#### **Decision Tree Algorithm on Donors\_Choose dataset**

```
In [1]: | %matplotlib inline
        import warnings
        warnings.filterwarnings("ignore")
        import sqlite3
        import pandas as pd
        import numpy as np
        import nltk
        import gc
        gc.enable()
        gc.DEBUG SAVEALL
        import string
        import matplotlib.pyplot as plt
        import seaborn as sns
        from sklearn.feature extraction.text import TfidfTransformer
        from sklearn.feature extraction.text import TfidfVectorizer
        from sklearn.feature extraction.text import CountVectorizer
        from sklearn.metrics import confusion matrix
        from sklearn import metrics
        from sklearn.metrics import roc curve, auc
        from nltk.stem.porter import PorterStemmer
        import re
        # Tutorial about Python regular expressions: https://pymotw.com/2/re/
        import string
        from nltk.corpus import stopwords
        from nltk.stem import PorterStemmer
        from nltk.stem.wordnet import WordNetLemmatizer
        from gensim.models import Word2Vec
        from gensim.models import KeyedVectors
        import pickle
        from tqdm import tqdm
        import os
        import math
        from plotly import plotly
        import plotly.offline as offline
        import plotly.graph objs as go
        offline.init notebook mode()
        from collections import Counter
        gc.set threshold(2, 1, 1)
```

# 2.1 Loading Input Data

```
In [2]: # %load ext memory profiler
        # s=0
        # We are taking samples of 0's and 1's and appending them to overcome memory erro
        project data = pd.read csv('train data.csv')
        # project data=project data.dropna(how='any')
        # project_data_1 = project_data[project_data['project_is_approved'] == s+1]
        # project data 0 = project data[project data['project is approved'] == s]
        # project data=project data.fillna("")
        project data 1=project data.head(25000)
        project_data_0=project_data.tail(25000)
        project data 1=project data 1.append(project data 0)
        project data=project data 1
        resource_data = pd.read_csv('resources.csv')
        #Sorting them by columns to spread the zeros and one's unevenly in the 'project
        project data.sort values(by=['teacher number of previously posted projects'])
        # project_data.sort_values(by=['project_essay_4'], ascending=False)
        project data 1=None
        project data 0=None
In [3]:
        print("Number of data points in train data", project data.shape)
        print('-'*50)
        print("The attributes of data :", project data.columns.values)
        Number of data points in train data (50000, 17)
        The attributes of data : ['Unnamed: 0' 'id' 'teacher_id' 'teacher_prefix' 'scho
        ol state'
         'project subject categories' 'project subject subcategories'
         'project_title' 'project_essay_1' 'project_essay_2' 'project_essay_3'
         'project essay 4' 'project resource summary'
         'teacher_number_of_previously_posted_projects' 'project_is_approved']
In [4]: | print("Number of data points in resource data", resource_data.shape)
        print(resource data.columns.values)
        resource data.head(1)
        # project data.head(2)
        Number of data points in resource data (1541272, 4)
        ['id' 'description' 'quantity' 'price']
Out[4]:
                id
                                                description quantity price
         0 p233245 LC652 - Lakeshore Double-Space Mobile Drying Rack
                                                               1 149.0
```

Out[5]: 32

```
In [5]: y = project_data['project_is_approved'].values
    X = project_data.drop(['project_is_approved'], axis=1)
    X.head(1)
    project_data=None
    gc.collect()
    gc.enable()
    gc.DEBUG_SAVEALL
```

# 2.2 Getting the Data Model Ready:Preprocessing and Vectorizing categorical features

#### 2.2.1 Preprocessing:project\_grade\_category

```
In [6]: | sub_catogories = list(X['project_grade_category'].values)
        # remove special characters from list of strings python: https://stackoverflow.co
        # https://www.geeksforgeeks.org/removing-stop-words-nltk-python/
        # https://stackoverflow.com/questions/23669024/how-to-strip-a-specific-word-from
        # https://stackoverflow.com/questions/8270092/remove-all-whitespace-in-a-string-
        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 on s
                    j=j.replace('The','') # if we have the words "The" we are going to re
                j = j.replace(' ','') # we are placeing all the ' '(space) with ''(empty)
                temp +=j.strip()+" "#" abc ".strip() will return "abc", remove the trail
                temp = temp.replace('&','_')
            sub_cat_list.append(temp.strip())
        X['project_grade_category'] = sub_cat_list
```

```
In [7]:
    sub_catogories=None
    sub_cat_list=None
    temp=None
    i=None
    catogories=None
    cat_list=None
    temp=None
    my_counter=None
    word=None
    cat_dict=None
    gc.collect()
    gc.enable()
    gc.DEBUG_SAVEALL
```

Out[7]: 32

```
In [8]: | catogories = list(X['project_subject_categories'].values)
        # remove special characters from list of strings python: https://stackoverflow.co
        # https://www.geeksforgeeks.org/removing-stop-words-nltk-python/
        # https://stackoverflow.com/questions/23669024/how-to-strip-a-specific-word-from
        # https://stackoverflow.com/questions/8270092/remove-all-whitespace-in-a-string-
        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 on s
                    j=j.replace('The','') # if we have the words "The" we are going to re
                j = j.replace(' ','') # we are placeing all the ' '(space) with ''(empty)
                temp+=j.strip()+" " #" abc ".strip() will return "abc", remove the trail
                temp = temp.replace('&','_') # we are replacing the & value into
            cat list.append(temp.strip())
        X['clean categories'] = cat list
        X.drop(['project_subject_categories'], axis=1, inplace=True)
        X.head(2)
```

## Out[8]:

	Unnamed: 0	id	teacher_id	teacher_prefix	school_state	project_sul
0	160221	p253737	c90749f5d961ff158d4b4d1e7dc665fc	Mrs.	IN	
1	140945	p258326	897464ce9ddc600bced1151f324dd63a	Mr.	FL	
4						<b>&gt;</b>

#### 2.2.3 Preprocessing:project\_subject\_subcategories

```
In [9]: | sub catogories = list(X['project subject subcategories'].values)
        # remove special characters from list of strings python: https://stackoverflow.co
        # https://www.geeksforgeeks.org/removing-stop-words-nltk-python/
        # https://stackoverflow.com/questions/23669024/how-to-strip-a-specific-word-from
        # https://stackoverflow.com/questions/8270092/remove-all-whitespace-in-a-string-
        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 on s
                    j=j.replace('The','') # if we have the words "The" we are going to re
                j = j.replace(' ','') # we are placeing all the ' '(space) with ''(empty)
                temp +=j.strip()+" "#" abc ".strip() will return "abc", remove the trail
                temp = temp.replace('&','_')
            sub_cat_list.append(temp.strip())
```

```
In [10]: X['clean_subcategories'] = sub_cat_list
    X.drop(['project_subject_subcategories'], axis=1, inplace=True)
    X.head(2)
```

#### Out[10]:

	Unnamed: 0	id	teacher_id	teacher_prefix	school_state	project_sul
0	160221	p253737	c90749f5d961ff158d4b4d1e7dc665fc	Mrs.	IN	
1	140945	p258326	897464ce9ddc600bced1151f324dd63a	Mr.	FL	
4						<b>&gt;</b>

## 2.2.4 New Column: digits in summary

```
In [11]: # Creating a new column 'digits_in_summary' which contains flags of 1 for /
         # 'project resource summary' containing numeric specification in their requiremnt
         project_resource_summary = []
         new=[]
         project_resource_summary = list(X['project_resource_summary'].values)
         for i in project_resource_summary:
             # consider we have text like this "Math & Science, Warmth, Care & Hunger"
             for j in i.split(' '):
                 if j.isdigit():
                     new.append(1)
                     break
                 else:
                     continue
             else:
                 new.append(0)
         X['digits in summary']=new
         X.sort_values(by=['digits_in_summary'])
```

## Out[11]:

].		Unnamed: 0	id	teacher_id	teacher_prefix	school_state	p
	0	160221	p253737	c90749f5d961ff158d4b4d1e7dc665fc	Mrs.	IN	
	91629	8616	p220748	32c24ca620f89faba3107e4aafcdbf07	Ms.	CA	
	91630	9156	p201859	0edc5fc97906c1fb271e1ffe50dfae24	Mrs.	OR	
	91631	74015	p167529	701ea8412bb39370abb8ebec7ddd7ea2	Mrs.	NC	<b>~</b>

```
In [12]: #To make best use of the memory we are setting the variable names to 'None' and project_resource_summary=None
    new=None
    new1=None
    i=None
    j=None
    a=None

gc.collect()
gc.enable()
gc.DEBUG_SAVEALL
Out[12]: 32
```

# 2.2.5 Preprocessing:Text features (Project Essay's)

```
In [14]: X = X.drop(['project_essay_1', 'project_essay_2', 'project_essay_3', 'proje
```

# 2.2.6 Adding column Cost per project in dataset

```
In [15]: # https://stackoverflow.com/questions/22407798/how-to-reset-a-dataframes-indexes
price_data = resource_data.groupby('id').agg({'price':'sum', 'quantity':'sum'}).
price_data.head(2)
type(price_data)
```

Out[15]: pandas.core.frame.DataFrame

```
In [16]: # join two dataframes in python:
          X = pd.merge(X, price_data, on='id', how='left')
          X.sort_values(by=['price'])
          X.head(2)
Out[16]:
              Unnamed:
                             id
                                                      teacher_id teacher_prefix school_state project_sul
           0
                160221 p253737
                                  c90749f5d961ff158d4b4d1e7dc665fc
                                                                         Mrs.
                                                                                       IN
           1
                140945 p258326 897464ce9ddc600bced1151f324dd63a
                                                                          Mr.
                                                                                       FL
```

```
In [17]: #To make best use of the memory we are setting the variable names to 'None' and presource_data=None
    price_data=None
    gc.collect()
    gc.enable()
    gc.DEBUG_SAVEALL
Out[17]: 32
```

