#### **SVM 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 1.head(34000)
        project data 0=project data 0.tail(16000)
        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=['project essay 1'])
        project data.sort values(by=['project essay 2'], ascending=False)
        project_data.sort_values(by=['project_essay_3'])
        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_submitted_datetime' 'project_grade_category'
          'project subject categories' 'project subject subcategories'
         'project title' 'project essay 1' 'project essay 2' 'project essay 3'
          'project_essay_4' 'project_resource_summary'
         'teacher_number_of_previously_posted_projects' 'project_is_approved']
        print("Number of data points in resource data", resource_data.shape)
In [4]:
        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]:
                                                 description quantity price
                Ы
         0 p233245 LC652 - Lakeshore Double-Space Mobile Drying Rack
                                                                   149.0
```

```
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
Out[5]: 32
```

# 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 split
                    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
    j=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.cd
        # 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
                                  ,'') # we are placeing all the ' '(space) with ''(empty)
                j = j.replace(' '
                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
1	140945	p258326	897464ce9ddc600bced1151f324dd63a	Mr.	FL	
3	45	p246581	f3cb9bffbba169bef1a77b243e620b60	Mrs.	KY	
4						<b>+</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
1	140945	p258326	897464ce9ddc600bced1151f324dd63a	Mr.	FL	
3	45	p246581	f3cb9bffbba169bef1a77b243e620b60	Mrs.	KY	
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
```

```
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)

#### 2.2.6 Adding column Cost per project in dataset

Out[14]: (50000, 14)

```
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.head(2)
Out[16]:
             Unnamed:
                            id
                                                   teacher_id teacher_prefix school_state project_sul
               140945 p258326 897464ce9ddc600bced1151f324dd63a
                                                                      Mr.
                                                                                  FL
          1
                                                                                  ΚY
                   45 p246581
                                f3cb9bffbba169bef1a77b243e620b60
                                                                     Mrs.
In [17]:
         #To make best use of the memory we are setting the variable names to 'None' and
          resource_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"\'re", " are", phrase)
    phrase = re.sub(r"\'s", " is", phrase)
    phrase = re.sub(r"\'d", " would", phrase)
    phrase = re.sub(r"\'d", " will", phrase)
    phrase = re.sub(r"\'t", " not", phrase)
    phrase = re.sub(r"\'t", " have", phrase)
    phrase = re.sub(r"\'ve", " have", phrase)
    phrase = re.sub(r"\'ve", " have", phrase)
    return phrase
```

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

What does \"The message in the music\" mean to you? To my kids, it means so muc h more! Music is not just for entertainment! My kids are learning how music ca n be used as a human rights tool! They are learning to express their feelings o f the world around them in positive ways.\r\n\r\nMy kids are every stereotype y ou can think of: poor, underprivileged, rough, angry. Turn that same prism aro und there is also joy, happiness, intelligence and creativity. They are learnin g to navigate the world around them and we are working on showing them how to u se this energy in a positive way. Our piano lab is aging. We have 3 classes of 35 kids that are bursting at the gills, which is a great problem to have! BUT keyboards are beginning to have the keys break off making them unuseable. We al so do not have enough keyboards for every student in class, even if you counted the slightly broken ones!\r\n\r\nWe are in desperate need of more keyboards for our labs to meet the needs of our students. A keyboard for each student in cla ss will facilitate their love of music and help them grow and appreciate the wo rld around them. We are down to 24 semi working keyboards and 35 kids in each piano class. These keyboards are needed sooner rather than later for our kids to have a successful year in class.nannan

\_\_\_\_\_

```
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)
```

What does The message in the music mean to you? To my kids, it means so much more! Music is not just for entertainment! My kids are learning how music can be used as a human rights tool! They are learning to express their feelings of the world around them in positive ways. My kids are every stereotype you can think of: poor, underprivileged, rough, angry. Turn that same prism around the re is also joy, happiness, intelligence and creativity. They are learning to na vigate the world around them and we are working on showing them how to use this energy in a positive way. Our piano lab is aging. We have 3 classes of 35 kids that are bursting at the gills, which is a great problem to have! BUT keyboard s are beginning to have the keys break off making them unuseable. We also do no t have enough keyboards for every student in class, even if you counted the sli ghtly broken ones! We are in desperate need of more keyboards for our labs t o meet the needs of our students. A keyboard for each student in class will fa cilitate their love of music and help them grow and appreciate the world around We are down to 24 semi working keyboards and 35 kids in each piano clas s. These keyboards are needed sooner rather than later for our kids to have a successful year in class.nannan

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

What does The message in the music mean to you To my kids it means so much more Music is not just for entertainment My kids are learning how music can be used as a human rights tool They are learning to express their feelings of the world around them in positive ways My kids are every stereotype you can think of poor underprivileged rough angry Turn that same prism around there is also joy happi ness intelligence and creativity They are learning to navigate the world around them and we are working on showing them how to use this energy in a positive wa y Our piano lab is aging We have 3 classes of 35 kids that are bursting at the gills which is a great problem to have BUT keyboards are beginning to have the keys break off making them unuseable We also do not have enough keyboards for e very student in class even if you counted the slightly broken ones We are in de sperate need of more keyboards for our labs to meet the needs of our students A keyboard for each student in class will facilitate their love of music and help them grow and appreciate the world around them We are down to 24 semi working k eyboards and 35 kids in each piano class These keyboards are needed sooner rath er than later for our kids to have a successful year in class 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('\\r', ' ')
        sent = sent.replace('\\n', ' ')
        sent = re.sub('[^A-Za-z0-9]+', ' ', sent)
        # https://gist.github.com/sebleier/554280
        sent = ' '.join(e for e in sent.split() if e not in stopwords)
        preprocessed_essays.append(sent.lower().strip())
```

100%| 50000/50000 [03:35<00:00, 232.14it/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** 140945 p258326 897464ce9ddc600bced1151f324dd63a Mr. FL

**1** 45 p246581 f3cb9bffbba169bef1a77b243e620b60 Mrs. KY

```
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())
```

100% | 50000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 10000 | 100000 | 100000 | 100000 | 100000 | 100000 | 100000 | 100000 | 100000 | 100000 | 1000000 | 1000

```
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
                                                                                 FL
               140945 p258326 897464ce9ddc600bced1151f324dd63a
                                                                     Mr.
          1
                   45 p246581
                               f3cb9bffbba169bef1a77b243e620b60
                                                                                 KY
                                                                    Mrs.
         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': 15726, 'Math_Science': 12827, 'Health_Sports': 43
         73, 'SpecialNeeds': 4305, 'AppliedLearning': 3801, 'Music_Arts': 3148, 'History
          Civics': 1762, 'Warmth': 379, 'Care Hunger': 379})
```

#### 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
```

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

pecialNeeds', 'Health\_Sports', 'Math\_Science', 'Literacy\_Language']

```
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', 'Ci
         vics_Government', 'Extracurricular', 'ForeignLanguages', 'Warmth', 'Care_Hunge
         r', 'NutritionEducation', 'SocialSciences', 'PerformingArts', 'CharacterEducati
         on', 'TeamSports', 'Other', 'College_CareerPrep', 'Music', 'History_Geography',
         'Health_LifeScience', 'EarlyDevelopment', 'ESL', 'Gym_Fitness', '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', 'NH', 'AK', 'DE', 'SD', 'NE', 'HI', 'WV', 'ME',
         'NM', 'DC', 'IA', 'ID', 'KS', 'AR', 'CO', 'OR', 'MN', 'MS', 'KY', 'MD', 'NV',
'UT', 'AL', 'TN', 'CT', 'WI', 'VA', 'OK', 'NJ', 'AZ', 'WA', 'LA', 'MA', 'OH',
         'MO', 'IN', 'PA', 'MI', 'SC', 'GA', '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_cat&
         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)
         (33500, 16) (33500,)
         (16500, 16) (16500,)
         After vectorizations
         (33500, 5) (33500,)
         (16500, 5) (16500,)
         ['Dr.', 'Teacher', 'Mr.', 'Ms.', 'Mrs.']
         ______
```

## 2.3 Make Data Model Ready: Vectorizing Numerical features

# 2.3.1 Vectorizing Numerical features--Price

