'Mathematics', 'Literacy']
Shape of matrix after one hot encodig (109248, 30)
```

In [65]:

```
# count of all the words in corpus python: https://stackoverflow.com/a/228985
from collections import Counter
state_counter = Counter()
for word in training_data['school_state'].values:
    state_counter.update(word.split())
```

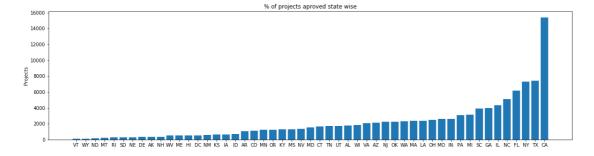
In [66]:

```
# dict sort by value python: https://stackoverflow.com/a/613218/4084039
school_state_dict = dict(state_counter)
sorted_school_state_dict = dict(sorted(school_state_dict.items(), key=lambda
```

In [67]:

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

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



In [68]:

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

VT 80 WY 98 : ND : 143 MT : 245 RΙ 285 SD 300 NE309 DE 343 ΑK : 345 NH348 WV: 503 ME 505 ΗI 507 : DC : 516 NM 557 KS : 634 IΑ : 666 ID 693 AR 1049 : CO : 1111 MN 1208 OR : 1242 ΚY 1304 MS : 1323 NV1367 MD 1514 : CT : 1663 TN 1688 UT : 1731 AL: 1762 WI : 1827 VA 2045 : AZ 2147 NJ 2237 OK : 2276 WA2334 MΑ : 2389 LA 2394 ОН 2467 MO : 2576 IN 2620 PΑ 3109 ΜI 3161 : SC 3936 GΑ 3963 ΙL 4350 NC 5091

```
FL : 6185
NY : 7318
TX : 7396
CA : 15388
```

In [69]:

```
vectorizer = CountVectorizer(vocabulary=list(sorted_school_state_dict.keys())
vectorizer.fit(training_data['school_state'].values)
print(vectorizer.get_feature_names())

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

```
['VT', 'WY', 'ND', 'MT', 'RI', 'SD', 'NE', 'DE', 'AK', 'NH', 'WV', 'ME', 'HI', 'DC', 'NM', 'KS', 'IA', 'ID', 'AR', 'CO', 'M N', 'OR', 'KY', 'MS', 'NV', 'MD', 'CT', 'TN', 'UT', 'AL', 'W I', 'VA', 'AZ', 'NJ', 'OK', 'WA', 'MA', 'LA', 'OH', 'MO', 'I N', 'PA', 'MI', 'SC', 'GA', 'IL', 'NC', 'FL', 'NY', 'TX', 'C A']
Shape of matrix after one hot encodig (109248, 51)
```

Vectorizing teacher_prefix

In [70]:

```
# count of all the words in corpus python: https://stackoverflow.com/a/228985
from collections import Counter
teacher_prefix_counter = Counter()
for word in training_data['teacher_prefix'].values:
    teacher_prefix_counter.update(word.split())
```

In [71]:

```
# dict sort by value python: https://stackoverflow.com/a/613218/4084039
teacher_prefix_dict = dict(teacher_prefix_counter)
sorted_teacher_prefix_dict = dict(sorted(teacher_prefix_dict.items(), key=lar
```

```
In [72]:
```

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

DR : 13
TEACHER : 2360
MR : 10648
MS : 38955
MRS : 57272

In [73]:

```
vectorizer = CountVectorizer(vocabulary=list(sorted_teacher_prefix_dict.keys(
vectorizer.fit(training_data['teacher_prefix'].values)
print(vectorizer.get_feature_names())

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

```
['DR', 'TEACHER', 'MR', 'MS', 'MRS']
Shape of matrix after one hot encodig (109248, 5)
```

Vectorizing project_grade_category

In [74]:

```
project_grade_counter = Counter()
for word in training_data['project_grade_category'].values:
    project_grade_counter.update(word.split())
```

In [75]:

```
project_grade_dict = dict(project_grade_counter)
sorted_project_grade_dict = dict(sorted(project_grade_dict.items(), key=lambo
```

In [76]:

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

Grades_9_12 : 10963 Grades_6_8 : 16923 Grades_3_5 : 37137 Grades_PreK_2 : 44225

```
In [77]:
```

```
vectorizer = CountVectorizer(vocabulary=list(sorted_project_grade_dict.keys()
vectorizer.fit(training_data['project_grade_category'].values)
print(vectorizer.get_feature_names())

project_grade_category_one_hot = vectorizer.transform(training_data['project_print("Shape of matrix after one hot encodig ",project_grade_category_one_hot
```

```
['Grades_9_12', 'Grades_6_8', 'Grades_3_5', 'Grades_PreK_2'] Shape of matrix after one hot encodig (109248, 4)
```

Vectorizing Text data

1.4.2.1

In [78]:

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

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

In [79]:

```
# We are considering only the words which appeared in at least 10 documents(r
vectorizer = CountVectorizer(min_df=10)
titles_bow = vectorizer.fit_transform(preprocessed_titles)
print("Shape of matrix after one hot encodig ",titles_bow.shape)
```

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

In [80]:

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

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

In [81]:

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

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

In [82]:

```
'''# Reading glove vectors in python: https://stackoverflow.com/a/38230349/40
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')
```

Out[82]:

```
'# Reading glove vectors in python: https://stackoverflow.com/
a/38230349/4084039\ndef (https://stackoverflow.com/a/38230349/
4084039\ndef) loadGloveModel(gloveFile):\n
                                              print ("Loading
Glove Model")\n
                   f = open(gloveFile,\'r\', encoding="utf8")
\n
      model = {}\n
                     for line in tqdm(f):\n
                                                    splitLine
= line.split()\n
                        word = splitLine[0]\n
                                                      embeddin
g = np.array([float(val) for val in splitLine[1:]])\n
                           print ("Done.",len(model)," words
odel[word] = embedding\n
                return model\nmodel = loadGloveModel(\'glove.4
loaded!")\n
2B.300d.txt\')\n\n'
```

In [83]:

```
'''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 col
      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))'''
```

Out[83]:

'words = []\nfor i in preprocessed_essays:\n words.extend
(i.split(\' \'))\n\nfor i in preprocessed_titles:\n words.e
xtend(i.split(\' \'))\nprint("all the words in the coupus", le
n(words))\nwords = set(words)\nprint("the unique words in the
coupus", len(words))\n\ninter_words = set(model.keys()).inters
ection(words)\nprint("The number of words that are present in
both glove vectors and our coupus", len(inter_words),"
(",np.round(len(inter_words)/len(words)*100,3),"%)")\n\nwords_
courpus = {}\nwords_glove = set(model.keys())\nfor i in word
s:\n if i in words_glove:\n words_courpus[i] = model
[i]\nprint("word 2 vec length", len(words_courpus))'

In [84]:

```
'''# stronging variables into pickle files python: http://www.jessicayung.com
import pickle
with open('glove_vectors', 'wb') as f:
    pickle.dump(words_courpus, f)
'''
```

Out[84]:

```
"# stronging variables into pickle files python: http://www.jessicayung.com/how-to-use-pickle-to-save-and-load-variables-in-python/\n\nimport (http://www.jessicayung.com/how-to-use-pickle-to-save-and-load-variables-in-python/\n\nimport) pickle\nwith open('glove_vectors', 'wb') as f:\n pickle.dump(words_courpus, f)\n"
```

In [85]:

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

In [86]:

```
# average Word2Vec
# compute average word2vec for each review.
avg_w2v_vectors = []; # the avg-w2v for each sentence/review is stored in the
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%|

| 109248/109248 [00:26<00:00, 4056.67it/s]

109248 300

In [87]:

100%| 109248/109248 [00:01<00:00, 81507.06it/s]

109248 300

Using Pretrained Models: TFIDF weighted W2V

In [88]:

```
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 [89]:

```
# average Word2Vec
# compute average word2vec for each review.
tfidf w2v vectors = []; # the avg-w2v for each sentence/review is stored in t
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/revi
   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
            tf idf = dictionary[word]*(sentence.count(word)/len(sentence.spli
            vector += (vec * tf_idf) # calculating tfidf weighted w2v
            tf idf weight += tf idf
   if tf_idf_weight != 0:
        vector /= tf_idf_weight
   tfidf w2v vectors.append(vector)
print(len(tfidf w2v vectors))
print(len(tfidf_w2v_vectors[0]))
```

100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%|

109248 300

In [90]:

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

In [91]:

```
# average Word2Vec
# compute average word2vec for each review.
tfidf w2v vectors titles = []; # the avg-w2v for each sentence/review is stor
for sentence in tqdm(preprocessed titles): # for each review/sentence
   vector = np.zeros(300) # as word vectors are of zero Length
   tf idf weight =0; # num of words with a valid vector in the sentence/revi
   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
            tf idf = dictionary[word]*(sentence.count(word)/len(sentence.spli
            vector += (vec * tf_idf) # calculating tfidf weighted w2v
            tf idf weight += tf idf
   if tf_idf_weight != 0:
        vector /= tf_idf_weight
   tfidf_w2v_vectors_titles.append(vector)
print(len(tfidf w2v vectors titles))
print(len(tfidf_w2v_vectors_titles[0]))
```

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

109248 300

In [92]:

```
# standardization sklearn: https://scikit-learn.org/stable/modules/generated/
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 3]
# Reshape your data either using array.reshape(-1, 1)

price_scalar = StandardScaler()
price_scalar.fit(training_data['price'].values.reshape(-1,1)) # finding the material print(f"Mean : {price_scalar.mean_[0]}, Standard deviation : {np.sqrt(price_scalar.mean_standardize the data with above maen and variance.
price_standardized = price_scalar.transform(training_data['price'].values.res
```

Mean: 298.1193425966608, Standard deviation: 367.49634838483

In [93]:

```
# standardization sklearn: https://scikit-learn.org/stable/modules/generated/
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 3
# Reshape your data either using array.reshape(-1, 1)

teacher_noppp_scalar = StandardScaler()
teacher_noppp_scalar.fit(training_data['teacher_number_of_previously_posted_print(f"Mean : {teacher_noppp_scalar.mean_[0]}, Standard deviation : {np.sqrf}
# Now standardize the data with above maen and variance.
teacher_noppp_standardized = teacher_noppp_scalar.transform(training_data['teacher_noppp_scalar.transform(training_data['teacher_noppp_scalar.transform(training_data['teacher_noppp_scalar.transform(training_data['teacher_noppp_scalar.transform(training_data['teacher_noppp_scalar.transform(training_data['teacher_noppp_scalar.transform(training_data['teacher_noppp_scalar.transform(training_data['teacher_noppp_scalar.transform(training_data['teacher_noppp_scalar.transform(training_data['teacher_noppp_scalar.transform(training_data['teacher_noppp_scalar.transform(training_data['teacher_noppp_scalar.transform(training_data['teacher_noppp_scalar.transform(training_data['teacher_noppp_scalar.transform(training_data['teacher_noppp_scalar.transform(training_data['teacher_noppp_scalar.transform(training_data['teacher_noppp_scalar.transform(training_data['teacher_noppp_scalar.transform(training_data['teacher_noppp_scalar.transform(training_data['teacher_noppp_scalar.transform(training_data['teacher_noppp_scalar.transform(training_data['teacher_noppp_scalar.transform(training_data['teacher_noppp_scalar.transform(training_data['teacher_noppp_scalar.transform(training_data['teacher_noppp_scalar.transform(training_data['teacher_noppp_scalar.transform(training_data['teacher_noppp_scalar.transform(training_data['teacher_noppp_scalar.transform(training_data['teacher_noppp_scalar.transform(training_data['teacher_noppp_scalar
```

Mean : 11.153165275336848, Standard deviation : 367.4963483848 3496

Merging all the above features

In [94]:

```
print('School State ',school_state_one_hot.shape)
print('Teacher Prefix ',teacher_prefix_one_hot.shape)
print('Project Grade ',project_grade_category_one_hot.shape)
print('Title ',titles_bow.shape)
print('Essay ',essays_bow.shape)
print('Categories ',categories_one_hot.shape)
print('Sub Categories',sub_categories_one_hot.shape)
print('Price',price_standardized.shape)
print('Teacher number of previously posted projects',teacher_noppp_standardized.shape)
```

```
School State (109248, 51)
Teacher Prefix (109248, 5)
Project Grade (109248, 4)
Title (109248, 3329)
Essay (109248, 16623)
Categories (109248, 9)
Sub Categories (109248, 30)
Price (109248, 1)
Teacher number of previously posted projects (109248, 1)
```

```
In [95]:
```

```
print('Title BOW', titles_bow.shape)
print('Essay BOW', essays_bow.shape)
print('Title TFIDF',titles_tfidf.shape)
print('Essay TFIDF',essays tfidf.shape)
print('Title AVG W2V (',len(avg_w2v_vectors_titles),',',len(avg_w2v_vectors_t
print('Essay AVG W2V (',len(avg_w2v_vectors),',',len(avg_w2v_vectors[0]),')')
print('Essay TFIDF AVG W2V (',len(tfidf w2v vectors),',',len(tfidf w2v vector
Title BOW (109248, 3329)
Essay BOW (109248, 16623)
Title TFIDF (109248, 3329)
Essay TFIDF (109248, 16623)
Title AVG W2V ( 109248 , 300 )
Essay AVG W2V ( 109248 , 300 )
Title TFIDF AVG W2V ( 109248 , 300 )
Essay TFIDF AVG W2V ( 109248 , 300 )
In [96]:
# merge two sparse matrices: https://stackoverflow.com/a/19710648/4084039
from scipy.sparse import hstack
# with the same hstack function we are concatinating a sparse matrix and a d\epsilon
X BOW = hstack((school state one hot, teacher prefix one hot, project grade (
           categories one hot, sub categories one hot, price standardized,
           teacher noppp standardized, titles bow)).tocsr()
X_BOW.shape
Out[96]:
(109248, 3430)
In [97]:
X_BOW_5000 = X_BOW[0:5000,:]
X BOW 5000.shape
Out[97]:
(5000, 3430)
```

```
In [98]:
X_BOW_1000 = X_BOW[0:1000,:]
X_BOW_1000.shape
Out[98]:
(1000, 3430)
In [99]:
X_BOW_10000 = X_BOW[0:10000,:]
X_BOW_10000.shape
Out[99]:
(10000, 3430)
In [100]:
labels = training_data['project_is_approved']
labels.shape
Out[100]:
(109248,)
In [101]:
labels 1000 = labels[0: 1000]
labels_1000.shape
Out[101]:
(1000,)
In [102]:
labels_5000 = labels[0: 5000]
labels_5000.shape
Out[102]:
(5000,)
```

In [103]:

```
labels_10000 = labels[0: 10000]
labels_10000.shape
```

Out[103]:

(10000,)

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

In [104]:

```
from sklearn.manifold import TSNE

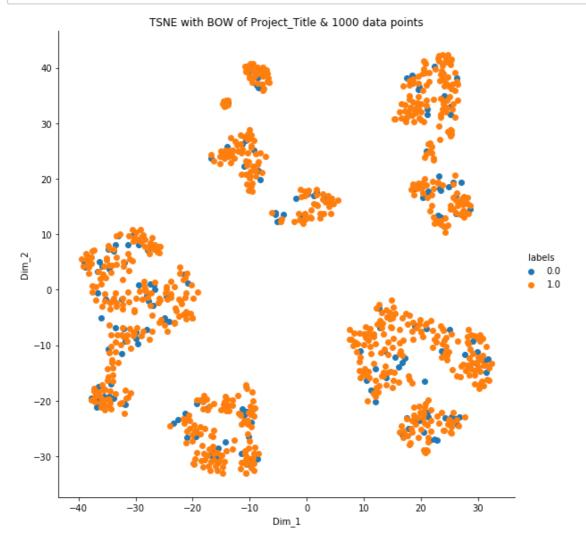
x = X_BOW_1000
y = labels_1000

model = TSNE(n_components=2, perplexity=30, learning_rate=300, random_state=0)

X_embedding = model.fit_transform(x.toarray())
tsne_data = model.fit_transform(X_embedding)

# creating a new data frame which help us in ploting the result data
tsne_data = np.vstack((tsne_data.T, y)).T
tsne_df = pd.DataFrame(tsne_data, columns=("Dim_1", "Dim_2", "labels"))

# Ploting the result of tsne
sns.FacetGrid(tsne_df, hue="labels", height=8).map(plt.scatter, 'Dim_1', 'Dir
plt.title('TSNE with BOW of Project_Title & 1000 data points')
plt.show()
```



In [105]:

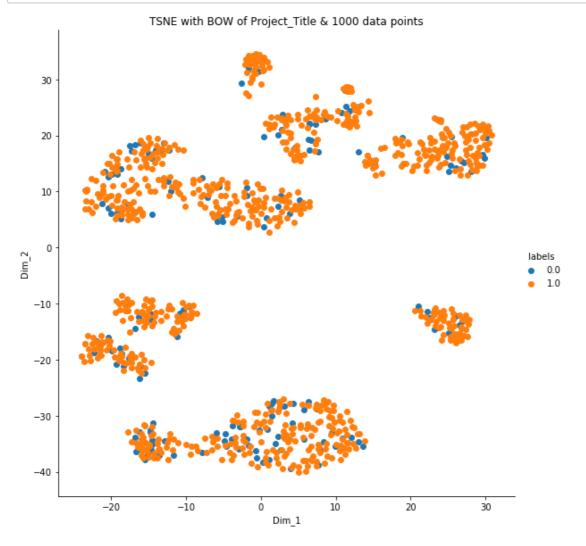
```
from sklearn.manifold import TSNE

x = X_BOW_1000
y = labels_1000

model = TSNE(n_components=2, perplexity=40, learning_rate=300, random_state=0
X_embedding = model.fit_transform(x.toarray())
tsne_data = model.fit_transform(X_embedding)

# creating a new data frame which help us in ploting the result data
tsne_data = np.vstack((tsne_data.T, y)).T
tsne_df = pd.DataFrame(tsne_data, columns=("Dim_1", "Dim_2", "labels"))

# Ploting the result of tsne
sns.FacetGrid(tsne_df, hue="labels", height=8).map(plt.scatter, 'Dim_1', 'Dir
plt.title('TSNE with BOW of Project_Title & 1000 data points')
plt.show()
```



In [106]:

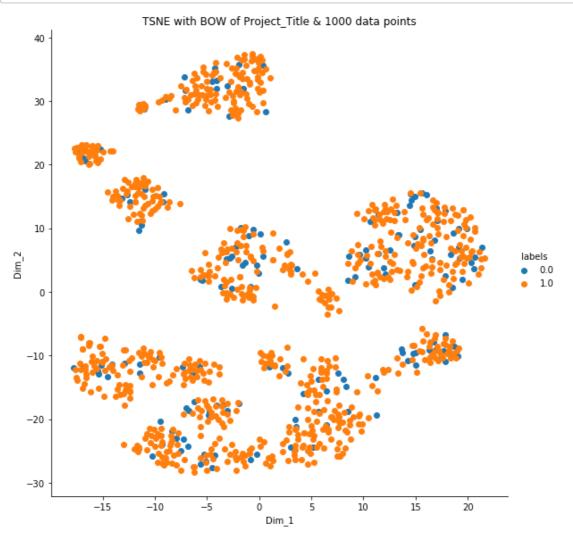
```
from sklearn.manifold import TSNE

x = X_BOW_1000
y = labels_1000

model = TSNE(n_components=2, perplexity=50, learning_rate=300, random_state=0
X_embedding = model.fit_transform(x.toarray())
tsne_data = model.fit_transform(X_embedding)