# 2.2.7 Text Preprocessing:Essay Text

```
In [18]:
          # https://stackoverflow.com/a/47091490/4084039
          import re
          def decontracted(phrase):
               # specific
               phrase = re.sub(r"won't", "will not", phrase)
               phrase = re.sub(r"can\'t", "can not", phrase)
               # general
               phrase = re.sub(r"n\'t", " not", phrase)
               phrase = re.sub(r"\'re", " are", phrase)
                                         <sup>" is"</sup>, phrase)
               phrase = re.sub(r"\'s",
               phrase = re.sub(r"\'d", " would", phrase)
               phrase = re.sub(r"\'ll", " will", phrase)
phrase = re.sub(r"\'t", " not", phrase)
               phrase = re.sub(r"\'ve", " have", phrase)
               phrase = re.sub(r"\'m", " am", phrase)
               return phrase
```

```
In [19]: sent = decontracted(X['essay'].values[99])
    print(sent)
    print("="*50)
```

My preschool students are children who are three to five years of age. My scho ol is in sunny San Pedro, California. The children from San Pedro come to schoo l each morning ready to learn and grow. There is never a dull moment in our cl ass; my students are busy bees moving from one interest area to another. They are eager to learn, explore, and experiment with the instructional materials an d centers I set up for them. We need more materials for the children to engage with, materials that will foster their interest in technology, literacy, math, science, art, and engineering. \r\nMy student is will learn number recognition and develop counting skills while engaging with the Learn to count picture puzz les and number Sequencing puzzles. While building with the 3-D Magnet Builders and Crystal Building Blocks, my student is mathematical skills will be supporte d and strengthened in concepts such as measurement, comparison, number estimati on, symmetry and balance. My student is will build number skills as the they si ft and make exciting number shell discoveries with every scoop at the sand tabl e. The sort a shape activity board will allow my youngest students to learn col ors, shapes and sorting skills as they fit various shape pieces into place.nann an

\_\_\_\_\_\_

```
In [20]: # \r \n \t remove from string python: http://texthandler.com/info/remove-line-browsent = sent.replace('\\r', ' ')
    sent = sent.replace('\\"', ' ')
    sent = sent.replace('\\n', ' ')
    print(sent)
```

My preschool students are children who are three to five years of age. My scho ol is in sunny San Pedro, California. The children from San Pedro come to schoo l each morning ready to learn and grow. There is never a dull moment in our cl ass; my students are busy bees moving from one interest area to another. They are eager to learn, explore, and experiment with the instructional materials an d centers I set up for them. We need more materials for the children to engage with, materials that will foster their interest in technology, literacy, math, science, art, and engineering. My student is will learn number recognition an d develop counting skills while engaging with the Learn to count picture puzzle s and number Sequencing puzzles. While building with the 3-D Magnet Builders an d Crystal Building Blocks, my student is mathematical skills will be supported and strengthened in concepts such as measurement, comparison, number estimatio n, symmetry and balance. My student is will build number skills as the they sif t and make exciting number shell discoveries with every scoop at the sand tabl e. The sort a shape activity board will allow my youngest students to learn col ors, shapes and sorting skills as they fit various shape pieces into place.nann an

```
In [21]: #remove spacial character: https://stackoverflow.com/a/5843547/4084039
sent = re.sub('[^A-Za-z0-9]+', ' ', sent)
print(sent)
```

My preschool students are children who are three to five years of age My school is in sunny San Pedro California The children from San Pedro come to school eac h morning ready to learn and grow There is never a dull moment in our class my students are busy bees moving from one interest area to another They are eager to learn explore and experiment with the instructional materials and centers I set up for them We need more materials for the children to engage with material s that will foster their interest in technology literacy math science art and e ngineering My student is will learn number recognition and develop counting ski lls while engaging with the Learn to count picture puzzles and number Sequencin g puzzles While building with the 3 D Magnet Builders and Crystal Building Bloc ks my student is mathematical skills will be supported and strengthened in conc epts such as measurement comparison number estimation symmetry and balance My s tudent is will build number skills as the they sift and make exciting number sh ell discoveries with every scoop at the sand table The sort a shape activity bo ard will allow my youngest students to learn colors shapes and sorting skills a s they fit various shape pieces into place nannan

```
In [23]: # Combining all the above statemennts
    from tqdm import tqdm
    preprocessed_essays = []
    # tqdm is for printing the status bar
    for sentance in tqdm(X['essay'].values):
        sent = decontracted(sentance)
        sent = sent.replace('\\r', '')
        sent = sent.replace('\\", '')
        sent = sent.replace('\\", '')
        sent = re.sub('[^A-Za-z0-9]+', '', sent)
        # https://gist.github.com/sebleier/554280
        sent = ''.join(e for e in sent.split() if e not in stopwords)
        preprocessed_essays.append(sent.lower().strip())
```

00%|**| 100**|| 000 | 001:39<00:00, 500.76it/s|

```
In [24]: # after preprocesing

# X['essay'] = None
X['essay'] = preprocessed_essays
X.head(2)
```

# Out[24]:

Unnamed: id teacher\_id teacher\_prefix school\_state project\_sul

**0** 160221 p253737 c90749f5d961ff158d4b4d1e7dc665fc Mrs. IN

1 140945 p258326 897464ce9ddc600bced1151f324dd63a Mr. FL

```
In [25]: # Combining all the above statemennts
    from tqdm import tqdm
    preprocessed_project_title = []
    # tqdm is for printing the status bar
    for sentance in tqdm(X['project_title'].values):
        sent = decontracted(sentance)
        sent = sent.replace('\\r', ' ')
        sent = sent.replace('\\r', ' ')
        sent = sent.replace('\\r', ' ')
        sent = re.sub('[^A-Za-z0-9]+', ' ', sent)
        # https://gist.github.com/sebleier/554280
        sent = ' '.join(e for e in sent.split() if e not in stopwords)
        preprocessed_project_title.append(sent.lower().strip())
```

| 50000/50000 [00:04<00:00, 10262.14it/s]

100%

```
In [26]: | preprocessed_project_title[4999]
          # after preprocesing
          # X['project title'] = None
         X['project title'] = preprocessed project title
          X.head(2)
Out[26]:
             Unnamed:
                           id
                                                   teacher_id teacher_prefix school_state project_sul
          0
                                                                                  IN
               160221 p253737
                                c90749f5d961ff158d4b4d1e7dc665fc
                                                                    Mrs.
          1
                                                                                 FL
               140945 p258326 897464ce9ddc600bced1151f324dd63a
                                                                     Mr.
         2.2.8 Splitting the data into Train and Test
In [27]: # train test split(67:33)
          from sklearn.model selection import train test split
         X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.33, strati
          # X train, X cv, y train, y cv = train test split(X train, y train, test size=0..
In [28]: # count of all the words in corpus python: https://stackoverflow.com/a/22898595/4
          from collections import Counter
          my counter = Counter()
          for word in X train['clean categories'].values:
              my counter.update(word.split())
          print(my counter)
          # dict sort by value python: https://stackoverflow.com/a/613218/4084039
          cat dict = dict(my counter)
          sorted cat dict train = dict(sorted(cat dict.items(), key=lambda kv: kv[1]))
         Counter({'Literacy Language': 16076, 'Math Science': 12619, 'Health Sports': 43
         84, 'SpecialNeeds': 4165, 'AppliedLearning': 3769, 'Music_Arts': 3165, 'History
          Civics': 1785, 'Warmth': 410, 'Care Hunger': 410})
```

## 2.2.9 Vectorizing Categorical data: clean\_categories(Project subject categories)

```
In [29]: print(X train.shape, y train.shape)
         # print(X_cv.shape, y_cv.shape)
         print(X test.shape, y test.shape)
         print("="*100)
         Bow features names1=[]
         from sklearn.feature extraction.text import CountVectorizer
         vectorizer = CountVectorizer(vocabulary=list(sorted cat dict train.keys()), lower
         vectorizer.fit(X_train['clean_categories'].values) # fit has to happen only on the
         # we use the fitted Countvectorizer to convert the text to vector
         X train clean cat ohe = vectorizer.transform(X train['clean categories'].values)
         # X_cv_clean_cat_ohe = vectorizer.transform(X_cv['clean_categories'].values)
         X test clean cat ohe = vectorizer.transform(X test['clean categories'].values)
         print("After vectorizations")
         print(X train clean cat ohe.shape, y train.shape)
         # print(X cv clean cat ohe.shape, y cv.shape)
         print(X_test_clean_cat_ohe.shape, y_test.shape)
         print(vectorizer.get feature names())
         # print(vectorizer test.get feature names())
         print("="*100)
         (33500, 16) (33500,)
         (16500, 16) (16500,)
         ------
         After vectorizations
```

```
(33500, 9) (33500,)
(16500, 9) (16500,)
['Warmth', 'Care_Hunger', 'History_Civics', 'Music_Arts', 'AppliedLearning', 'S
pecialNeeds', 'Health_Sports', 'Math_Science', 'Literacy_Language']
 ______
```