```
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,)
```

(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
         (33500, 1) (33500,)
         (16500, 1) (16500,)
```

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

\_\_\_\_\_

**Vectorizing Text data** 

=============

2.4.1 Bag of words: Essays

```
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)
        # We are considering only the words which appeared in at least 10 documents(rows
        vectorizer = CountVectorizer(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_bow = vectorizer.transform(X_train['essay'].values)
        # X_cv_essay_bow = vectorizer.transform(X_cv['essay'].values)
        X test essay bow = vectorizer.transform(X test['essay'].values)
        print("After vectorizations")
        print(X_train_essay_bow.shape, y_train.shape)
        # print(X_cv_essay_bow.shape, y_cv.shape)
        print(X test essay bow.shape, y test.shape)
        print("="*100)
        (33500, 16) (33500,)
        (16500, 16) (16500,)
        ______
        After vectorizations
        (33500, 5000) (33500,)
        (16500, 5000) (16500,)
        ______
```

#### 2.4.2 Bag of words: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)
         vectorizer = CountVectorizer(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_bow = vectorizer.transform(X_train['project_title'].values
         # X cv project title bow = vectorizer.transform(X cv['project title'].values)
         X test project title bow = vectorizer.transform(X test['project title'].values)
         print("After vectorizations")
         print(X_train_project_title_bow.shape, y_train.shape)
         # print(X_cv_project_title_bow.shape, y_cv.shape)
         print(X test project title bow.shape, y test.shape)
         print("="*100)
         (33500, 16) (33500,)
         (16500, 16) (16500,)
         ==============
         After vectorizations
         (33500, 1041) (33500,)
         (16500, 1041) (16500,)
In [46]: from scipy.sparse import hstack
         # with the same hstack function we are concatinating a sparse matrix and a dense
         X BOW TRAIN = hstack((X train digits in summary norm, X train teacher number of p
         X BOW TRAIN=X BOW TRAIN.todense()
         X_BOW_TRAIN=np.array(X_BOW_TRAIN)
         # X BOW cv = hstack((X cv project title bow, X cv essay bow , X cv digits in summar
         # X BOW cv=X BOW cv.todense()
         # X BOW cv=np.array(X BOW cv)
         X_BOW_test = hstack((X_test_digits_in_summary_norm,X_test_teacher_number_of_prev)
         X BOW test=X_BOW_test.todense()
         X BOW test=np.array(X BOW test)
         X train project title bow=None
         X_train_essay_bow =None
         X test project title bow=None
         X test essay bow =None
```

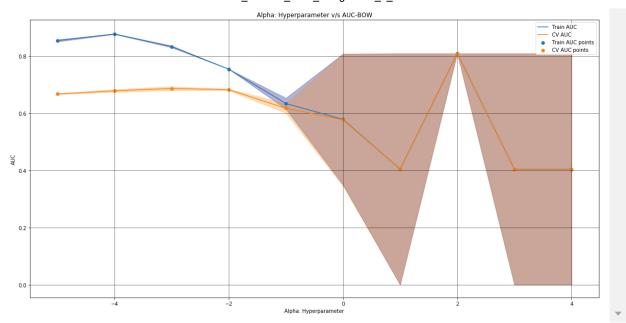
# 2.5 Applying SVM on BOW, SET 1

# 2.5.1 Applying SVM & GridSearchCV on Train data to obtain the best C

```
In [48]: #code source: http://occam.olin.edu/sites/default/files/DataScienceMaterials/mack
         from sklearn.model selection import train test split
         from sklearn.model selection import GridSearchCV
         from sklearn.datasets import *
         from sklearn.linear model import LogisticRegression
         from sklearn.linear_model import SGDClassifier
         from sklearn.metrics import f1 score
         from sklearn.calibration import CalibratedClassifierCV
         from sklearn.svm import LinearSVC
         from sklearn.svm import SVC
         # data = load breast cancer() #refer: http://scikit-learn.org/stable/modules/gene
         tuned parameters = { 'alpha': [10**-4, 10**-3, 10**-2, 10**-1, 10**0, 10**1, 10**]
         # X_train, X_test, y_train, y_test = train_test_split(data.data, data.target, tre
         # LR=SGDClassifier(loss='log',class weight="balanced")
         #Using GridSearchCV
         # SGDClassifier(loss='log',class weight="balanced",max iter=1500)
         # LogisticRegression(penalty = 'l1', C = i,class_weight="balanced")
         model = GridSearchCV(SGDClassifier(penalty = '12',loss='hinge',class_weight="bal
         # model=GridSearchCV(SVC(kernel='rbf',class weight="balanced",max iter = 2000),
         model.fit(X BOW TRAIN, y train)
         print(model.best estimator )
         print(model.score(X BOW test, y test))
         C:\Users\Admin\Anaconda3\lib\site-packages\sklearn\metrics\classification.py:
         1437: UndefinedMetricWarning:
         F-score is ill-defined and being set to 0.0 due to no predicted samples.
         C:\Users\Admin\Anaconda3\lib\site-packages\sklearn\metrics\classification.py:
         1437: UndefinedMetricWarning:
         F-score is ill-defined and being set to 0.0 due to no predicted samples.
         C:\Users\Admin\Anaconda3\lib\site-packages\sklearn\metrics\classification.py:
         1437: UndefinedMetricWarning:
         F-score is ill-defined and being set to 0.0 due to no predicted samples.
         C:\Users\Admin\Anaconda3\lib\site-packages\sklearn\metrics\classification.py:
         1437: UndefinedMetricWarning:
         F-score is ill-defined and being set to 0.0 due to no predicted samples.
         C:\Users\Admin\Anaconda3\lib\site-packages\sklearn\metrics\classification.py:
         1437: UndefinedMetricWarning:
         F-score is ill-defined and being set to 0.0 due to no predicted samples.
         C:\Users\Admin\Anaconda3\lib\site-packages\sklearn\metrics\classification.py:
         1437: UndefinedMetricWarning:
```

F-score is ill-defined and being set to 0.0 due to no predicted samples.

```
In [49]: # from sklearn.model selection import GridSearchCV
         # from sklearn.naive bayes import MultinomialNB
         # # nb = MultinomialNB(class prior=[0.5,0.5])
         # parameters = [0.00001, 0.0001, 0.001, 0.01, 0.1, 0.5, 0.8, 1, 10, 100, 1000]
         alpha=[*tuned parameters.values()]
         alpha=alpha[0]
         log alphas=[math.log10(num) for num in alpha]
         # clf = GridSearchCV(nb, parameters, cv= 10, scoring='roc auc',return train score
         # clf.fit(X_BOW_TRAIN, y_train)
         #Selecting the best parameter
         best_alpha1=model.best_params_
         # print(best alpha1)
         train_auc= model.cv_results_['mean_train_score']
         # print(train auc)
         train_auc_std= model.cv_results_['std_train_score']
         # print(train auc std)
         cv_auc = model.cv_results_['mean_test_score']
         # print(cv auc)
         cv auc std= model.cv results ['std test score']
         # print(cv auc std)
         # print(log alphas)
         plt.figure(figsize=(20,10))
         plt.plot(log alphas, train auc, label='Train AUC')
         # this code is copied from here: https://stackoverflow.com/a/48803361/4084039
         plt.gca().fill_between(log_alphas,train_auc - train_auc_std,train_auc + train_au
         plt.plot(log alphas, cv auc, label='CV AUC')
         # this code is copied from here: https://stackoverflow.com/a/48803361/4084039
         plt.gca().fill between(log alphas,cv auc - cv auc std,cv auc + cv auc std,alpha=
         plt.scatter(log alphas, train auc, label='Train AUC points')
         plt.scatter(log alphas, cv auc, label='CV AUC points')
         plt.legend()
         plt.xlabel("Alpha: Hyperparameter")
         plt.ylabel("AUC")
         plt.title("Alpha: Hyperparameter v/s AUC-BOW")
         plt.grid(color='black', linestyle='-', linewidth=0.5)
         plt.show()
```