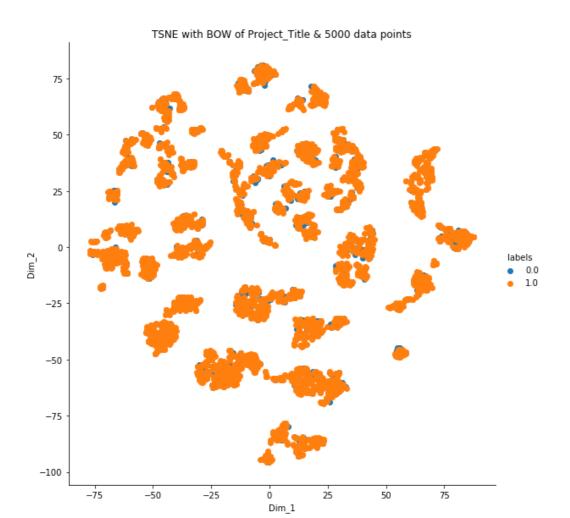
# creating a new data frame which help us in ploting the result data
tsne_data = np.vstack((tsne_data.T, y)).T
tsne_df = pd.DataFrame(tsne_data, columns=("Dim_1", "Dim_2", "labels"))

# Ploting the result of tsne
sns.FacetGrid(tsne_df, hue="labels", height=8).map(plt.scatter, 'Dim_1', 'Dir
plt.title('TSNE with BOW of Project_Title & 1000 data points')
plt.show()
```



In [107]:

```
# TSNE with 5000 data points
X_BOW_1000 = X_BOW[0:1000,:]
X BOW 1000.shape
from sklearn.manifold import TSNE
# Picking the top 1000 points as TSNE takes a lot of time for 15K points
x = X BOW 5000
y = labels 5000
model = TSNE(n_components=2, perplexity=30, learning_rate=300, random_state={
X embedding = model.fit_transform(x.toarray())
# configuring the parameteres
# the number of components = 2
# default perplexity = 30
# default learning rate = 200
# default Maximum number of iterations for the optimization = 1000
tsne data = model.fit transform(X embedding)
# creating a new data frame which help us in ploting the result data
tsne data = np.vstack((tsne data.T, y)).T
tsne_df = pd.DataFrame(tsne_data, columns=("Dim_1", "Dim_2", "labels"))
# Ploting the result of tsne
sns.FacetGrid(tsne_df, hue="labels", height=8).map(plt.scatter, 'Dim_1', 'Dir
plt.title('TSNE with BOW of Project Title & 5000 data points')
plt.show()
```



In [108]:

```
# TSNE with 10000 data points
from sklearn.manifold import TSNE
# Picking the top 1000 points as TSNE takes a lot of time for 15K points
x = X_BOW_10000
y = labels 10000
model = TSNE(n_components=2, perplexity=30, learning_rate=300, random_state={
X embedding = model.fit transform(x.toarray())
# configuring the parameteres
# the number of components = 2
# default perplexity = 30
# default learning rate = 200
# default Maximum number of iterations for the optimization = 1000
tsne data = model.fit transform(X embedding)
# creating a new data frame which help us in ploting the result data
tsne data = np.vstack((tsne data.T, y)).T
tsne df = pd.DataFrame(tsne data, columns=("Dim 1", "Dim 2", "labels"))
# Ploting the result of tsne
sns.FacetGrid(tsne_df, hue="labels", height=8).map(plt.scatter, 'Dim_1', 'Dir
plt.title('TSNE with BOW of Project_Title & 10000 data points')
plt.show()
```

- · Vectorized categorical and numerical features and stacked it with title BOW vector
- Plotted TSNE with different Perplexity and Random State values for 1000 data points
- After experimenting with different combinations, choosed the final values of perplexity and random_state as 30 and 300 respectively.
- Plotted TSNE with all categorical, numerical features and Title BOW vector for 1K, 5K and 10K data points
- Observed that the accepted projects are more than the rejected projects.
- · Accepted and Rejected classes are heavily overlaping and not seperated at this point

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

In [109]:

In [110]:

```
from sklearn.manifold import TSNE

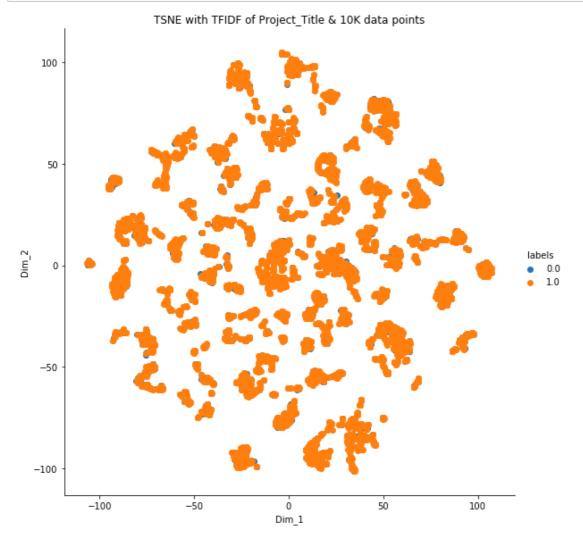
x = X_TFIDF_10000
y = labels_10000

model = TSNE(n_components=2, perplexity=30, learning_rate=300, random_state=0
X_embedding = model.fit_transform(x.toarray())

tsne_data = model.fit_transform(X_embedding)