# 2.2.10 Vectorizing Categorical data: clean\_subcategories(Project subject subcategories)

```
In [30]: # count of all the words in corpus python: https://stackoverflow.com/a/22898595/4
         from collections import Counter
         my counter = Counter()
         for word in X train['clean subcategories'].values:
             my_counter.update(word.split())
         sub cat dict = dict(my counter)
          sorted_sub_cat_dict_train = dict(sorted(sub_cat_dict.items(), key=lambda kv: kv[
```

```
In [31]: print(X train.shape, y train.shape)
         # print(X_cv.shape, y_cv.shape)
         print(X_test.shape, y_test.shape)
         print("="*100)
         vectorizer = CountVectorizer(vocabulary=list(sorted sub cat dict train.keys()),
         vectorizer.fit(X train['clean subcategories'].values) # fit has to happen only of
         # we use the fitted Countvectorizer to convert the text to vector
         X train clean sub cat ohe = vectorizer.transform(X train['clean subcategories'].
         # X_cv_clean_sub_cat_ohe = vectorizer.transform(X_cv['clean_subcategories'].value
         X_test_clean_sub_cat_ohe = vectorizer.transform(X_test['clean_subcategories'].val
         print("After vectorizations")
         print(X_train_clean_sub_cat_ohe.shape, y_train.shape)
         # print(X cv clean sub cat ohe.shape, y cv.shape)
         print(X_test_clean_sub_cat_ohe.shape, y_test.shape)
         print(vectorizer.get feature names())
         # print(vectorizer test.get feature names())
         print("="*100)
         (33500, 16) (33500,)
         (16500, 16) (16500,)
         After vectorizations
         (33500, 30) (33500,)
         (16500, 30) (16500,)
         ['Economics', 'CommunityService', 'FinancialLiteracy', 'ParentInvolvement', 'Ex
         tracurricular', 'Civics Government', 'ForeignLanguages', 'Warmth', 'Care Hunge
         r', 'NutritionEducation', 'SocialSciences', 'PerformingArts', 'CharacterEducati
         on', 'TeamSports', 'Other', 'College_CareerPrep', 'Music', 'History_Geography',
         'Health_LifeScience', 'EarlyDevelopment', 'Gym_Fitness', 'ESL', 'EnvironmentalS
        cience', 'VisualArts', 'Health_Wellness', 'AppliedSciences', 'SpecialNeeds', 'L
         iterature_Writing', 'Mathematics', 'Literacy']
         ______
```

-----

# 2.2.11 Vectorizing Categorical data: school\_state

```
In [33]: | print(X_train.shape, y_train.shape)
         # print(X_cv.shape, y_cv.shape)
         print(X test.shape, y test.shape)
         print("="*100)
         vectorizer = CountVectorizer(vocabulary=list(sorted school state dict train.keys
         vectorizer.fit(X train['school state'].values) # fit has to happen only on train
         # we use the fitted Countvectorizer to convert the text to vector
         X_train_School_state_ohe = vectorizer.transform(X_train['school_state'].values)
         # X_cv_School_state_ohe = vectorizer.transform(X_cv['school_state'].values)
         X test School state ohe = vectorizer.transform(X test['school state'].values)
         print("After vectorizations")
         print(X train School state ohe.shape, y train.shape)
         # print(X cv School state ohe.shape, y cv.shape)
         print(X_test_School_state_ohe.shape, y_test.shape)
         print(vectorizer.get feature names())
         # print(vectorizer_test.get_feature_names())
         print("="*100)
         (33500, 16) (33500,)
         (16500, 16) (16500,)
         After vectorizations
         (33500, 51) (33500,)
         (16500, 51) (16500,)
         ['VT', 'WY', 'ND', 'MT', 'RI', 'SD', 'NE', 'AK', 'NH', 'DE', 'ME', 'WV', 'NM',
          'DC', 'HI', 'ID', 'IA', 'KS', 'AR', 'CO', 'MN', 'KY', 'OR', 'MS', 'NV', 'MD',
              'WI', 'UT',
                                 'TN',
                           'AL',
                                      'VA', 'AZ', 'OK', 'NJ', 'MA', 'WA',
          'MO', 'IN', 'PA', 'MI', 'GA', 'SC', 'IL', 'NC', 'FL', 'NY', 'TX', 'CA']
```

#### 2.2.12 Vectorizing Categorical data: project\_grade\_category

```
In [34]: # count of all the words in corpus python: https://stackoverflow.com/a/22898595/4
from collections import Counter
my_counter = Counter()
for word in X_train['project_grade_category'].values:
    my_counter.update(word.split())
project_grade_dict = dict(my_counter)
sorted_project_grade_dict_train = dict(sorted(project_grade_dict.items(), key=lar
```

```
In [35]: print(X train.shape, y train.shape)
         # print(X_cv.shape, y_cv.shape)
         print(X_test.shape, y_test.shape)
         print("="*100)
         vectorizer= CountVectorizer(vocabulary=list(sorted project grade dict train.keys
         vectorizer.fit(X train['project grade category'].values) # fit has to happen only
         # we use the fitted Countvectorizer_pro_gradeto convert the text to vector
         X train project grade category ohe = vectorizer.transform(X train['project grade
         # X_cv_project_grade_category_ohe = vectorizer.transform(X_cv['project_grade_cate
         X_test_project_grade_category_ohe = vectorizer.transform(X_test['project_grade_c
         print("After vectorizations")
         print(X_train_project_grade_category_ohe.shape, y_train.shape)
         # print(X_cv_project_grade_category_ohe.shape, y_cv.shape)
         print(X_test_project_grade_category_ohe.shape, y_test.shape)
         print(vectorizer.get feature names())
         # print(vectorizer_test.get_feature_names())
         print("="*100)
         (33500, 16) (33500,)
         (16500, 16) (16500,)
         After vectorizations
         (33500, 4) (33500,)
         (16500, 4) (16500,)
         ['Grades9-12', 'Grades6-8', 'Grades3-5', 'GradesPreK-2']
             -----
```

#### 2.2.13 Vectorizing Categorical data: teacher\_prefix

```
In [36]: #To overcome the blanks in the teacher_prefix categry the .fillna is used
X_train['teacher_prefix']=X_train['teacher_prefix'].fillna("")
# project_data1=project_data.dropna()
```

```
In [37]: #To overcome the blanks in the teacher_prefix categry the .fillna is used
    X_test['teacher_prefix']=X_test['teacher_prefix'].fillna("")
    # project_data1=project_data.dropna()
```

```
In [38]: # count of all the words in corpus python: https://stackoverflow.com/a/22898595/4
         from collections import Counter
         my counter = Counter()
         my counter1=[]
         # project data['teacher prefix']=str(project data['teacher prefix'])
         for word in X_train['teacher_prefix'].values:
             my counter.update(word.split())
         teacher prefix dict = dict(my counter)
         sorted teacher prefix dict train = dict(sorted(teacher prefix dict.items(), key=
         # teacher_prefix_dict
In [39]:
         print(X_train.shape, y_train.shape)
         # print(X cv.shape, y cv.shape)
         print(X test.shape, y test.shape)
         print("="*100)
         vectorizer = CountVectorizer(vocabulary=list(sorted teacher prefix dict train.ke
         vectorizer.fit(X train['teacher prefix'].values) # fit has to happen only on tra
         # we use the fitted Countvectorizer to convert the text to vector
         X_train_teacher_prefix_ohe = vectorizer.transform(X_train['teacher_prefix'].value
         # X cv teacher prefix ohe = vectorizer.transform(X cv['teacher prefix'].values)
         X test teacher prefix ohe = vectorizer.transform(X test['teacher prefix'].values
         print("After vectorizations")
         print(X train teacher prefix ohe.shape, y train.shape)
         # print(X cv teacher prefix ohe.shape, y cv.shape)
         print(X_test_teacher_prefix_ohe.shape, y_test.shape)
         print(vectorizer.get feature names())
         # print(vectorizer test.get feature names())
         print("="*100)
```

2.3 Make Data Model Ready: Vectorizing Numerical features

# 2.3.1 Vectorizing Numerical features--Price

(33500, 16) (33500,)

```
In [40]: from sklearn.preprocessing import Normalizer
         from sklearn.preprocessing import StandardScaler
         normalizer = StandardScaler() #Normalizer()
         # normalizer test = Normalizer()
         # normalizer.fit(X_train['price'].values)
         # this will rise an error Expected 2D array, got 1D array instead:
         # array=[105.22 215.96 96.01 ... 368.98 80.53 709.67].
         # Reshape your data either using
         # array.reshape(-1, 1) if your data has a single feature
         # array.reshape(1, -1) if it contains a single sample.
         normalizer.fit(X train['price'].values.reshape(-1,1))
         # normalizer_test.fit(X_test['price'].values.reshape(1,-1))
         X train price norm = normalizer.transform(X train['price'].values.reshape(-1,1))
         # X cv price norm = normalizer.transform(X cv['price'].values.reshape(-1,1))
         X_test_price_norm = normalizer.transform(X_test['price'].values.reshape(-1,1))
         # X_train_price_norm=np.reshape(X_train_price_norm,(1,-1))
         # X test price norm=np.reshape(X test price norm,(1,-1))
         print("After vectorizations")
         # np.reshape(X_train_price_norm,
         print(X train price norm.shape, y train.shape)
         # print(X cv price norm.shape, y cv.shape)
         print(X test price norm.shape, y test.shape)
         print("="*100)
         After vectorizations
         (33500, 1) (33500,)
         (16500, 1) (16500,)
         _______
```