```
In [50]: print("The best alpha from the above graph is ",best_alpha1)
best_alpha1['alpha']
```

The best alpha from the above graph is {'alpha': 100}

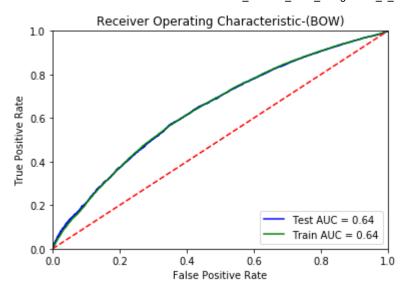
Out[50]: 100

# 2.5.2 Receiver Operating Characteristic- (BOW)

```
In [51]: # https://scikit-learn.org/stable/modules/generated/sklearn.metrics.roc curve.html
         from sklearn.metrics import roc curve, auc
         from sklearn.linear model import SGDClassifier
         from sklearn.calibration import CalibratedClassifierCV
         from sklearn.svm import SVC
         # train model on the best k
         # SVM=CalibratedClassifierCV(SGDClassifier(alpha=best alpha1['C'],penalty = 'l2']
         SVM=CalibratedClassifierCV(SGDClassifier(alpha=best alpha1['alpha'],penalty = 'l
         SVM.fit(X BOW TRAIN, y train)
         # roc_auc_score(y_true, y_score) the 2nd parameter should be probability estimate
         # not the predicted outputs
         probs = SVM.predict proba(X BOW test)
         # print(len(probs[:,1]))
         probs1 = SVM.predict proba(X BOW TRAIN)
         # print(len(probs1[:,1]))
         preds = probs[:,1]
         # print(preds)
         preds1 = probs1[:,1]
         # print(preds1)
         fpr, tpr, threshold = metrics.roc curve(y test, preds)
         # print(fpr,tpr,threshold)
         fpr1, tpr1, threshold_1 = metrics.roc_curve(y_train, preds1)
         # print(fpr1,tpr1,threshold)
         roc auc = metrics.auc(fpr, tpr)
         print(roc_auc)
         roc auc1 = metrics.auc(fpr1, tpr1)
         print(roc auc1)
         # https://www.programcreek.com/python/example/81207/sklearn.metrics.roc curve
         import matplotlib.pyplot as plt
         plt.title('Receiver Operating Characteristic-(BOW)')
         plt.plot(fpr, tpr, 'b', label = 'Test AUC = %0.2f' % roc_auc)
         plt.plot(fpr1, tpr1, 'g', label = 'Train AUC = %0.2f' % roc_auc1)
         plt.legend(loc = 'lower right')
         plt.plot([0, 1], [0, 1], 'r--')
         plt.xlim([0, 1])
         plt.ylim([0, 1])
         plt.ylabel('True Positive Rate')
         plt.xlabel('False Positive Rate')
         plt.show()
         # print(model.best estimator )
```

0.6427751951331497

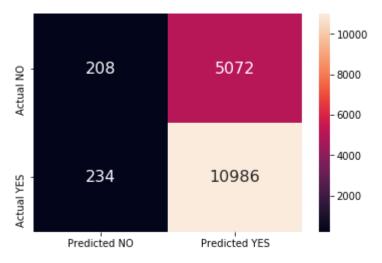
0.6427527870415264



#### 2.5.3 Confusion matrix- BOW

```
In [52]: # 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(SVM,X_BOW_test,y_test)
```



In [53]: #To make best use of the memory we are setting the variable names to 'None' and p
X\_BOW\_TRAIN=None
y\_BOW\_train=None
gc.collect()

Out[53]: 5784

#### 2.6 TFIDF vectorizer

#### 2.6.1 TFIDF vectorizer: Essay

```
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)
        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 [55]: # 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, 1034) (33500,)
        (16500, 1034) (16500,)
        ______
        ================
In [56]: 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)
```

## 2.6.3 Applying SVM on TFIDF, SET 2

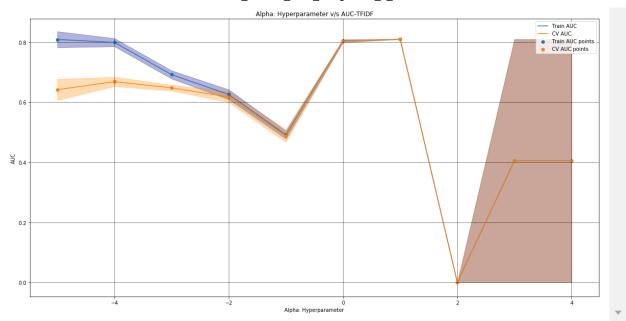
Applying SVM & GridSearchCV on Train data to obtain the best C

```
In [58]: #code source: http://occam.olin.edu/sites/default/files/DataScienceMaterials/mac/
         from sklearn.model selection import train test split
         from sklearn.model selection import GridSearchCV
         from sklearn.datasets import *
         from sklearn.linear model import LogisticRegression
         # data = load breast cancer() #refer: http://scikit-learn.org/stable/modules/gene
         tuned parameters = {'alpha': [10**-4, 10**-3, 10**-2, 10**-1, 10**0, 10**1, 10**]
         # X_train, X_test, y_train, y_test = train_test_split(data.data, data.target, tr@
         #Using GridSearchCV
         model = GridSearchCV(SGDClassifier(penalty = '12',loss='hinge',class weight="bal
         model.fit(X Tfidf train, y train)
         print(model.best estimator )
         print(model.score(X_Tfidf_test, y_test))
         C:\Users\Admin\Anaconda3\lib\site-packages\sklearn\metrics\classification.py:
         1437: UndefinedMetricWarning:
         F-score is ill-defined and being set to 0.0 due to no predicted samples.
         C:\Users\Admin\Anaconda3\lib\site-packages\sklearn\metrics\classification.py:
         1437: UndefinedMetricWarning:
         F-score is ill-defined and being set to 0.0 due to no predicted samples.
         C:\Users\Admin\Anaconda3\lib\site-packages\sklearn\metrics\classification.py:
         1437: UndefinedMetricWarning:
         F-score is ill-defined and being set to 0.0 due to no predicted samples.
         C:\Users\Admin\Anaconda3\lib\site-packages\sklearn\metrics\classification.py:
         1437: UndefinedMetricWarning:
         F-score is ill-defined and being set to 0.0 due to no predicted samples.
         C:\Users\Admin\Anaconda3\lib\site-packages\sklearn\metrics\classification.py:
         1437: UndefinedMetricWarning:
         F-score is ill-defined and being set to 0.0 due to no predicted samples.
         C:\Users\Admin\Anaconda3\lib\site-packages\sklearn\metrics\classification.py:
         1437: UndefinedMetricWarning:
         F-score is ill-defined and being set to 0.0 due to no predicted samples.
         C:\Users\Admin\Anaconda3\lib\site-packages\sklearn\metrics\classification.py:
         1437: UndefinedMetricWarning:
         F-score is ill-defined and being set to 0.0 due to no predicted samples.
         C:\Users\Admin\Anaconda3\lib\site-packages\sklearn\metrics\classification.py:
         1437: UndefinedMetricWarning:
```

F-score is ill-defined and being set to 0.0 due to no predicted samples.