# creating a new data frame which help us in ploting the result data
tsne_data = np.vstack((tsne_data.T, y)).T
tsne_df = pd.DataFrame(tsne_data, columns=("Dim_1", "Dim_2", "labels"))

# Ploting the result of tsne
sns.FacetGrid(tsne_df, hue="labels", height=8).map(plt.scatter, 'Dim_1', 'Dir
plt.title('TSNE with TFIDF of Project_Title & 10K data points')
plt.show()
```

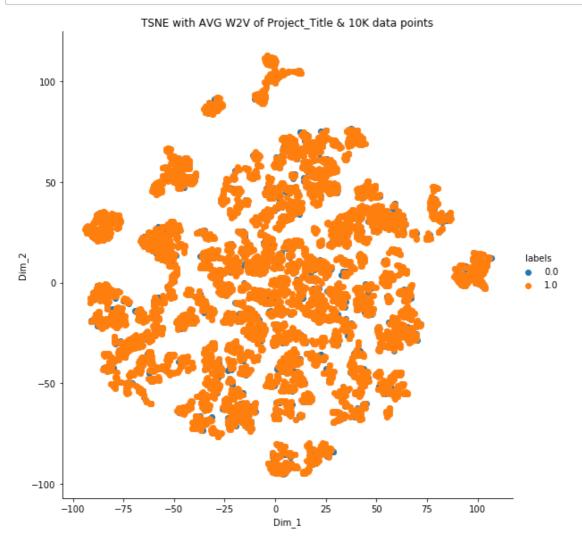


- Plotted TSNE with all categorical, numerical features and Title TFIDF vector for 10K data points
- Observed that the accepted projects are more than the rejected projects.
- · Accepted and Rejected classes are heavily overlaping and not seperated at this point
- · Didn't find any major difference when compared to TSNE BOW title vector

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

In [111]:

In [112]:



- Plotted TSNE with all categorical and numerical features and Title AVG W2V vector for 10K data points
- Observed that the accepted projects are more than the rejected projects.
- · Accepted and Rejected classes are heavily overlaping and not seperated at this point
- Found that the plot has dense data points when compared to TSNE of BOW and TFIDF title vectors

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

In [113]:

In [114]:

```
from sklearn.manifold import TSNE

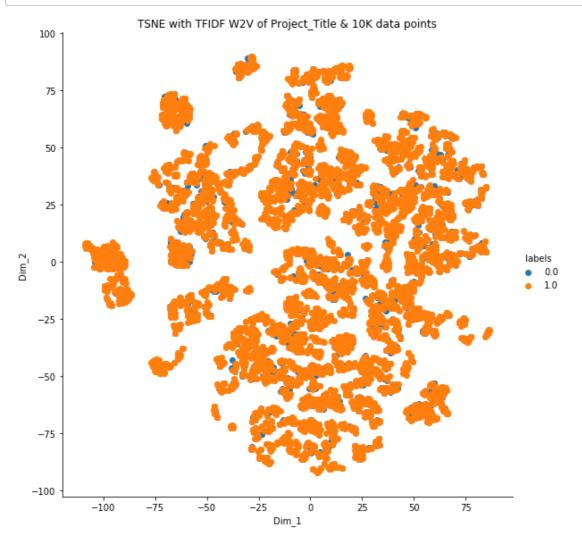
x = X_TFIDF_W2V_10000
y = labels_10000

model = TSNE(n_components=2, perplexity=30, learning_rate=300, random_state=0
X_embedding = model.fit_transform(x.toarray())

tsne_data = model.fit_transform(X_embedding)

# creating a new data frame which help us in ploting the result data
tsne_data = np.vstack((tsne_data.T, y)).T
tsne_df = pd.DataFrame(tsne_data, columns=("Dim_1", "Dim_2", "labels"))

# Ploting the result of tsne
sns.FacetGrid(tsne_df, hue="labels", height=8).map(plt.scatter, 'Dim_1', 'Dir
plt.title('TSNE with TFIDF W2V of Project_Title & 10K data points')
plt.show()
```

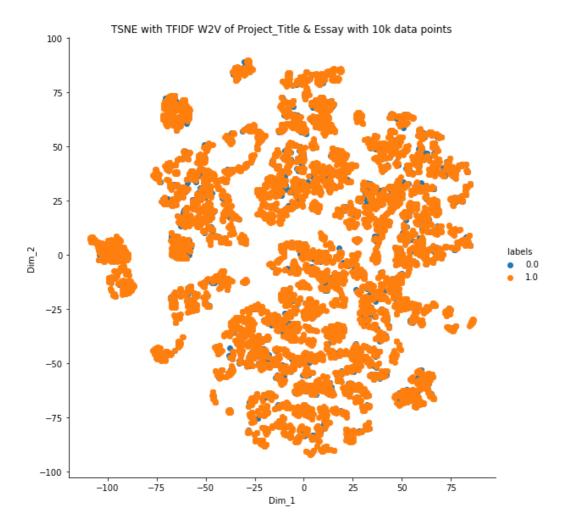


- Plotted TSNE with all categorical and numerical features and Title TFIDF W2V vector for 10K data points
- Observed that the accepted projects are more than the rejected projects.
- · Accepted and Rejected classes are heavily overlaping and not seperated at this point
- Found that the plot has dense data points when compared to TSNE of BOW and TFIDF title vectors, but no major difference when compared to TSNE of AVG W2V Title vector

2.4.1 TSNE with 'TFIDF Weighted W2V' encoding of 'project_title' and 'project_essay' features (Extra)

In [115]:

```
X_TFIDF_W2V_E = hstack((school_state_one_hot, teacher_prefix_one_hot, project
            categories_one_hot, sub_categories_one_hot, price_standardized,
            teacher noppp standardized, tfidf w2v vectors, tfidf w2v vectors
X_TFIDF_W2V_E_10000 = X_TFIDF_W2V_E[0:10000,:]
from sklearn.manifold import TSNE
x = X TFIDF W2V 10000
y = labels_10000
model = TSNE(n_components=2, perplexity=30, learning_rate=300, random_state={
X_embedding = model.fit_transform(x.toarray())
tsne data = model.fit transform(X embedding)
# creating a new data frame which help us in ploting the result data
tsne data = np.vstack((tsne data.T, y)).T
tsne df = pd.DataFrame(tsne data, columns=("Dim 1", "Dim 2", "labels"))
# Ploting the result of tsne
sns.FacetGrid(tsne_df, hue="labels", height=8).map(plt.scatter, 'Dim_1', 'Dir
plt.title('TSNE with TFIDF W2V of Project_Title & Essay with 10k data points
plt.show()
```

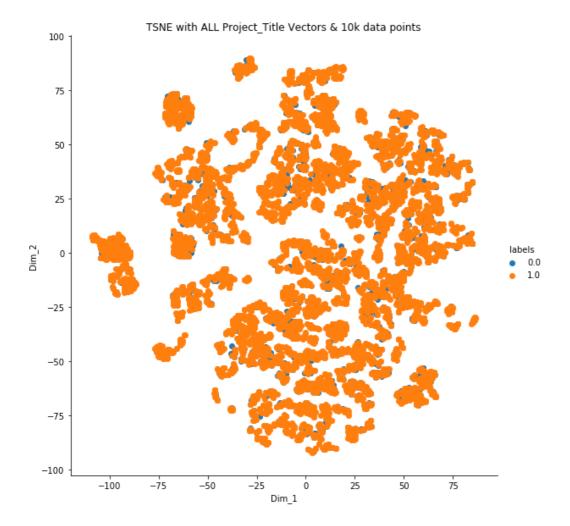


- Plotted TSNE with all categorical and numerical features, Title TFIDF W2V vector and Essay TFIDF W2V vector for 10K data points
- Observed that the accepted projects are more than the rejected projects.
- Accepted and Rejected classes are heavily overlaping and not seperated at this point
- Found that the plot has dense data points when compared to TSNE of BOW and TFIDF title vectors, but no major difference when compared to TSNE of AVG W2V Title vector and TSNE of Title TFIDF W2V vector

2.4 TSNE with all encoding of 'project_title' feature

In [116]:

```
X_ALL_TITLES = hstack((school_state_one_hot, teacher_prefix_one_hot, project)
            categories one hot, sub categories one hot, price standardized,
            teacher noppp standardized, titles bow, titles tfidf, avg w2v vect
            tfidf w2v vectors titles)).tocsr()
X ALL TITLES 10000 = X ALL TITLES[0:10000,:]
from sklearn.manifold import TSNE
x = X_TFIDF_W2V_10000
y = labels 10000
model = TSNE(n_components=2, perplexity=30, learning_rate=300, random_state=
X embedding = model.fit_transform(x.toarray())
tsne_data = model.fit_transform(X_embedding)
# creating a new data frame which help us in ploting the result data
tsne data = np.vstack((tsne data.T, y)).T
tsne_df = pd.DataFrame(tsne_data, columns=("Dim_1", "Dim_2", "labels"))
# Ploting the result of tsne
sns.FacetGrid(tsne_df, hue="labels", height=8).map(plt.scatter, 'Dim_1', 'Dir
plt.title('TSNE with ALL Project_Title Vectors & 10k data points')
plt.show()
```



- Plotted TSNE with all categorical and numerical features and all variations of Title vector for 10K data points
- Observed that the accepted projects are more than the rejected projects.
- Accepted and Rejected classes are heavily overlaping and not seperated at this point
- Found that the plot has dense data points when compared to TSNE of BOW and TFIDF title vectors, but no major difference when compared to TSNE of AVG W2V Title vector and TSNE of Title TFIDF W2V vector

Summary

- The objective is to automate project approval task using the text and meta data of the project
- 92706 projects were approved for funding out of 109248 applications i.e. 84.85 %
- 16542 projects were not approved for funding out of 109248 applications i.e. 15.14%
- Looks like an imbalanced dataset
- Removed np.nan values from teacher prefix column and converted to MRS
- Removed Full stops from teacher prefix column and converted to Upper case
- Ploted all the 51 states of United States of America with rate of acceptance using plotly.graph objs

- Delaware (DE) is the state with Most Acceptance percentage i.e. 89.79 %
- North Dakota (ND) is the state with Second most Acceptance percentage i.e. 88.81 %
- Washington (WA) is the state with Third most Acceptance percentage i.e. 87.61 %
- Vermont state has the lowest percentage of acceptance i.e. 80.00 %
- Washington, D.C. has the second lowest percentage of acceptance i.e. 80.23 %
- Texas state has the third lowest percentage of acceptance i.e. 81.31 %
- Every state has greater than 80 % acceptance rate
- California state has the most number of project submissions i.e. 15388
- Texas state has the second most number of project submissions i.e. 7396
- Wyoming state has the second least number of project submissions i.e. 98
- Vermont state has the least number of project submissions i.e. 80
- MRS teacher_prefix has the most number of project submissions i.e. 57272 with 85.55 % acceptance rate
- MS teacher_prefix has the second most number of project submissions i.e. 38955 with 84.35 % acceptance rate
- Teacher teacher_prefix has the second least number of project submissions i.e. 2360 with 79.53 % acceptance rate
- DR teacher_prefix has the least number of project submissions i.e. 13 with 69.23 % acceptance rate
- Most number of projects are for Grades pre K 2 i.e. 44225 with 84.87 % acceptance rate
- Second Most number of projects are for Grades 3-5 i.e. 37137 with 85.43 % acceptance rate
- Third Most number of projects are for Grades 6 8 i.e. 16923 with 84.25 % acceptance rate
- Least number of projects are for Grades 9 12 i.e. 10963 with 83.76 % acceptance rate
- Most number of projects are for Category Literacy & Language i.e. 23655 with 86.74 % acceptance rate
- Second Most number of projects are for Category Math & Science i.e. 17072 with 81.95
 % acceptance rate
- Third Most number of projects are for Category Literacy, Language, Math & Science i.e. 14636 with 0.86.94 % acceptance rate
- Least number of projects are for Category AppliedLearning, Math & Science i.e. 1052 with 81.27 % acceptance rate
- More than 50 % of accepted projects have less than 250 words in their essays
- More than 50 % of rejected projects have less than 250 words in their essays
- More than 75 % of accepted projects have less than 300 words in their essays
- More than 75 % of rejected projects have less than 300 words in their essays
- All the projects prices are between 0.66 to 9999.0
- 50 % of accepted project prices are under 198.99
- 50 % of rejected project prices are under 263.145
- · Accepted project prices are less when compared to rejected projects prices
- 27.46 % of teachers have posted the projects for the first time
- The first timers have posted 30014 projects with 82.13 % acceptance rate
- The maximum number of projects posted by any teacher is 451

- 93492 projects were submitted without any digits in their essays with 84.0885 % acceptance ratio
- 15756 projects were submitted without any digits in their essays with 89.4263 % acceptance ratio
- Projects with digits in their essays have 5.33 % more acceptance ratio than projects without digits
- · Vectorized categorical and numerical features and stacked it with title BOW vector
- Plotted TSNE with different Perplexity and Random State values for 1000 data points
- After experimenting with different combinations, choosed the final values of perplexity and random_state as 30 and 300 respectively.
- Plotted TSNE with all categorical, numerical features and Title BOW vector for 1K, 5K and 10K data points
- Plotted TSNE with all categorical, numerical features and Title TFIDF vector for 10K data points
- Didn't find any major difference when compared to TSNE of BOW title vector
- Plotted TSNE with all categorical and numerical features and Title AVG W2V vector for 10K data points
- Found that the plot has dense data points when compared to TSNE of BOW and TFIDF title vectors
- Plotted TSNE with all categorical and numerical features and Title TFIDF W2V vector for 10K data points
- Found that the plot has dense data points when compared to TSNE of BOW and TFIDF title vectors, but no major difference when compared to TSNE of AVG W2V Title vector
- Plotted TSNE with all categorical and numerical features, Title TFIDF W2V vector and Essay TFIDF W2V vector for 10K data points
- Found that the plot has dense data points when compared to TSNE of BOW and TFIDF title vectors, but no major difference when compared to TSNE of AVG W2V Title vector and TSNE of Title TFIDF W2V vector
- Plotted TSNE with all categorical and numerical features and all variations of Title vector for 10K data points
- Found that the plot has dense data points when compared to TSNE of BOW and TFIDF title vectors, but no major difference when compared to TSNE of AVG W2V Title vector and TSNE of Title TFIDF W2V vector
- The accepted and rejected points are heavily overlaping and not seperated at this point.