2.3.2 Vectorizing Numerical features-teacher\_number\_of\_previously\_posted\_projects

```
In [41]: | from sklearn.preprocessing import Normalizer
                        normalizer train = StandardScaler() #Normalizer()
                       normalizer test = StandardScaler() #Normalizer()
                       # normalizer.fit(X train['teacher number of previously posted projects'].values)
                       # this will rise an error Expected 2D array, got 1D array instead:
                       # array=[105.22 215.96 96.01 ... 368.98 80.53 709.67].
                       # Reshape your data either using
                       # array.reshape(-1, 1) if your data has a single feature
                       # array.reshape(1, -1) if it contains a single sample.
                       normalizer.fit(X_train['teacher_number_of_previously_posted_projects'].values.re
                       X_train_teacher_number_of_previously_posted_projects_norm = normalizer.transform
                       # X_cv_teacher_number_of_previously_posted_projects_norm = normalizer.transform()
                       X test teacher number of previously posted projects norm = normalizer.transform()
                       # X_train_teacher_number_of_previously_posted_projects_norm=np.reshape(X_train_t&
                       # X_test_teacher_number_of_previously_posted_projects_norm=np.reshape(X_test_teacher_number_of_previously_posted_projects_norm=np.reshape(X_test_teacher_number_of_previously_posted_projects_norm=np.reshape(X_test_teacher_number_of_previously_posted_projects_norm=np.reshape(X_test_teacher_number_of_previously_posted_projects_norm=np.reshape(X_test_teacher_number_of_previously_posted_projects_norm=np.reshape(X_test_teacher_number_of_previously_posted_projects_norm=np.reshape(X_test_teacher_number_of_previously_posted_projects_norm=np.reshape(X_test_teacher_number_of_previously_posted_projects_norm=np.reshape(X_test_teacher_number_of_previously_posted_projects_norm=np.reshape(X_test_teacher_number_of_previously_posted_projects_norm=np.reshape(X_test_teacher_number_of_previously_posted_projects_norm=np.reshape(X_test_teacher_number_of_previously_posted_projects_number_of_previously_posted_projects_number_of_previously_posted_projects_number_of_previously_posted_projects_number_of_previously_posted_projects_number_of_previously_posted_projects_number_of_previously_posted_projects_number_of_previously_posted_projects_number_of_previously_posted_projects_number_of_previously_posted_projects_number_of_previously_posted_projects_number_of_previously_posted_projects_number_of_previously_posted_projects_number_of_previously_posted_projects_number_of_previously_posted_projects_number_of_previously_posted_projects_number_of_previously_posted_projects_number_of_previously_posted_projects_number_of_previously_posted_projects_number_of_previously_posted_projects_number_of_previously_posted_projects_number_of_previously_posted_projects_number_of_previously_posted_projects_number_of_previously_posted_projects_number_of_previously_posted_projects_number_of_previously_posted_projects_number_of_previously_posted_projects_number_of_previously_projects_number_of_previously_projects_number_of_previously_projects_number_of_previously_projects_number_of_previously_projects_number_of_previously_projects_number_of_previously_projects_numb
                        print("After vectorizations")
                       print(X train teacher number of previously posted projects norm.shape, y train.sl
                       # print(X_cv_teacher_number_of_previously_posted_projects_norm.shape, y_cv.shape)
                        print(X_test_teacher_number_of_previously_posted_projects_norm.shape, y_test.sha
                        print("="*100)
                       After vectorizations
                       (33500, 1) (33500,)
                       (16500, 1) (16500,)
```

# 2.3.3 Vectorizing Numerical features--digits\_in\_summary

```
In [42]: X_train['digits_in_summary'].fillna(X_train['digits_in_summary'].mean(), inplace:
# X_cv['digits_in_summary'].fillna(X_cv['digits_in_summary'].mean(), inplace=True
X_test['digits_in_summary'].fillna(X_test['digits_in_summary'].mean(), inplace=True
```

```
In [43]: | from sklearn.preprocessing import Normalizer
         normalizer train = StandardScaler() #Normalizer()
         normalizer test = StandardScaler() #Normalizer()
         # normalizer.fit(X train['digits in summary'].values)
         # this will rise an error Expected 2D array, got 1D array instead:
         # array=[105.22 215.96 96.01 ... 368.98 80.53 709.67].
         # Reshape your data either using
         # array.reshape(-1, 1) if your data has a single feature
         # array.reshape(1, -1) if it contains a single sample.
         normalizer.fit(X_train['digits_in_summary'].values.reshape(-1,1))
         X_train_digits_in_summary_norm = normalizer.transform(X_train['digits_in_summary
         # X_cv_digits_in_summary_norm = normalizer.transform(X_cv['digits_in_summary'].ve
         X_test_digits_in_summary_norm = normalizer.transform(X_test['digits_in_summary']
         # X train digits in summary norm=np.reshape(X train digits in summary norm,(1,-1)
         # X test digits in summary norm=np.reshape(X test digits in summary norm,(1,-1))
         print("After vectorizations")
         print(X train digits in summary norm.shape, y train.shape)
         # print(X_cv_digits_in_summary_norm.shape, y_cv.shape)
         print(X test digits in summary norm.shape, y test.shape)
         print("="*100)
         After vectorizations
```

2.6 Make Data Model Ready: Vectorizing Essay and Project\_title feature into BOW & TFIDF

**Vectorizing Text data** 

- 2.6 TFIDF vectorizer
- 2.6.1 TFIDF vectorizer: Essay

```
In [44]: print(X train.shape, y train.shape)
        # print(X_cv.shape, y_cv.shape)
        print(X_test.shape, y_test.shape)
        print("="*100)
        from sklearn.feature extraction.text import TfidfVectorizer
        vectorizer = TfidfVectorizer(min_df=10,ngram_range=(2,2), max_features=5000)
        vectorizer.fit(X_train['essay'].values) # fit has to happen only on train data
        # we use the fitted Countvectorizer to convert the text to vector
        X train essay Tfidf = vectorizer.transform(X train['essay'].values)
        # X_cv_essay_Tfidf = vectorizer.transform(X_cv['essay'].values)
        X test essay Tfidf = vectorizer.transform(X test['essay'].values)
        print("After vectorizations")
        print(X_train_essay_Tfidf.shape, y_train.shape)
        # print(X cv essay Tfidf.shape, y cv.shape)
        print(X test essay Tfidf.shape, y test.shape)
        print("="*100)
        (33500, 16) (33500,)
        (16500, 16) (16500,)
        _______
        After vectorizations
        (33500, 5000) (33500,)
        (16500, 5000) (16500,)
        ______
```

#### 2.6.2 TFIDF vectorizer:Project Title

```
In [45]: # print(X train.shape, y train.shape)
         # print(X cv.shape, y cv.shape)
         # print(X test.shape, y test.shape)
         print("="*100)
         # We are considering only the words which appeared in at least 10 documents(rows
         vectorizer = TfidfVectorizer(min df=10,ngram range=(2,2), max features=5000)
         vectorizer.fit(X train['project title'].values) # fit has to happen only on trail
         # we use the fitted Countvectorizer to convert the text to vector
         X_train_project_title_tfidf = vectorizer.transform(X_train['project_title'].value
         # X_cv_project_title_tfidf = vectorizer.transform(X_cv['project_title'].values)
         X test project title tfidf = vectorizer.transform(X test['project title'].values
         print("After vectorizations")
         print(X train project title tfidf.shape, y train.shape)
         # print(X_cv_project_title_tfidf.shape, y_cv.shape)
         print(X_test_project_title_tfidf.shape, y_test.shape)
         print("="*100)
         ______
         After vectorizations
         (33500, 1054) (33500,)
         (16500, 1054) (16500,)
In [46]: from scipy.sparse import hstack
         X Tfidf train = hstack(( X train project title tfidf,X train essay Tfidf,X train
         X_Tfidf_train=X_Tfidf_train.todense()
         X Tfidf train=np.array(X Tfidf train)
         # X_Tfidf_cv = hstack(( X_cv_project_title_tfidf,X_cv_essay_Tfidf,X_cv_digits_in
         # X_Tfidf_cv=X_Tfidf_cv.todense()
         # X_Tfidf_cv=np.array(X_Tfidf_cv)
         X Tfidf test = hstack(( X test project title tfidf,X test essay Tfidf,X test dig
         X Tfidf test=X Tfidf test.todense()
         X Tfidf test=np.array(X Tfidf test)
```