```
In [59]: # from sklearn.model selection import GridSearchCV
         # from sklearn.naive bayes import MultinomialNB
         # # nb = MultinomialNB(class prior=[0.5,0.5])
         # parameters = [0.00001, 0.0001, 0.001, 0.01, 0.1, 0.5, 0.8, 1, 10, 100, 1000]
         alpha=[*tuned parameters.values()]
         alpha=alpha[0]
         log alphas=[math.log10(num) for num in alpha]
         # clf = GridSearchCV(nb, parameters, cv= 10, scoring='roc auc',return train score
         # clf.fit(X_BOW_TRAIN, y_train)
         #Selecting the best parameter
         best_alpha1=model.best_params_
         # print(best alpha1)
         train auc= model.cv results ['mean train score']
         # print(train auc)
         train auc std= model.cv_results_['std_train_score']
         # print(train auc std)
         cv_auc = model.cv_results_['mean_test_score']
         # print(cv auc)
         cv auc std= model.cv results ['std test score']
         # print(cv auc std)
         print(log alphas)
         plt.figure(figsize=(20,10))
         plt.plot(log alphas, train auc, label='Train AUC')
         # this code is copied from here: https://stackoverflow.com/a/48803361/4084039
         plt.gca().fill_between(log_alphas,train_auc - train_auc_std,train_auc + train_au
         plt.plot(log alphas, cv auc, label='CV AUC')
         # this code is copied from here: https://stackoverflow.com/a/48803361/4084039
         plt.gca().fill between(log alphas,cv auc - cv auc std,cv auc + cv auc std,alpha=
         plt.scatter(log alphas, train auc, label='Train AUC points')
         plt.scatter(log alphas, cv auc, label='CV AUC points')
         plt.legend()
         plt.xlabel("Alpha: Hyperparameter")
         plt.ylabel("AUC")
         plt.title("Alpha: Hyperparameter v/s AUC-TFIDF")
         plt.grid(color='black', linestyle='-', linewidth=0.5)
         plt.show()
```

[-5.0, -4.0, -3.0, -2.0, -1.0, 0.0, 1.0, 2.0, 3.0, 4.0]

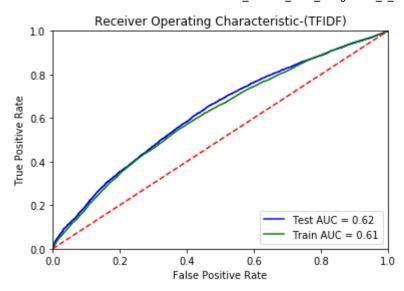


```
In [60]: # best_alpha1=0.01
print("The best alpha from the above graph is ",best_alpha1['alpha'])
```

The best alpha from the above graph is 10

# 2.6.4 Receiver Operating Characteristic- (TFIDF)

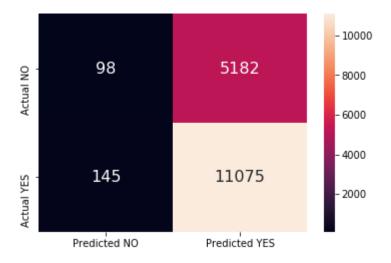
```
In [61]: # https://scikit-learn.org/stable/modules/generated/sklearn.metrics.roc_curve.htm
         from sklearn.metrics import roc_curve, auc
         # train model on the best k
         SVM=CalibratedClassifierCV(SGDClassifier(alpha=best alpha1['alpha'],penalty = 'l
         SVM.fit(X_Tfidf_train, y_train)
         # roc auc score(y true, y score) the 2nd parameter should be probability estimate
         # not the predicted outputs
         probs = SVM.predict_proba(X_Tfidf_test)
         # print(len(probs[:,1]))
         probs1 = SVM.predict_proba(X_Tfidf_train)
         # print(len(probs1[:,1]))
         preds = probs[:,1]
         # print(preds)
         preds1 = probs1[:,1]
         # print(preds1)
         fpr, tpr, threshold = metrics.roc_curve(y_test, preds)
         # print(fpr,tpr,threshold)
         fpr1, tpr1, threshold = metrics.roc curve(y train, preds1)
         # print(fpr1,tpr1,threshold)
         roc_auc = metrics.auc(fpr, tpr)
         # print(roc auc)
         roc_auc1 = metrics.auc(fpr1, tpr1)
         # print(roc auc1)
         # https://www.programcreek.com/python/example/81207/sklearn.metrics.roc curve
         import matplotlib.pyplot as plt
         plt.title('Receiver Operating Characteristic-(TFIDF)')
         plt.plot(fpr, tpr, 'b', label = 'Test AUC = %0.2f' % roc_auc)
         plt.plot(fpr1, tpr1, 'g', label = 'Train AUC = %0.2f' % roc auc1)
         plt.legend(loc = 'lower right')
         plt.plot([0, 1], [0, 1], 'r--')
         plt.xlim([0, 1])
         plt.ylim([0, 1])
         plt.ylabel('True Positive Rate')
         plt.xlabel('False Positive Rate')
         plt.show()
```



### 2.6.5 Confusion matrix-TFIDF

```
In [62]: # 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','/outsine since sheatmap(df_cm, annot=True,annot_kws={"size": 16}, fmt='g')

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



```
In [63]: #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[63]: 3158

## 2.7 AVG\_W2V

### 2.7.1 Using Pretrained Models: Avg W2V-Essays

```
In [79]: # train test split
         from sklearn.model_selection import train_test_split
         avg_w2v_vectors_essay_train_, avg_w2v_vectors_essay_test_, y_train, y_test = train
         # avg w2v vectors essay train, avg w2v vectors essay cv, y train, y cv = train te
In [80]: # 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 [81]: # average Word2Vec
         # compute average word2vec for each review.
         from tqdm import tqdm
         # sentence=[]
         avg w2v vectors essay train = []; # the avg-w2v for each sentence/review is store
         for sentence in tqdm(avg_w2v_vectors_essay_train_): # 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 essay train.append(vector)
         print(len(avg_w2v_vectors_essay_train))
         print(len(avg w2v vectors essay train[0]))
         100%
                                                 | 33500/33500 [01:12<00:00, 462.71it/s]
         33500
```

localhost:8888/notebooks/Donors\_Choose\_SVM\_Assignment\_6\_1.ipynb#

300

```
In [82]: # average Word2Vec
         # compute average word2vec for each review.
         from tqdm import tqdm
         avg w2v vectors essay test = []; # the avg-w2v for each sentence/review is stored
         for sentence in tqdm(avg_w2v_vectors_essay_test_): # 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 essay test.append(vector)
         print(len(avg_w2v_vectors_essay_test))
         print(len(avg w2v vectors essay test[0]))
         100%
                                                  || 16500/16500 [00:36<00:00, 451.02it/s]
         16500
         300
In [83]: # # We are considering only the words which appeared in at least 10 documents(rol
         # vectorizer = CountVectorizer(min_df=10, max_features=1000)
         # project essay avq w2v = vectorizer.fit transform(avq w2v vectors essay)
         # print("Shape of matrix after one hot encodig ",project essay avg w2v.shape)
         import scipy
         avg_w2v_vectors_essay_train=scipy.sparse.csr_matrix(avg_w2v_vectors_essay_train)
         # type(avg w2v vectors essay)
         import scipy
         avg w2v vectors essay test=scipy.sparse.csr matrix(avg w2v vectors essay test)
         # type(avg w2v vectors essay test)
```

## 2.7.2 Using Pretrained Models: AVG W2V on project\_title

```
In [84]: # train test split
    from sklearn.model_selection import train_test_split
    avg_w2v_vectors_Pro_title_train_, avg_w2v_vectors_Pro_title_test_, y_train, y_te:
    # avg_w2v_vectors_Pro_title_train, avg_w2v_vectors_Pro_title_cv, y_train, y_cv =
```

```
In [85]: # S = ["abc def pqr", "def def def abc", "pqr pqr def"]
    tfidf_model = TfidfVectorizer()
    tfidf_model.fit(avg_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 [86]: # average Word2Vec
         # compute average word2vec for each review.
         avg w2v vectors Pro title train = []; # the avg-w2v for each sentence/review is
         for sentence in tqdm(avg w2v vectors Pro title train ): # for each review/senten
             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_Pro_title_train.append(vector)
         print(len(avg w2v vectors Pro title train))
         print(len(avg_w2v_vectors_Pro_title_train[0]))
                                          | 33500/33500 [00:04<00:00, 8303.94it/s]
         100%
         33500
         300
In [87]: # average Word2Vec
         # compute average word2vec for each review.
         avg w2v vectors Pro title test = []; # the avg-w2v for each sentence/review is s
         for sentence in tqdm(avg w2v vectors Pro title test ): # 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 Pro title test.append(vector)
         print(len(avg_w2v_vectors_Pro_title_test))
         print(len(avg w2v vectors Pro title test[0]))
         100%
                                       16500/16500 [00:02<00:00, 8071.95it/s]
         16500
         300
```

```
In [88]: import scipy
    avg_w2v_vectors_Pro_title_train=scipy.sparse.csr_matrix(avg_w2v_vectors_Pro_title
    type(avg_w2v_vectors_Pro_title_train)

import scipy
    avg_w2v_vectors_Pro_title_test=scipy.sparse.csr_matrix(avg_w2v_vectors_Pro_title_type(avg_w2v_vectors_Pro_title_test)