#### 2.6.3 Applying Decision Tree on TFIDF, SET 1

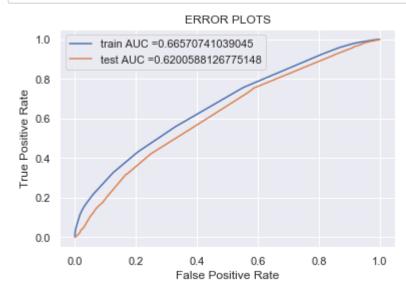
Applying Decision Tree & GridSearchCV on Train data to obtain the best C

```
In [47]: from sklearn.metrics import roc auc score
         import matplotlib.pyplot as plt
         from sklearn.model selection import train test split
         from sklearn.model selection import GridSearchCV
         from sklearn.model selection import cross val score
         # from sklearn.multioutput import MultiOutputClassifier
         # from sklearn.datasets import make multilabel classification
         from sklearn.tree import DecisionTreeClassifier
         dt1 = DecisionTreeClassifier(class weight = 'balanced')
         parameters = {'max_depth': [1, 5, 10, 50], 'min_samples_split': [5, 10, 100,500]]
         clf1 = GridSearchCV(dt1, parameters, cv=2, scoring='roc_auc',return_train_score='
         se1 = clf1.fit(X Tfidf train, y train)
         # train_auc = model.cv_results_['mean_train_score']
         # cv auc= model.cv results ['mean test score']
         # plt.plot(parameters['alpha'], train_auc, label='Train AUC')
         # plt.plot(parameters['alpha'], cv_auc, label='CV AUC')
         # plt.scatter(parameters['alpha'], train_auc, label='Train AUC points')
         # plt.scatter(parameters['alpha'], cv_auc, label='CV AUC points')
         # plt.legend()
         # plt.xlabel("log")
         # plt.xscale('log')
         # plt.ylabel("ROC AUC score")
         # plt.title("ROC AUC vs log plot")
         # plt.grid()
         # plt.show()
         # model.fit(X_Tfidf_train, y_train)
         # print(model.best estimator )
         # print(model.score(X Tfidf test, y test))
```

```
In [48]:
           import seaborn as sns; sns.set()
           max scores1 = pd.DataFrame(clf1.cv results ).groupby(['param max depth','param m'
           fig, ax = plt.subplots(1,2, figsize=(20,6))
           sns.heatmap(max_scores1.mean_train_score, annot = True, fmt='.4g', ax=ax[0])
           sns.heatmap(max_scores1.mean_test_score, annot = True, fmt='.4g', ax=ax[1])
           ax[0].set title('Train Set')
           ax[1].set title('CV Set')
           plt.show()
                                                                                                     - 0.60
                                                                                   0.5434
                         0.5451
                                 0.5451
                 0.5451
                                          0.5451
                                                   - 0.88
                                                                   0.5434
                                                                           0.5434
                                                                                            0.5434
                                                                                                     - 0.59
           _max_depth
5
                                                                                            0.5915
                 0.6293
                                 0.6269
                                          0.6254
                                                                   0.5922
                                                                           0.5917
                                                                                   0.5912
                                                                                                     - 0.57
                                          0.6783
                                                                   0.5989
                                                                           0.5981
                                                                                   0.5962
                                                                                            0.6012
                                                                                                     - 0.56
                                                    0.64
                                                                   0.5614
                                                                                   0.5645
                 0.9281
                         0.9135
                                 0.8682
                                                                           10 100 param_min_samples_split
                                                                                             500
In [50]:
           print(clf1.best estimator )
           #Mean cross-validated score of the best estimator
           print(clf1.score(X Tfidf train,y train))
           print(clf1.score(X Tfidf test,y test))
           DecisionTreeClassifier(ccp_alpha=0.0, class_weight='balanced', criterion='gin
           i',
                                      max depth=10, max features=None, max leaf nodes=None,
                                      min impurity decrease=0.0, min impurity split=None,
                                      min_samples_leaf=1, min_samples_split=500,
                                      min_weight_fraction_leaf=0.0, presort='deprecated',
                                      random state=None, splitter='best')
           0.66570741039045
           0.6200588126775148
          best_tune_parameters=[{'max_depth':[10], 'min_samples_split':[500] } ]
In [51]:
```

#### 2.6.4 Receiver Operating Characteristic- (TFIDF)

```
In [65]:
         # https://scikitlearn.org/stable/modules/generated/sklearn.metrics.roc curve.html
         from sklearn.metrics import roc curve, auc
         from sklearn.tree import DecisionTreeClassifier
         from sklearn.metrics import roc curve, auc
         clf11= DecisionTreeClassifier(class weight = 'balanced',max depth=10,min samples
         # clfV1=DecisionTreeClassifier (class_weight = 'balanced',max_depth=10,min_sample
         clf11.fit(X Tfidf train, y train)
         # for visulation
         # clfV1.fit(X_Tfidf_train, y_train)
         #https://scikitlearn.org/stable/modules/generated/sklearn.linear_model.SGDClassij
         y train pred1 = clf11.predict proba(X Tfidf train) [:,1]
         y_test_pred1 = clf11.predict_proba(X_Tfidf_test) [:,1]
         train_fpr1, train_tpr1, tr_thresholds1 = roc_curve(y_train, y_train_pred1)
         test fpr1, test tpr1, te thresholds1 = roc curve(y test, y test pred1)
         plt.plot(train fpr1, train tpr1, label="train AUC ="+str(auc(train fpr1, train t
         plt.plot(test_fpr1, test_tpr1, label="test AUC ="+str(auc(test_fpr1, test_tpr1)))
         plt.legend()
         plt.xlabel("False Positive Rate")
         plt.ylabel("True Positive Rate")
         plt.title("ERROR PLOTS")
         plt.grid(True)
         plt.show()
```



#### 2.6.5 Confusion matrix- TFIDF

```
In [66]: # https://stackoverflow.com/questions/35572000/how-can-i-plot-a-confusion-matrix
#function to get heatmap confusion matrix
def get_confusion_matrix(clf,X_te,y_test):
    y_pred = clf.predict(X_te)
    df_cm = pd.DataFrame(confusion_matrix(y_test, y_pred), index =['Actual NO','/
    sns.heatmap(df_cm, annot=True,annot_kws={"size": 16}, fmt='g')

# %%time
get_confusion_matrix(clf11,X_Tfidf_test,y_test)
```

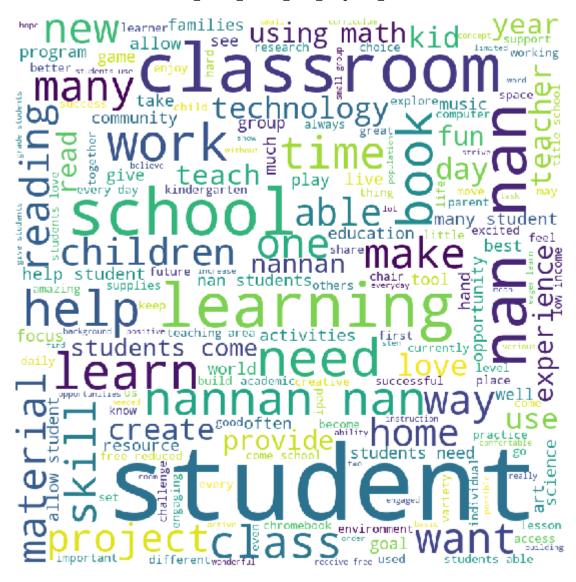


2.7.1 Selecting the FPR data points from Essay, Price and Teacher\_number\_of\_previously\_posted\_projects test data set.

```
In [68]:
         import numpy as np
         import pandas as pd
         y pred = clf11.predict(X Tfidf test)
         #Essay
         df=X_test['essay']
         df.to frame()
         #Price
         df =X test['price']
         df_.to_frame()
         #Teacher number of previously posted projects
         df_1=X_test['teacher_number_of_previously_posted_projects']
         df_1.to_frame()
         df1=pd.DataFrame(y test,columns=['y test'])
         df2=pd.DataFrame(y_pred,columns=['y_pred'])
         df_row = pd.concat([df,df1, df2],axis=1)
         df_row1 = pd.concat([df_,df1, df2],axis=1)
         df row2 = pd.concat([df 1,df1, df2],axis=1)
         #Selecting the false positive data points
         df row = df row[(df row.y pred == 1) & (df row.y test == 0) ]
         df_row1 = df_row1[(df_row1.y_pred == 1) & (df_row1.y_test == 0) ]
         df_row2 = df_row2[(df_row2.y_pred == 1) & (df_row2.y_test == 0) ]
         df row2=df row2.fillna(0)
```

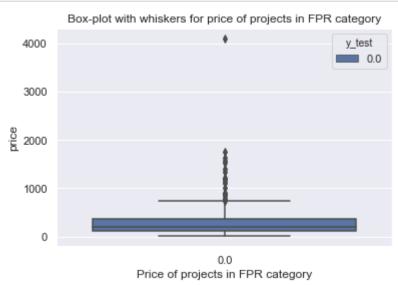
#### 2.7.2 WordCloud of False positive data points- (TFIDF)

```
In [69]: # https://www.geeksforgeeks.org/generating-word-cloud-python/
         from wordcloud import WordCloud, STOPWORDS
         import matplotlib.pyplot as plt
         import pandas as pd
         # Reads 'Youtube04-Eminem.csv' file
         # df = pd.read csv(r"Youtube04-Eminem.csv", encoding ="latin-1")
         comment words = ' '
         stopwords = set(STOPWORDS)
         # iterate through the csv file
         for val in df row['essay']:
             # typecaste each val to string
             val = str(val)
             # split the value
             tokens = val.split()
             # Converts each token into Lowercase
             for i in range(len(tokens)):
                  tokens[i] = tokens[i].lower()
             for words in tokens:
                  comment words = comment words + words + ' '
         wordcloud = WordCloud(width = 800, height = 800,
                          background_color ='white',
                          stopwords = stopwords,
                          min font size = 10).generate(comment words)
         # plot the WordCloud image
         plt.figure(figsize = (8, 8), facecolor = None)
         plt.imshow(wordcloud)
         plt.axis("off")
         plt.tight layout(pad = 0)
         plt.show()
```



# 2.7.3 Box plot of price of False positive data points

# In [70]: import seaborn as sns import matplotlib.pyplot as plt g=sns.boxplot(x='y\_test',y='price',hue='y\_test',data=df\_row1) plt.title("Box-plot with whiskers for price of projects in FPR category") plt.xlabel("Price of projects in FPR category") # plt.legend(loc=1) plt.show()

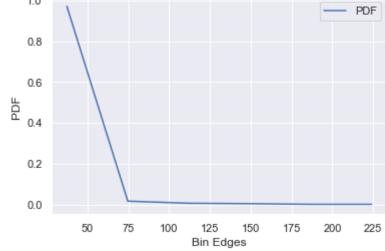


# 2.7.4 PDF of teacher\_number\_of\_previously\_posted\_projects for false positive data points.

```
The pdf is : [0.97068063 0.01675393 0.00628272 0.00418848 0.00104712 0.0010471 2]
The bin edges of pdf are: [ 0. 37.33333333 74.66666667 112. 149.33333333 186.66666667 224. ]
```

# Out[71]: <matplotlib.legend.Legend at 0x3229cb70>





```
In [72]: # print("PDF1 of Patients who survived falling in >=5years from the graph-Blue to
print("The probability of finding project quote less than ",round(bin_edges[1],2)
print("The probability of finding project quote less than ",round(bin_edges[2],2)
print("The probability of finding project quote less than ",round(bin_edges[3],2)
print("The probability of finding project quote less than ",round(bin_edges[4],2)
print("The probability of finding project quote less than ",round(bin_edges[5],2)
print("The probability of finding project quote less than ",round(bin_edges[6],2)
```

The probability of finding project quote less than 37.33 that got approved and falling in FPR category is 97.07 %

The probability of finding project quote less than 74.67 that got approved and falling in FPR category is 1.68 %

The probability of finding project quote less than  $\,$  112.0 that got approved and falling in FPR category is 0.63 %

The probability of finding project quote less than  $\,$  149.33 that got approved an d falling in FPR category is 0.42 %

The probability of finding project quote less than 186.67 that got approved an d falling in FPR category is 0.1 %

The probability of finding project quote less than  $\,$  224.0 that got approved and falling in FPR category is 0.1 %

## 2.8.1 Getting top features using feature importances from TFIDF model

```
In [76]: from sklearn.feature_selection import SelectFromModel
    from sklearn.linear_model import LogisticRegression

    selector = SelectFromModel(estimator=DecisionTreeClassifier(max_depth=None,class)
# selector.estimator_.coef_
    selector.threshold_
    selector.get_support()

dut[76]: array([False, False, False, ..., True, True, True])

In [77]: x=selector.transform(X_Tfidf_train)
    x1=selector.transform(X_Tfidf_test)
# X_Tfidf_train.shape
# X_Tfidf_test.shape
```

# 2.8.2 Applying Decision Tree & GridSearchCV on Train data with best selected features to obtain the best max depth, min samples split

```
In [61]:
          from sklearn.metrics import roc auc score
          import matplotlib.pyplot as plt
          from sklearn.model selection import train test split
          from sklearn.model selection import GridSearchCV
          from sklearn.model selection import cross val score
          from sklearn.tree import DecisionTreeClassifier
          dt4= DecisionTreeClassifier(class weight = 'balanced')
          parameters = {'max depth': [1, 5, 10, 50],'min samples split': [5, 10, 100,500]}
          clf4 = GridSearchCV(dt4, parameters, cv=5, scoring='roc auc',return train score='
          set4= clf4.fit(x, y_train)
In [62]:
          import seaborn as sns; sns.set()
          max_scores1 = pd.DataFrame(clf4.cv_results_).groupby(['param_max_depth','param_m']
          fig, ax = plt.subplots(1,2, figsize=(20,6))
          sns.heatmap(max_scores1.mean_train_score, annot = True, fmt='.4g', ax=ax[0])
          sns.heatmap(max scores1.mean test score, annot = True, fmt='.4g', ax=ax[1])
          ax[0].set title('Train Set')
          ax[1].set_title('CV Set')
          plt.show()
                           Train Set
                0.5444
                        0.5444
                               0.5444
                                        0.5444
                                                               0.5422
                                                                       0.5422
                                                                               0.5422
                                                                                       0.5422
                                                                                                - 0.600
                                                               0.5971
                0.6256
                        0.6255
                                                                       0.5971
                                                                               0.5973
                                                                                       0.5979
                                0.624
                                        0.6228
                                                                                                - 0.585
           max
                                                 0.72
                                                                                                0.570
                                0.6809
                                                               0.6078
                                                                       0.6072
                                                                               0.6079
                                                                                       0.6126
                                                                                                0.555
                                                               0.5722
                0.9397
                        0.9262
                                0.8564
                                                                       0.5706
                                                                                       0.5987
                                        500
In [63]:
          #Best Estimator and Best tune parameters
          print(clf4.best estimator )
          #Mean cross-validated score of the best_estimator
          print(clf4.score(x,y train))
          print(clf4.score(x1,y_test))
          DecisionTreeClassifier(ccp_alpha=0.0, class_weight='balanced', criterion='gin
          i',
                                    max depth=10, max features=None, max leaf nodes=None,
                                    min impurity decrease=0.0, min impurity split=None,
                                    min_samples_leaf=1, min_samples_split=500,
                                    min weight fraction leaf=0.0, presort='deprecated',
                                    random state=None, splitter='best')
          0.66570741039045
          0.6200588126775148
```

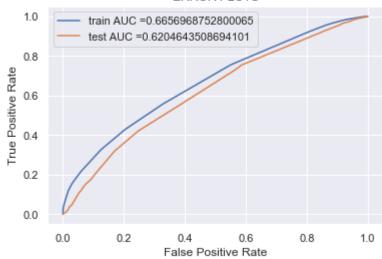
best\_tune\_parameters=[{'max\_depth':[10], 'min\_samples\_split':[500] } ]

## 2.8.3 Receiver Operating Characteristic- (TFIDF)

#### In [81]:

```
#Fitting Model to Hyper-Parameter Curve
from sklearn.metrics import roc curve, auc
from sklearn.tree import DecisionTreeClassifier
from sklearn.metrics import roc curve, auc
clf11= DecisionTreeClassifier(class weight = 'balanced',max depth=10,min samples
# clfV1=DecisionTreeClassifier (class weight = 'balanced', max depth=5, min sample
clf11.fit(x, y train)
# for visulation
# clfV1.fit(X_tfidf_w2v_train, y_train)
#https://scikitlearn.org/stable/modules/generated/sklearn.linear_model.SGDClassi
y train pred1 = clf11.predict proba(x) [:,1]
y test pred1 = clf11.predict proba(x1) [:,1]
train fpr1, train tpr1, tr thresholds1 = roc curve(y train, y train pred1)
test_fpr1, test_tpr1, te_thresholds1 = roc_curve(y_test, y_test_pred1)
plt.plot(train fpr1, train tpr1, label="train AUC ="+str(auc(train fpr1, train t
plt.plot(test_fpr1, test_tpr1, label="test AUC ="+str(auc(test_fpr1, test_tpr1))
plt.legend()
plt.xlabel("False Positive Rate")
plt.ylabel("True Positive Rate")
plt.title("ERROR PLOTS")
plt.grid(True)
plt.show()
```

#### ERROR PLOTS



```
In [82]: # https://stackoverflow.com/questions/35572000/how-can-i-plot-a-confusion-matrix
#function to get heatmap confusion matrix
def get_confusion_matrix(clf,X_te,y_test):
    y_pred = clf.predict(X_te)
    df_cm = pd.DataFrame(confusion_matrix(y_test, y_pred), index =['Actual NO','/s sns.heatmap(df_cm, annot=True,annot_kws={"size": 16}, fmt='g')

# %%time
get_confusion_matrix(clf11,x1,y_test)
```



```
In [84]: #To make best use of the memory we are setting the variable names to 'None' and p
X_Tfidf_test=None
X_Tfidf_train=None
gc.collect()
```

Out[84]: 0

# 2.9 TFIDF weighted W2V-Essay

# 2.9.1 Using Pretrained Models: TFIDF weighted W2V-Essay

```
In [85]: # train test split
         from sklearn.model selection import train test split
         Xtfidf w2v vectors train , Xtfidf w2v vectors test , y train, y test = train test
         # Xtfidf_w2v_vectors_train, Xtfidf_w2v_vectors_cv, y_train, y_cv = train_test_spl
In [86]: # S = ["abc def pqr", "def def def abc", "pqr pqr def"]
         tfidf model = TfidfVectorizer()
         tfidf model.fit(Xtfidf w2v vectors train )
         # we are converting a dictionary with word as a key, and the idf as a value
         dictionary = dict(zip(tfidf_model.get_feature_names(), list(tfidf_model.idf_)))
         tfidf words = set(tfidf model.get feature names())
In [87]: | # stronging variables into pickle files python: http://www.jessicayung.com/how-to
         # make sure you have the glove vectors file
         import pickle
         with open('glove_vectors', 'rb') as f:
             model = pickle.load(f)
             glove words = set(model .keys())
In [88]: # average Word2Vec
         # compute average word2vec for each review.
         tfidf_w2v_vectors_essay_train = []; # the avg-w2v for each sentence/review is stell
         for sentence in tqdm(Xtfidf_w2v_vectors_train_): # for each review/sentence
             vector = np.zeros(300) # as word vectors are of zero length
             tf idf weight =0; # num of words with a valid vector in the sentence/review
             for word in sentence.split(): # for each word in a review/sentence
                 if (word in glove words) and (word in tfidf words):
                     vec = model [word] # getting the vector for each word
                     # here we are multiplying idf value(dictionary[word]) and the tf value
                     tf idf = dictionary[word]*(sentence.count(word)/len(sentence.split())
                     vector += (vec * tf idf) # calculating tfidf weighted w2v
                     tf idf weight += tf idf
             if tf idf weight != 0:
                 vector /= tf idf weight
             tfidf_w2v_vectors_essay_train.append(vector)
         print(len(tfidf w2v vectors essay train))
         print(len(tfidf w2v vectors essay train[0]))
         100%
                                         33500/33500 [05:20<00:00, 104.46it/s]
         33500
         300
```

localhost:8888/notebooks/Donors\_Choose\_Decision\_Trees\_Assignment\_7.ipynb#

```
In [89]: # average Word2Vec
         # compute average word2vec for each review.
         tfidf w2v vectors essay test = []; # the avg-w2v for each sentence/review is stol
         for sentence in tqdm(Xtfidf_w2v_vectors_test_): # for each review/sentence
             vector = np.zeros(300) # as word vectors are of zero length
             tf idf weight =0; # num of words with a valid vector in the sentence/review
             for word in sentence.split(): # for each word in a review/sentence
                 if (word in glove words) and (word in tfidf words):
                     vec = model_[word] # getting the vector for each word
                     # here we are multiplying idf value(dictionary[word]) and the tf value
                     tf_idf = dictionary[word]*(sentence.count(word)/len(sentence.split())
                     vector += (vec * tf idf) # calculating tfidf weighted w2v
                     tf idf weight += tf idf
             if tf idf weight != 0:
                 vector /= tf idf weight
             tfidf w2v vectors essay test.append(vector)
         print(len(tfidf w2v vectors essay test))
         print(len(tfidf w2v vectors essay test[0]))
         100%
                                                  | 16500/16500 [02:57<00:00, 92.76it/s]
         16500
         300
```