# import scipy
# avg_w2v_vectors_Pro_title_cv=scipy.sparse.csr_matrix(avg_w2v_vectors_Pro_title_type(avg_w2v_vectors_Pro_title_cv))

Out[88]: scipy.sparse.csr.csr_matrix
```

```
In [89]: X_avg_w2v_train = hstack(( avg_w2v_vectors_Pro_title_train,avg_w2v_vectors_essay.
X_avg_w2v_train=X_avg_w2v_train.todense()
X_avg_w2v_train=np.array(X_avg_w2v_train)

# X_avg_w2v_cv = hstack(( avg_w2v_vectors_Pro_title_cv,avg_w2v_vectors_essay_cv,)
# X_avg_w2v_cv=X_avg_w2v_cv.todense()
# X_avg_w2v_cv=np.array(X_avg_w2v_cv)

X_avg_w2v_test = hstack(( avg_w2v_vectors_Pro_title_test,avg_w2v_vectors_essay_text_X_avg_w2v_test=X_avg_w2v_test.todense()
X_avg_w2v_test=None
avg_w2v_vectors_Pro_title_train=None
avg_w2v_vectors_Pro_title_test=None
avg_w2v_vectors_essay_test=None
avg_w2v_vectors_essay_test=None
avg_w2v_vectors_essay_test=None
```

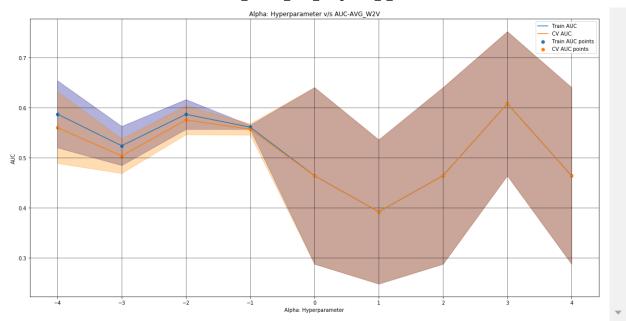
### 2.7.3 Applying SVM on AVG W2V, SET 3

Applying SVM & GridSearchCV on Train data to obtain the best C

```
#code source: http://occam.olin.edu/sites/default/files/DataScienceMaterials/macl
from sklearn.model selection import train test split
from sklearn.model selection import GridSearchCV
from sklearn.datasets import *
from sklearn.linear model import LogisticRegression
# data = load breast cancer() #refer: http://scikit-learn.org/stable/modules/gene
tuned parameters = {'alpha': [10**-4, 10**-3, 10**-2, 10**-1, 10**0, 10**1, 10**]
#Using GridSearchCV
model = GridSearchCV(SGDClassifier(penalty = '12',loss='hinge',class weight="bal
model.fit(X avg w2v train, y train)
print(model.best estimator )
print(model.score(X_avg_w2v_test, y_test))
SGDClassifier(alpha=1000, average=False, class weight='balanced',
             early stopping=False, epsilon=0.1, eta0=0.0, fit intercept=True,
             11_ratio=0.15, learning_rate='optimal', loss='hinge',
             max iter=1500, n iter no change=5, n jobs=None, penalty='12',
             power_t=0.5, random_state=None, shuffle=True, tol=0.001,
             validation fraction=0.1, verbose=0, warm start=False)
0.68
```

```
In [97]: # from sklearn.model selection import GridSearchCV
         # from sklearn.naive bayes import MultinomialNB
         # # nb = MultinomialNB(class prior=[0.5,0.5])
         # parameters = [0.00001, 0.0001, 0.001, 0.01, 0.1, 0.5, 0.8, 1, 10, 100, 1000]
         alpha=[*tuned parameters.values()]
         alpha=alpha[0]
         log alphas=[math.log10(num) for num in alpha]
         # clf = GridSearchCV(nb, parameters, cv= 10, scoring='roc auc',return train score
         # clf.fit(X BOW TRAIN, y train)
         #Selecting the best parameter
         best_alpha1=model.best_params_
         # print(best alpha1)
         train auc= model.cv results ['mean train score']
         # print(train auc)
         train auc std= model.cv_results_['std_train_score']
         # print(train auc std)
         cv_auc = model.cv_results_['mean_test_score']
         # print(cv auc)
         cv auc std= model.cv results ['std test score']
         # print(cv auc std)
         print(log alphas)
         plt.figure(figsize=(20,10))
         plt.plot(log alphas, train auc, label='Train AUC')
         # this code is copied from here: https://stackoverflow.com/a/48803361/4084039
         plt.gca().fill_between(log_alphas,train_auc - train_auc_std,train_auc + train_au
         plt.plot(log alphas, cv auc, label='CV AUC')
         # this code is copied from here: https://stackoverflow.com/a/48803361/4084039
         plt.gca().fill between(log alphas,cv auc - cv auc std,cv auc + cv auc std,alpha=
         plt.scatter(log alphas, train auc, label='Train AUC points')
         plt.scatter(log alphas, cv auc, label='CV AUC points')
         plt.legend()
         plt.xlabel("Alpha: Hyperparameter")
         plt.ylabel("AUC")
         plt.title("Alpha: Hyperparameter v/s AUC-AVG_W2V")
         plt.grid(color='black', linestyle='-', linewidth=0.5)
         plt.show()
```

[-4.0, -3.0, -2.0, -1.0, 0.0, 1.0, 2.0, 3.0, 4.0]

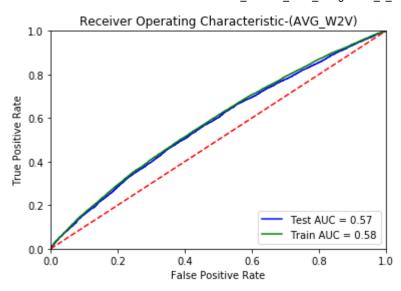


In [98]: print("The best alpha from the above graph is ",best\_alpha1['alpha'])
# best\_alpha1['alpha']

The best alpha from the above graph is 1000

# 2.7.4 Receiver Operating Characteristic- (AVG\_W2V)

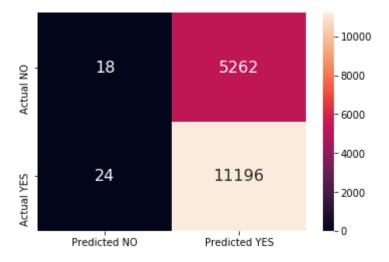
```
In [99]: # https://scikit-learn.org/stable/modules/generated/sklearn.metrics.roc_curve.htm
         from sklearn.metrics import roc_curve, auc
         # train model on the best k
         SVM=CalibratedClassifierCV(SGDClassifier(alpha=best alpha1['alpha'],penalty = 'l
         SVM.fit(X_avg_w2v_train, y_train)
         # roc_auc_score(y_true, y_score) the 2nd parameter should be probability estimate
         # not the predicted outputs
         probs = SVM.predict_proba(X_avg_w2v_test)
         # print(len(probs[:,1]))
         probs1 = SVM.predict_proba(X_avg_w2v_train)
         # print(len(probs1[:,1]))
         preds = probs[:,1]
         # print(preds)
         preds1 = probs1[:,1]
         # print(preds1)
         fpr, tpr, threshold = metrics.roc_curve(y_test, preds)
         # print(fpr,tpr,threshold)
         fpr1, tpr1, threshold = metrics.roc curve(y train, preds1)
         # print(fpr1,tpr1,threshold)
         roc_auc = metrics.auc(fpr, tpr)
         # print(roc auc)
         roc_auc1 = metrics.auc(fpr1, tpr1)
         # print(roc auc1)
         # https://www.programcreek.com/python/example/81207/sklearn.metrics.roc curve
         import matplotlib.pyplot as plt
         plt.title('Receiver Operating Characteristic-(AVG W2V)')
         plt.plot(fpr, tpr, 'b', label = 'Test AUC = %0.2f' % roc_auc)
         plt.plot(fpr1, tpr1, 'g', label = 'Train AUC = %0.2f' % roc auc1)
         plt.legend(loc = 'lower right')
         plt.plot([0, 1], [0, 1], 'r--')
         plt.xlim([0, 1])
         plt.ylim([0, 1])
         plt.ylabel('True Positive Rate')
         plt.xlabel('False Positive Rate')
         plt.show()
```



# 2.7.5 Confusion matrix-AVG\_W2V