In [90]: import scipy
 tfidf\_w2v\_vectors\_essay\_train=scipy.sparse.csr\_matrix(tfidf\_w2v\_vectors\_essay\_train)
 tfidf\_w2v\_vectors\_essay\_train)

tfidf\_w2v\_vectors\_essay\_test=scipy.sparse.csr\_matrix(tfidf\_w2v\_vectors\_essay\_test
type(tfidf\_w2v\_vectors\_essay\_test)

# Xtfidf\_w2v\_vectors\_cv=scipy.sparse.csr\_matrix(Xtfidf\_w2v\_vectors\_cv)
# type(Xtfidf\_w2v\_vectors\_cv)

Out[90]: scipy.sparse.csr.csr matrix

#### 2.9.2 Using Pretrained Models: TFIDF weighted W2V on project\_title

```
In [91]: # train test split
from sklearn.model_selection import train_test_split

tfidf_w2v_vectors_Pro_title_train_, tfidf_w2v_vectors_Pro_title_test_, y_train, y_d

# tfidf_w2v_vectors_Pro_title_train, tfidf_w2v_vectors_Pro_title_cv, y_train, y_d
```

```
In [92]: # S = ["abc def pqr", "def def def abc", "pqr pqr def"]
         tfidf model = TfidfVectorizer()
         tfidf model.fit(tfidf w2v vectors Pro title train )
         # 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 [93]: # average Word2Vec
         # compute average word2vec for each review.
         tfidf w2v vectors Pro title train = []; # the avg-w2v for each sentence/review is
         for sentence in tqdm(tfidf w2v vectors Pro title train ): # for each review/sente
             vector = np.zeros(300) # as word vectors are of zero length
             tf idf weight =0; # num of words with a valid vector in the sentence/review
             for word in sentence.split(): # for each word in a review/sentence
                 if (word in glove words) and (word in tfidf words):
                     vec = model [word] # getting the vector for each word
                     # here we are multiplying idf value(dictionary[word]) and the tf value
                     tf idf = dictionary[word]*(sentence.count(word)/len(sentence.split())
                     vector += (vec * tf_idf) # calculating tfidf weighted w2v
                     tf idf weight += tf idf
             if tf_idf_weight != 0:
                 vector /= tf idf weight
             tfidf w2v vectors Pro title train.append(vector)
         print(len(tfidf_w2v_vectors_Pro_title_train))
         print(len(tfidf_w2v_vectors_Pro_title_train[0]))
                                         | 33500/33500 [00:07<00:00, 4204.60it/s]
         100%
         33500
         300
In [94]: | # average Word2Vec
         # compute average word2vec for each review.
         tfidf w2v vectors Pro title test = []; # the avg-w2v for each sentence/review is
         for sentence in tqdm(tfidf_w2v_vectors_Pro_title_test_): # for each review/sentel
             vector = np.zeros(300) # as word vectors are of zero length
             tf_idf_weight =0; # num of words with a valid vector in the sentence/review
             for word in sentence.split(): # for each word in a review/sentence
                 if (word in glove words) and (word in tfidf words):
                     vec = model_[word] # getting the vector for each word
                     # here we are multiplying idf value(dictionary[word]) and the tf value
                     tf_idf = dictionary[word]*(sentence.count(word)/len(sentence.split())
                     vector += (vec * tf_idf) # calculating tfidf weighted w2v
                     tf idf weight += tf idf
             if tf idf weight != 0:
                 vector /= tf idf weight
             tfidf w2v vectors Pro title test.append(vector)
         print(len(tfidf_w2v_vectors_Pro_title_test))
         print(len(tfidf w2v vectors Pro title test[0]))
         100%
                                           16500/16500 [00:03<00:00, 4892.35it/s]
         16500
         300
```

```
In [95]:
         import scipy
         tfidf w2v vectors Pro title train=scipy.sparse.csr matrix(tfidf w2v vectors Pro
         type(tfidf w2v vectors Pro title train)
         tfidf w2v vectors Pro title test=scipy.sparse.csr matrix(tfidf w2v vectors Pro t
         type(tfidf_w2v_vectors_Pro_title_test)
         # tfidf_w2v_vectors_Pro_title_cv=scipy.sparse.csr_matrix(tfidf w2v vectors Pro t
         # type(tfidf w2v vectors Pro title cv)
Out[95]: scipy.sparse.csr.csr matrix
In [96]: X_tfidf_w2v_train = hstack((tfidf_w2v_vectors_essay_train,tfidf_w2v_vectors_Pro_
         X tfidf w2v train=X tfidf w2v train.todense()
         X tfidf w2v train=np.array(X tfidf w2v train)
         # X_tfidf_w2v_cv = hstack((Xtfidf_w2v_vectors_cv,tfidf_w2v_vectors_Pro_title_cv
         # X tfidf w2v cv=X tfidf w2v cv.todense()
         # X tfidf w2v cv=np.array(X tfidf w2v cv)
         X tfidf w2v test = hstack((tfidf w2v vectors essay test,tfidf w2v vectors Pro ti
```

# X\_All = hstack((categories\_one\_hot,sub\_categories\_one\_hot,school\_state\_one\_hot,

#### 2.9.3 Applying Decision Tree on TFIDF W2V, SET 2

X\_tfidf\_w2v\_test=X\_tfidf\_w2v\_test.todense()
X tfidf w2v test=np.array(X tfidf w2v test)

# Applying Decision Tree & GridSearchCV on Train data to obtain the best max\_depth, min\_samples\_split

```
In [97]: from sklearn.metrics import roc_auc_score
    import matplotlib.pyplot as plt
    from sklearn.model_selection import train_test_split
    from sklearn.model_selection import GridSearchCV
    from sklearn.model_selection import cross_val_score
    from sklearn.tree import DecisionTreeClassifier
    dt4= DecisionTreeClassifier(class_weight = 'balanced')
    parameters = {'max_depth': [1, 5, 10, 50], 'min_samples_split': [5, 10, 100,500]]
    clf4 = GridSearchCV(dt4, parameters, cv=2, scoring='roc_auc',return_train_score='set4= clf4.fit(X_tfidf_w2v_train, y_train)
```

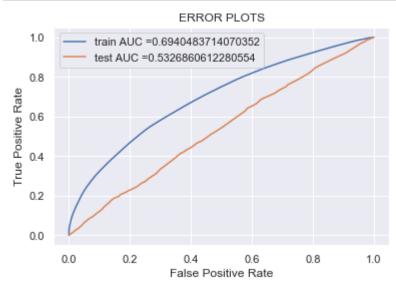
```
In [98]:
            import seaborn as sns; sns.set()
            max_scores1 = pd.DataFrame(clf4.cv_results_).groupby(['param_max_depth','param_m']
            fig, ax = plt.subplots(1,2, figsize=(20,6))
             sns.heatmap(max scores1.mean train score, annot = True, fmt='.4g', ax=ax[0])
            sns.heatmap(max_scores1.mean_test_score, annot = True, fmt='.4g', ax=ax[1])
             ax[0].set_title('Train Set')
             ax[1].set_title('CV Set')
            plt.show()
                                 Train Set
                   0.5414
                             0.5414
                                      0.5414
                                                0.5414
                                                                                                                   - 0.525
                   0.6278
                             0.6278
                                      0.6257
                                                                            0.5281
                                                                                      0.5282
                                                                                               0.5286
                                                                                                         0.5285
                                                0.618
                                                                                                                   - 0.520
                                                           - 0.80
             max
                                                                                                                    - 0.515
                                                           - 0.72
                                                                             0.514
                                                                                                         0.5313
                                                                                                                    - 0.510
                                      0.9103
                                                                            0.5027
                                                                                      0.5037
                             0.9987
                                                0.7403
                                                                                                0.508
                   0.9997
                             10 100
param_min_samples_split
                                                 500
                                                                                                          500
```

```
In [99]:
          #Best Estimator and Best tune parameters
          print(clf4.best estimator )
          #Mean cross-validated score of the best estimator
          print(clf4.score(X_tfidf_w2v_train,y_train))
          print(clf4.score(X_tfidf_w2v_test,y_test))
          DecisionTreeClassifier(ccp alpha=0.0, class weight='balanced', criterion='gin
          i',
                                  max_depth=10, max_features=None, max_leaf_nodes=None,
                                  min_impurity_decrease=0.0, min_impurity_split=None,
                                  min_samples_leaf=1, min_samples_split=500,
                                  min weight fraction leaf=0.0, presort='deprecated',
                                  random state=None, splitter='best')
          0.6940483714070352
          0.532880611595636
In [100]:
          best_tune_parameters= [{'max_depth': [10], 'min_samples_split':[500] }]
```