```
In [100]: # 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','/sins.heatmap(df_cm, annot=True,annot_kws={"size": 16}, fmt='g')

# %%time
get_confusion_matrix(SVM,X_avg_w2v_test,y_test)
```



```
In [101]: #To make best use of the memory we are setting the variable names to 'None' and p
X_avg_w2v_test=None
X_avg_w2v_train=None
y_avg_w2v_train=None
gc.collect()
```

Out[101]: 9064

## 2.8.1 Using Pretrained Models: TFIDF weighted W2V-Essay

```
In [102]: # 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
          # S = ["abc def pqr", "def def def abc", "pqr pqr def"]
In [103]:
          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 [104]: # 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]))
```

```
33500/33500 [08:47<00:00, 63.53it/s]
100%
```

33500 300

```
In [105]: # 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 [04:23<00:00, 62.60it/s]
                         16500
                         300
In [106]:
                         import scipy
                          tfidf_w2v_vectors_essay_train=scipy.sparse.csr_matrix(tfidf_w2v_vectors_essay_train=scipy.sparse.csr_matrix(tfidf_w2v_vectors_essay_train=scipy.sparse.csr_matrix(tfidf_w2v_vectors_essay_train=scipy.sparse.csr_matrix(tfidf_w2v_vectors_essay_train=scipy.sparse.csr_matrix(tfidf_w2v_vectors_essay_train=scipy.sparse.csr_matrix(tfidf_w2v_vectors_essay_train=scipy.sparse.csr_matrix(tfidf_w2v_vectors_essay_train=scipy.sparse.csr_matrix(tfidf_w2v_vectors_essay_train=scipy.sparse.csr_matrix(tfidf_w2v_vectors_essay_train=scipy.sparse.csr_matrix(tfidf_w2v_vectors_essay_train=scipy.sparse.csr_matrix(tfidf_w2v_vectors_essay_train=scipy.sparse.csr_matrix(tfidf_w2v_vectors_essay_train=scipy.sparse.csr_matrix(tfidf_w2v_vectors_essay_train=scipy.sparse.csr_matrix(tfidf_w2v_vectors_essay_train=scipy.sparse.csr_matrix(tfidf_w2v_vectors_essay_train=scipy.sparse.csr_matrix(tfidf_w2v_vectors_essay_train=scipy.sparse.csr_matrix(tfidf_w2v_vectors_essay_train=scipy.sparse.csr_matrix(tfidf_w2v_vectors_essay_train=scipy.sparse.csr_matrix(tfidf_w2v_vectors_essay_train=scipy.sparse.csr_watrix(tfidf_w2v_vectors_essay_train=scipy.sparse.csr_watrix(tfidf_w2v_vectors_essay_train=scipy.sparse.csr_watrix(tfidf_w2v_vectors_essay_train=scipy.sparse.csr_watrix(tfidf_w2v_vectors_essay_train=scipy.sparse.csr_watrix(tfidf_w2v_vectors_essay_train=scipy.sparse.csr_watrix(tfidf_w2v_vectors_essay_train=scipy.sparse.csr_watrix(tfidf_w2v_vectors_essay_train=scipy.sparse.csr_watrix(tfidf_w2v_vectors_essay_train=scipy.sparse.csr_watrix(tfidf_w2v_vectors_essay_train=scipy.sparse.csr_watrix(tfidf_w2v_vectors_essay_train=scipy.sparse.csr_watrix(tfidf_w2v_vectors_essay_train=scipy.sparse.csr_watrix(tfidf_w2v_vectors_essay_train=scipy.sparse.csr_watrix(tfidf_w2v_vectors_essay_train=scipy.sparse.csr_watrix(tfidf_w2v_vectors_essay_train=scipy.sparse.csr_watrix(tfidf_w2v_vectors_essay_train=scipy.sparse.csr_watrix(tfidf_w2v_vectors_essay_train=scipy.sparse.csr_watrix(tfidf_w2v_vectors_essay_train=scipy.sparse.csr_watrix(tfidf_w2v_vectors_essay_train=scipy.sparse.csr_w
                          type(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[106]: scipy.sparse.csr.csr matrix
```

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

tfidf words = set(tfidf model.get feature names())

# 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 )))

```
In [109]: # 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 [00:08<00:00, 3828.79it/s]
          100%
          33500
          300
In [110]: # 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/senter
              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 [00:04<00:00, 3643.78it/s]
          16500
          300
```

```
In [111]:
          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[111]: scipy.sparse.csr.csr matrix
In [112]: 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_tfidf_w2v_test=X_tfidf_w2v_test.todense()
          X tfidf w2v test=np.array(X tfidf w2v test)
          # X All = hstack((categories one hot, sub categories one hot, school state one hot,
```

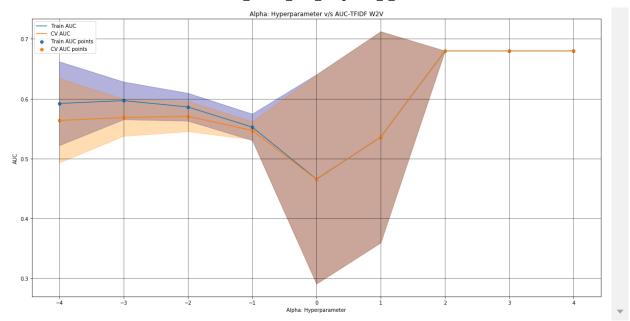
### 2.8.3 Applying SVM on TFIDF W2V, SET 4

Applying SVM & GridSearchCV on Train data to obtain the best C

```
#code source: http://occam.olin.edu/sites/default/files/DataScienceMaterials/macl
from sklearn.model selection import train test split
from sklearn.model selection import GridSearchCV
from sklearn.datasets import *
from sklearn.linear model import LogisticRegression
from sklearn.linear_model import SGDClassifier
# data = load breast cancer() #refer: http://scikit-learn.org/stable/modules/gene
tuned_parameters = {'alpha': [10**-4, 10**-3, 10**-2, 10**-1, 10**0, 10**1, 10**]
# X train, X test, y train, y test = train test split(data.data, data.target, tre
#Using GridSearchCV
model = GridSearchCV(SGDClassifier(penalty = '12',loss='hinge',class weight="bald")
model.fit(X_tfidf_w2v_train, y_train)
print(model.best estimator )
print(model.score(X_tfidf_w2v_test, y_test))
SGDClassifier(alpha=100, average=False, class weight='balanced',
              early_stopping=False, epsilon=0.1, eta0=0.0, fit_intercept=True,
              11_ratio=0.15, learning_rate='optimal', loss='hinge',
              max_iter=1500, n_iter_no_change=5, n_jobs=None, penalty='12',
              power t=0.5, random state=None, shuffle=True, tol=0.001,
              validation_fraction=0.1, verbose=0, warm_start=False)
0.68
```

```
In [114]: # from sklearn.model selection import GridSearchCV
          # from sklearn.naive bayes import MultinomialNB
          # # nb = MultinomialNB(class prior=[0.5,0.5])
          # parameters = [0.00001, 0.0001, 0.001, 0.01, 0.1, 0.5, 0.8, 1, 10, 100, 1000]
          alpha=[*tuned parameters.values()]
          alpha=alpha[0]
          log alphas=[math.log10(num) for num in alpha]
          # clf = GridSearchCV(nb, parameters, cv= 10, scoring='roc auc',return train score
          # clf.fit(X_BOW_TRAIN, y_train)
          #Selecting the best parameter
          best_alpha1=model.best_params_
          # print(best alpha1)
          train auc= model.cv results ['mean train score']
          # print(train auc)
          train auc std= model.cv_results_['std_train_score']
          # print(train auc std)
          cv_auc = model.cv_results_['mean_test_score']
          # print(cv auc)
          cv auc std= model.cv results ['std test score']
          # print(cv auc std)
          print(log alphas)
          plt.figure(figsize=(20,10))
          plt.plot(log alphas, train auc, label='Train AUC')
          # this code is copied from here: https://stackoverflow.com/a/48803361/4084039
          plt.gca().fill_between(log_alphas,train_auc - train_auc_std,train_auc + train_au
          plt.plot(log alphas, cv auc, label='CV AUC')
          # this code is copied from here: https://stackoverflow.com/a/48803361/4084039
          plt.gca().fill between(log alphas,cv auc - cv auc std,cv auc + cv auc std,alpha=
          plt.scatter(log alphas, train auc, label='Train AUC points')
          plt.scatter(log alphas, cv auc, label='CV AUC points')
          plt.legend()
          plt.xlabel("Alpha: Hyperparameter")
          plt.ylabel("AUC")
          plt.title("Alpha: Hyperparameter v/s AUC-TFIDF W2V")
          plt.grid(color='black', linestyle='-', linewidth=0.5)
          plt.show()
```

[-4.0, -3.0, -2.0, -1.0, 0.0, 1.0, 2.0, 3.0, 4.0]

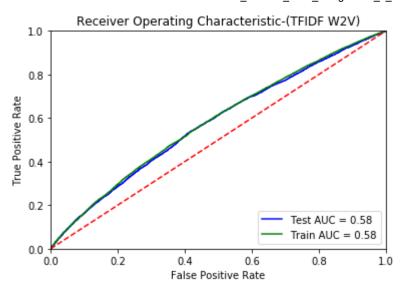


In [115]: print("The best alpha from the above graph is ",best\_alpha1['alpha'])
# best\_alpha1['alpha']

The best alpha from the above graph is 100

# 2.8.4 Receiver Operating Characteristic- TFIDF W2V

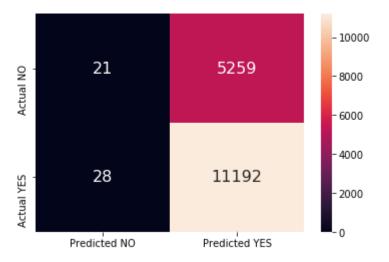
```
In [116]: # https://scikit-learn.org/stable/modules/generated/sklearn.metrics.roc curve.html
          from sklearn.metrics import roc curve, auc
          from sklearn.calibration import CalibratedClassifierCV
          # train model on the best k
          SVM=CalibratedClassifierCV(SGDClassifier(alpha=best alpha1['alpha'],penalty = 'l
          SVM.fit(X_tfidf_w2v_train, y_train)
          # roc auc score(y true, y score) the 2nd parameter should be probability estimate
          # not the predicted outputs
          probs = SVM.predict_proba(X_tfidf_w2v_test)
          # print(len(probs[:,1]))
          probs1 = SVM.predict_proba(X_tfidf_w2v_train)
          # print(len(probs1[:,1]))
          preds = probs[:,1]
          # print(preds)
          preds1 = probs1[:,1]
          # print(preds1)
          fpr, tpr, threshold = metrics.roc_curve(y_test, preds)
          # print(fpr,tpr,threshold)
          fpr1, tpr1, threshold = metrics.roc curve(y train, preds1)
          # print(fpr1,tpr1,threshold)
          roc_auc = metrics.auc(fpr, tpr)
          # print(roc auc)
          roc_auc1 = metrics.auc(fpr1, tpr1)
          # print(roc auc1)
          # https://www.programcreek.com/python/example/81207/sklearn.metrics.roc curve
          import matplotlib.pyplot as plt
          plt.title('Receiver Operating Characteristic-(TFIDF W2V)')
          plt.plot(fpr, tpr, 'b', label = 'Test AUC = %0.2f' % roc_auc)
          plt.plot(fpr1, tpr1, 'g', label = 'Train AUC = %0.2f' % roc auc1)
          plt.legend(loc = 'lower right')
          plt.plot([0, 1], [0, 1], 'r--')
          plt.xlim([0, 1])
          plt.ylim([0, 1])
          plt.ylabel('True Positive Rate')
          plt.xlabel('False Positive Rate')
          plt.show()
```



#### 2.8.5 Confusion matrix- TFIDF W2V

```
In [117]: # 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(SVM,X_tfidf_w2v_test,y_test)
```



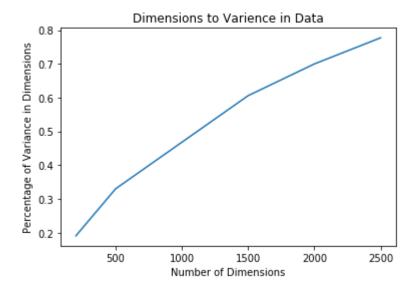
```
In [118]: #To make best use of the memory we are setting the variable names to 'None' and p
X_tfidf_w2v_test=None
X_tfidf_w2v_train=None
y_tfidf_w2v_train=None
gc.collect()
```

Out[118]: 5591

### 2.9.1 Applying TruncatedSVD to obtain the best dimension.

## 2.9.2 Plot of the Percentage of variance in Dimensions

```
In [47]: plt.xlabel("Number of Dimensions")
    plt.ylabel("Percentage of Variance in Dimensions")
    plt.title("Dimensions to Varience in Data")
    plt.plot(Di,Varience_sum)
    plt.show()
```



## 2.9.3 Applying the TruncatedSVD on Train Essay feature

```
In [48]: #At 2500 dimensions we are getting 80% efficiency.
    #But we are taking 3000 dimensions since we can approx that it would increase the
    svd = TruncatedSVD(n_components= 3000)
    svd.fit(X_train_essay_Tfidf)
    #Transforms:
    #Train SVD
    X_train_essay_Tfidf= svd.transform(X_train_essay_Tfidf)
    #Test SVD
    X_test_essay_Tfidf = svd.transform(X_test_essay_Tfidf)
```

```
In [50]: from scipy.sparse import hstack
    X_NO_text_train = hstack((X_train_essay_Tfidf,X_train_digits_in_summary_norm,X_train_train=X_NO_text_train.todense()
    X_NO_text_train=np.array(X_NO_text_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_NO_text_test = hstack((X_test_essay_Tfidf,X_test_digits_in_summary_norm,X_test_X_NO_text_test=X_NO_text_test.todense()
    X_NO_text_test=np.array(X_NO_text_test)
```

### 2.9.4 Applying SVM on No text features, SET 5

Applying SVM & GridSearchCV on Train data to obtain the best C

```
In [54]: #code source: http://occam.olin.edu/sites/default/files/DataScienceMaterials/macl
from sklearn.model_selection import train_test_split
from sklearn.model_selection import GridSearchCV
from sklearn.datasets import *
from sklearn.linear_model import LogisticRegression
from sklearn.linear_model import SGDClassifier

# data = Load_breast_cancer() #refer: http://scikit-learn.org/stable/modules/gene
tuned_parameters = {'alpha': [0.00001,0.0001,0.001,0.01,1,10,100,1000,10000]]
# X_train, X_test, y_train, y_test = train_test_split(data.data, data.target, tree

#Using GridSearchCV
model = GridSearchCV(SGDClassifier(penalty ='12',loss='hinge',class_weight="balamodel.fit(X_NO_text_train, y_train))
print(model.best_estimator_)
print(model.score(X_NO_text_test, y_test))
```

C:\Users\Admin\Anaconda3\lib\site-packages\sklearn\metrics\classification.py:14
37: UndefinedMetricWarning:

F-score is ill-defined and being set to 0.0 due to no predicted samples.

C:\Users\Admin\Anaconda3\lib\site-packages\sklearn\metrics\classification.py:14
37: UndefinedMetricWarning:

F-score is ill-defined and being set to 0.0 due to no predicted samples.

C:\Users\Admin\Anaconda3\lib\site-packages\sklearn\metrics\classification.py:14
37: UndefinedMetricWarning:

F-score is ill-defined and being set to 0.0 due to no predicted samples.

C:\Users\Admin\Anaconda3\lib\site-packages\sklearn\metrics\classification.py:14
37: UndefinedMetricWarning:

F-score is ill-defined and being set to 0.0 due to no predicted samples.

C:\Users\Admin\Anaconda3\lib\site-packages\sklearn\metrics\classification.py:14
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C:\Users\Admin\Anaconda3\lib\site-packages\sklearn\metrics\classification.py:14
37: UndefinedMetricWarning:

F-score is ill-defined and being set to 0.0 due to no predicted samples.

C:\Users\Admin\Anaconda3\lib\site-packages\sklearn\metrics\classification.py:14
37: UndefinedMetricWarning:

F-score is ill-defined and being set to 0.0 due to no predicted samples.

C:\Users\Admin\Anaconda3\lib\site-packages\sklearn\metrics\classification.py:14

37: UndefinedMetricWarning:

F-score is ill-defined and being set to 0.0 due to no predicted samples.

C:\Users\Admin\Anaconda3\lib\site-packages\sklearn\metrics\classification.py:14
37: UndefinedMetricWarning:

F-score is ill-defined and being set to 0.0 due to no predicted samples.

C:\Users\Admin\Anaconda3\lib\site-packages\sklearn\metrics\classification.py:14
37: UndefinedMetricWarning:

F-score is ill-defined and being set to 0.0 due to no predicted samples.

C:\Users\Admin\Anaconda3\lib\site-packages\sklearn\metrics\classification.py:14
37: UndefinedMetricWarning:

F-score is ill-defined and being set to 0.0 due to no predicted samples.

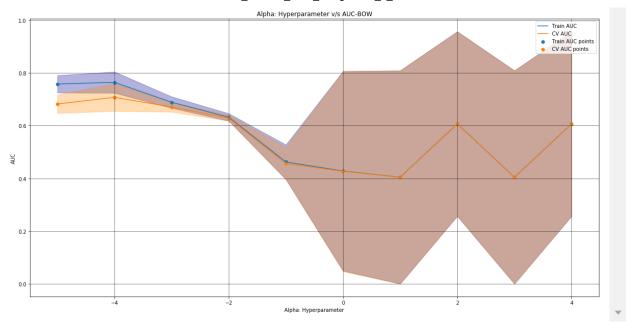
C:\Users\Admin\Anaconda3\lib\site-packages\sklearn\metrics\classification.py:14
37: UndefinedMetricWarning:

F-score is ill-defined and being set to 0.0 due to no predicted samples.

0.7101860920666014

```
In [63]: # from sklearn.model selection import GridSearchCV
         # from sklearn.naive bayes import MultinomialNB
         # # nb = MultinomialNB(class prior=[0.5,0.5])
         # parameters = [0.00001, 0.0001, 0.001, 0.01, 0.1, 0.5, 0.8, 1, 10, 100, 1000]
         alpha=[*tuned parameters.values()]
         alpha=alpha[0]
         log alphas=[math.log10(num) for num in alpha]
         # clf = GridSearchCV(nb, parameters, cv= 10, scoring='roc auc',return train score
         # clf.fit(X BOW TRAIN, y train)
         #Selecting the best parameter
         best_alpha1=model.best_params_
         # print(best alpha1)
         train auc= model.cv results ['mean train score']
         # print(train auc)
         train auc std= model.cv results ['std train score']
         # print(train auc std)
         cv_auc = model.cv_results_['mean_test_score']
         # print(cv auc)
         cv auc std= model.cv results ['std test score']
         # print(cv auc std)
         print(log alphas)
         plt.figure(figsize=(20,10))
         plt.plot(log alphas, train auc, label='Train AUC')
         # this code is copied from here: https://stackoverflow.com/a/48803361/4084039
         plt.gca().fill_between(log_alphas,train_auc - train_auc_std,train_auc + train_au
         plt.plot(log alphas, cv auc, label='CV AUC')
         # this code is copied from here: https://stackoverflow.com/a/48803361/4084039
         plt.gca().fill between(log alphas,cv auc - cv auc std,cv auc + cv auc std,alpha=
         plt.scatter(log alphas, train auc, label='Train AUC points')
         plt.scatter(log alphas, cv auc, label='CV AUC points')
         plt.legend()
         plt.xlabel("Alpha: Hyperparameter")
         plt.ylabel("AUC")
         plt.title("Alpha: Hyperparameter v/s AUC-BOW")
         plt.grid(color='black', linestyle='-', linewidth=0.5)
         plt.show()
```

[-5.0, -4.0, -3.0, -2.0, -1.0, 0.0, 1.0, 2.0, 3.0, 4.0]

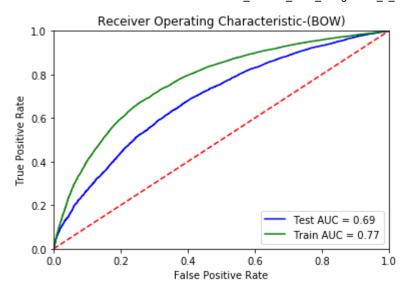


In [56]: print("The best alpha from the above graph is ",best\_alpha1)
# best\_alpha1['alpha']

The best alpha from the above graph is {'alpha': 0.0001}

# 2.9.5 Receiver Operating Characteristic- (No text features)

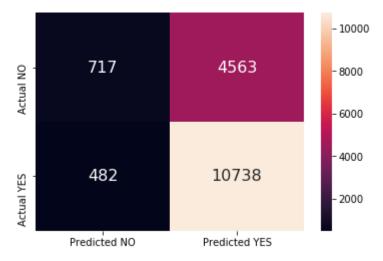
```
In [60]: # https://scikit-learn.org/stable/modules/generated/sklearn.metrics.roc curve.html
         from sklearn.metrics import roc_curve, auc
         # train model on the best k
         SVM=CalibratedClassifierCV(SGDClassifier(alpha=best alpha1['alpha'],penalty = 'l
         SVM.fit(X_NO_text_train, y_train)
         # roc_auc_score(y_true, y_score) the 2nd parameter should be probability estimate
         # not the predicted outputs
         probs = SVM.predict_proba(X_NO_text_test)
         # print(len(probs[:,1]))
         probs1 = SVM.predict_proba(X_NO_text_train)
         # print(len(probs1[:,1]))
         preds = probs[:,1]
         # print(preds)
         preds1 = probs1[:,1]
         # print(preds1)
         fpr, tpr, threshold = metrics.roc_curve(y_test, preds)
         # print(fpr,tpr,threshold)
         fpr1, tpr1, threshold = metrics.roc curve(y train, preds1)
         # print(fpr1,tpr1,threshold)
         roc_auc = metrics.auc(fpr, tpr)
         # print(roc auc)
         roc_auc1 = metrics.auc(fpr1, tpr1)
         # print(roc auc1)
         # https://www.programcreek.com/python/example/81207/sklearn.metrics.roc curve
         import matplotlib.pyplot as plt
         plt.title('Receiver Operating Characteristic-(BOW)')
         plt.plot(fpr, tpr, 'b', label = 'Test AUC = %0.2f' % roc_auc)
         plt.plot(fpr1, tpr1, 'g', label = 'Train AUC = %0.2f' % roc auc1)
         plt.legend(loc = 'lower right')
         plt.plot([0, 1], [0, 1], 'r--')
         plt.xlim([0, 1])
         plt.ylim([0, 1])
         plt.ylabel('True Positive Rate')
         plt.xlabel('False Positive Rate')
         plt.show()
```



#### 2.9.6 Confusion matrix- No text features

```
In [61]: # 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','/sins.heatmap(df_cm, annot=True,annot_kws={"size": 16}, fmt='g')

# %%time
get_confusion_matrix(SVM,X_NO_text_test,y_test)
```



In [ ]: #To make best use of the memory we are setting the variable names to 'None' and p
X\_NO\_text\_test=None
X\_NO\_text\_train=None
# y\_tfidf\_w2v\_train=None
gc.collect()

### 2.10 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", "SVM", '0.0001', 0.64])
x.add_row(["Tfidf", "SVM" ,'0.0001', 0.61])
x.add_row(["avg_w2v", "SVM", '0.0001', 0.57])
x.add_row(["tfidf_w2v", "SVM", '0.0001' , 0.58])
x.add_row(["No_Text", "SVM", '0.0001', 0.69])
print(x)
```

+	+	<b>+</b>	
Vectorizer	Model +	Hyper Parameter-Alpha +	Test-AUC
BOW Tfidf avg_w2v tfidf_w2v No_Text	SVM SVM SVM SVM	0.0001   0.0001   0.0001   0.0001   0.0001	0.64     0.61     0.57     0.58     0.69
+	+	<del> </del>	r

```
In [ ]:
```