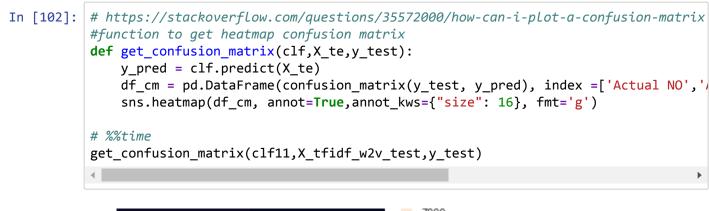
#### 2.9.4 Receiver Operating Characteristic- TFIDF W2V

# In [101]:

```
#Fitting Model to Hyper-Parameter Curve
from sklearn.metrics import roc curve, auc
from sklearn.tree import DecisionTreeClassifier
from sklearn.metrics import roc curve, auc
clf11= DecisionTreeClassifier(class_weight = 'balanced',max_depth=10,min_samples)
# clfV1=DecisionTreeClassifier (class_weight = 'balanced',max_depth=5,min_sample
clf11.fit(X tfidf w2v train, y train)
# for visulation
# clfV1.fit(X_tfidf_w2v_train, y_train)
#https://scikitlearn.org/stable/modules/generated/sklearn.linear model.SGDClassi
y_train_pred1 = clf11.predict_proba(X_tfidf_w2v_train) [:,1]
y_test_pred1 = clf11.predict_proba(X_tfidf_w2v_test) [:,1]
train_fpr1, train_tpr1, tr_thresholds1 = roc_curve(y_train, y_train_pred1)
test fpr1, test tpr1, te thresholds1 = roc curve(y test, y test pred1)
plt.plot(train_fpr1, train_tpr1, label="train AUC ="+str(auc(train_fpr1, train_t)
plt.plot(test fpr1, test tpr1, label="test AUC ="+str(auc(test fpr1, test tpr1))
plt.legend()
plt.xlabel("False Positive Rate")
plt.vlabel("True Positive Rate")
plt.title("ERROR PLOTS")
plt.grid(True)
plt.show()
```



## 2.9.5 Confusion matrix- TFIDF W2V



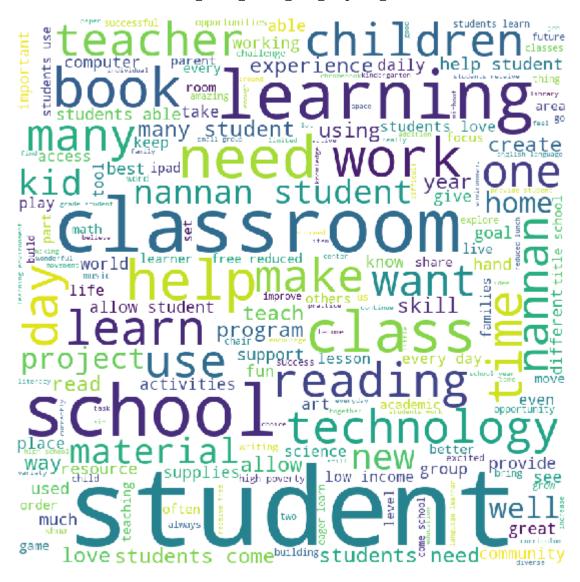


2.10 Selecting the FPR data points from Essay,Price and Teacher\_number\_of\_previously\_posted\_projects test data set.

```
In [103]:
          import numpy as np
          import pandas as pd
          y pred = clf11.predict(X tfidf w2v test)
          #Essay
          df=pd.DataFrame(Xtfidf_w2v_vectors_test_)
          #Price
          df =X test['price']
          df .to frame()
          #Teacher_number_of_previously_posted_projects
          df 1=X test['teacher number of previously posted projects']
          df_1.to_frame()
          df1=pd.DataFrame(y_test,columns=['y_test'])
          df2=pd.DataFrame(y pred,columns=['y pred'])
          df row = pd.concat([df,df1, df2],axis=1)
          df_row1 = pd.concat([df_,df1, df2],axis=1)
          df_row2 = pd.concat([df_1,df1, df2],axis=1)
          #Selecting the false positive data points
          df_row = df_row[(df_row.y_pred == 1) & (df_row.y_test == 0) ]
          df_row1 = df_row1[(df_row1.y_pred == 1) & (df_row1.y_test == 0) ]
          df_row2 = df_row2[(df_row2.y_pred == 1) & (df_row2.y_test == 0) ]
          df row2=df row2.fillna(0)
```

#### 2.11 WordCloud of False positive data points- TFIDF W2V

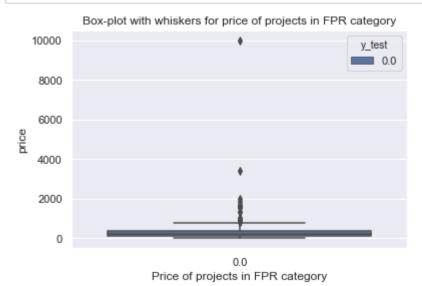
```
In [104]: # https://www.geeksforgeeks.org/generating-word-cloud-python/
          from wordcloud import WordCloud, STOPWORDS
           import matplotlib.pyplot as plt
           import pandas as pd
          # Reads 'Youtube04-Eminem.csv' file
          # df = pd.read csv(r"Youtube04-Eminem.csv", encoding ="latin-1")
          comment words = ' '
           stopwords = set(STOPWORDS)
          # iterate through the csv file
          for val in df row[0]:
              # typecaste each val to string
              val = str(val)
              # split the value
              tokens = val.split()
              # Converts each token into Lowercase
              for i in range(len(tokens)):
                   tokens[i] = tokens[i].lower()
              for words in tokens:
                   comment words = comment words + words + ' '
          wordcloud = WordCloud(width = 800, height = 800,
                           background_color ='white',
                           stopwords = stopwords,
                           min font size = 10).generate(comment words)
          # plot the WordCloud image
          plt.figure(figsize = (8, 8), facecolor = None)
          plt.imshow(wordcloud)
          plt.axis("off")
          plt.tight layout(pad = 0)
          plt.show()
```



2.12 Box plot of "price" of False positive data points

### In [108]:

```
import seaborn as sns
import matplotlib.pyplot as plt
g=sns.boxplot(x='y_test',y='price',hue='y_test',data=df_row1)
plt.title("Box-plot with whiskers for price of projects in FPR category")
plt.xlabel("Price of projects in FPR category")
# plt.legend(loc=1)
plt.show()
```



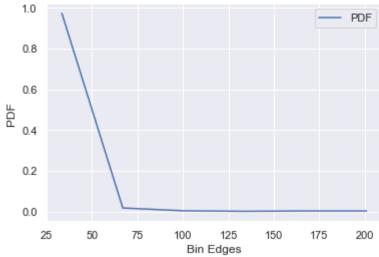
2.13 PDF of teacher\_number\_of\_previously\_posted\_projects for false positive data points.

## 

The pdf is: [9.73252804e-01 1.72562554e-02 3.45125108e-03 8.62812770e-04 2.58843831e-03 2.58843831e-03]
The bin edges of pdf are: [ 0. 33.5 67. 100.5 134. 167.5 201. ]

#### Out[106]: <matplotlib.legend.Legend at 0x3b103f28>





```
In [109]: # print("PDF1 of Patients who survived falling in >=5years from the graph-Blue to
print("The probability of finding project quote's less than ",round(bin_edges1[1])
print("The probability of finding project quote's less than ",round(bin_edges1[2)
print("The probability of finding project quote's less than ",round(bin_edges1[3)
print("The probability of finding project quote's less than ",round(bin_edges1[4)
print("The probability of finding project quote's less than ",round(bin_edges1[5)
print("The probability of finding project quote's less than ",round(bin_edges1[6)
```

The probability of finding project quote's less than 33.5 that got approved and falling in FPR category are 97.07 %

The probability of finding project quote's less than  $\,$  67.0 that got approved an d falling in FPR category are 1.68 %

The probability of finding project quote's less than  $\,$  100.5 that got approved a nd falling in FPR category are 0.63 %

The probability of finding project quote's less than  $\,$  134.0 that got approved a nd falling in FPR category are 0.42 %

The probability of finding project quote's less than 167.5 that got approved a nd falling in FPR category are 0.1 %

The probability of finding project quote's less than 201.0 that got approved a nd falling in FPR category are 0.1 %

```
In [ ]: #To make best use of the memory we are setting the variable names to 'None' and p
X_tfidf_w2v_test=None
X_tfidf_w2v_train=None
y_tfidf_w2v_train=None
gc.collect()
```

#### 2.14 Pretty table summary

```
In [119]: #http://zetcode.com/python/prettytable/
         from prettytable import PrettyTable
        x = PrettyTable()
        x.field_names = ["Vectorizer","Model", "Hyper Parameter-Alpha", "Test-AUC"]
        # x.add_row(["BOW","Decision Tree", '0.0001', 0.64])
        x.add_row(["Tfidf","Decision Tree" ,'max_depth:[10], min_samples_split:[500]', 0
        # x.add_row(["avg_w2v","Decision Tree",'0.0001', 0.58])
        x.add_row(["Tfidf- Non zero feature importance","Decision Tree",'max_depth: [10]
        x.add_row(["Tfidf_w2v","Decision Tree",'max_depth: [10], min_samples_split:[500]
        # x.add_row(["No_Text", "Decision Tree", '0.0001', 0.52])
        print(x)
        Model
                                                              Hyper Parameter
                    Vectorizer
                      | Test-AUC |
                      Tfidf
                                       | Decision Tree | max_depth:[10], min_samp
        les_split:[500] | 0.62 |
         | Tfidf- Non zero feature importance | Decision Tree | max depth: [10], min sam
        ples_split:[500] | 0.62 |
                                       | Decision Tree | max_depth: [10], min_sam
                    Tfidf_w2v
        ples_split:[500] | 0.53 |
         -----
          -----+
```

In [ ]: