Install and Import the libraries

Tn [41•

```
In [1]:
%matplotlib inline
import warnings
warnings.filterwarnings("ignore", category=UserWarning)
warnings.filterwarnings("ignore", category=FutureWarning)
warnings.filterwarnings("ignore", category=RuntimeWarning)
# Data Manipulation libraries
import pandas as pd
import numpy as np
#Plotting libraries
import matplotlib.pyplot as plt
import seaborn as sns
# Estimators and metrics
from sklearn.preprocessing import Normalizer
from sklearn.feature extraction.text import CountVectorizer, TfidfTransformer, TfidfVectorizer
from sklearn.model selection import train test split, KFold
from sklearn.linear model import SGDClassifier
from sklearn.svm import SVC
from sklearn.decomposition import TruncatedSVD
from sklearn.metrics import confusion matrix, roc curve, auc, roc auc score, accuracy score
from sklearn.calibration import CalibratedClassifierCV
# NLP libraries
import nltk
from nltk.corpus import stopwords
from gensim.models import Word2Vec, KeyedVectors
import re
import pickle
from tqdm import tqdm
from collections import Counter
from scipy.sparse import hstack
#Code Reference: https://ptable.readthedocs.io/en/latest/tutorial.html
from prettytable import PrettyTable
In [2]:
# Read the data into Pandas Dataframe
project data= pd.read csv('../train data.csv')
resource data = pd.read csv('../resources.csv')
In [3]:
print('Number of data points in the Train dataset :',project data.shape[0])
print("-"*53)
print('Number of features in the Train dataset :',project data.shape[1])
print("-"*53)
print("List of Features in the Train dataset:\n",project data.columns.values.tolist())
Number of data points in the Train dataset: 109248
_____
Number of features in the Train dataset : 17
List of Features in the Train dataset:
['Unnamed: 0', 'id', 'teacher id', 'teacher prefix', 'school state', 'project submitted datetime', 'project grade category', 'project subject categories', 'project subject subcategories', '
project title', 'project essay 1', 'project essay 2', 'project essay 3', 'project resource summary', 'teacher number of previously posted projects', 'project is approved']
```

```
in fal.
print('Number of data points in the Resourse dataset :',resource_data.shape[0])
print("-"*55)
print('Number of features in the Resourse dataset :',resource_data.shape[1])
print("-"*55)
print("List of Features in the Resourse dataset:", resource data.columns.values.tolist())
Number of data points in the Resourse dataset : 1541272
Number of features in the Resourse dataset : 4
List of Features in the Resourse dataset: ['id', 'description', 'quantity', 'price']
In [5]:
cols=['Date' if each col=='project submitted datetime' else each col for each col in project data.columns.values.tolist()]
project_data['Date']=pd.to_datetime(project_data['project_submitted_datetime'])
project data.drop('project submitted datetime',axis=1,inplace=True)
project_data.sort_values(by=['Date'],inplace=True)
In [6]:
project data=project data[cols]
```

```
project_data=project_data[cols]
print("Sample records from Training data ")
project_data.head()
```

Sample records from Training data

Out[6]:

	Unnamed: 0	id	teacher_ic	l teacher_prefix	school_state	Date	project_grade_category	project_subject_categories	project_subject_subcategories	project_title	project_essay_1	project_essay_2	project_essay_3	project_essay_4
55660	8393	3 p205479	2bf07ba08945e5d8b2a3f269b2b3cfe	5 Mrs.	CA	2016- 04-27 00:27:36	Grades PreK-2	Math & Science	Applied Sciences, Health & Life Science	Engineering STEAM into the Primary Classroom	I have been fortunate enough to use the Fairy	My students come from a variety of backgrounds	try to do	It is challenging to develop high quality scie
76127	37728	3 p043609	3f60494c61921b3b43ab61bdde2904d	f Ms.	UT	2016- 04-27 00:31:25	Grades 3-5	Special Needs	Special Needs	Sensory Tools for Focus	Imagine being 8-9 years old. You're in your th	Most of my students have autism, anxiety, anot	It is tough to do more than one thing at a tim	When my students are able to calm themselves d
51140	74477	′ p189804	4a97f3a390bfe21b99cf5e2b81981c7	3 Mrs.	CA	2016- 04-27 00:46:53	Grades PreK-2	Literacy & Language	Literacy	Mobile Learning with a Mobile Listening Center	Having a class of 24 students comes with diver	I have a class of twenty-four kindergarten stu	By having a mobile listening and storage cente	A mobile listening center will help keep equip
473	100660) p234804	cbc0e38f522143b86d372f8b43d4cff3	B Mrs.	GA	2016- 04-27 00:53:00	Grades PreK-2	Applied Learning	Early Development	Flexible Seating for Flexible Learning	I recently read an article about giving studen	I teach at a low- income (Title 1) school. Ever	We need a classroom rug that we can use as a c	Benjamin Franklin once said, \"Tell me and I f
41558	33679) p137682	06f6e62e17de34fcf81020c77549e1d	5 Mrs.	WA	2016- 04-27 01:05:25	Grades 3-5	Literacy & Language	Literacy	Going Deep: The Art of Inner Thinking!	My students crave challenge, they eat obstacle	We are an urban, public k- 5 elementary school	With the new common core standards that have b	These remarkable gifts will provide students w
4														Þ

In [7]:

```
print("Sample records from Resourse data ")
resource_data.head()
```

Sample records from Resourse data

Out[7]:

description quantity price

0	p233245	LC652 - Lakeshore Double-Space Mobile Drying Hack	quantity	149.66
1	p069063	Bouncy Bands for Desks (Blue support pipes)	3	14.95
2	p069063	Cory Stories: A Kid's Book About Living With Adhd	1	8.45
3	p069063	Dixon Ticonderoga Wood-Cased #2 HB Pencils, Bo	2	13.59
4	p069063	EDUCATIONAL INSIGHTS FLUORESCENT LIGHT FILTERS	3	24.95

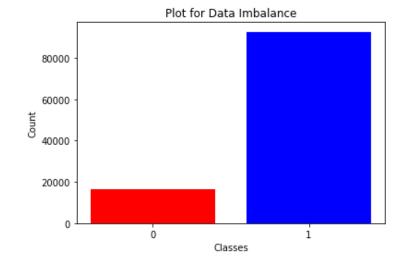
Data Analysis

```
In [8]:
def check class bal(dataset, target class):
    count_per_class=list(dataset[target_class].value_counts())
    classes=list(dataset[target_class].value_counts().index)
    print("Ratio of the classes :")
    for each cls, cls count in zip(classes, count per class):
       print("Class {} has {} records with a ratio of {}%".
              format(each cls,cls count,np.round((cls count/dataset.shape[0]*100),2)))
    plt.bar(classes, count_per_class, color=['b', 'r'])
    plt.xticks(classes)
    plt.ylabel("Count")
    plt.xlabel("Classes")
    plt.title("Plot for Data Imbalance")
    plt.show()
    del classes
    del count_per_class
```

In [9]:

```
check_class_bal(project_data, 'project_is_approved')

Ratio of the classes:
Class 1 has 92706 records with a ratio of 84.86%
Class 0 has 16542 records with a ratio of 15.14%
```



Data Preprocessing

chek for null values

```
In [10]:
print("Null values from Train data :\n")
```

```
print(project_data.isnull().sum())
Null values from Train data:
                                                    0
Unnamed: 0
                                                    0
id
teacher_id
                                                    0
teacher prefix
                                                    3
school state
                                                    0
                                                    0
Date
                                                    0
project_grade_category
                                                    0
project_subject_categories
                                                    0
project_subject_subcategories
                                                    0
project title
                                                    0
project essay 1
                                                    0
project essay 2
                                                105490
project essay 3
                                                105490
project essay 4
project_resource summary
                                                    0
teacher number of previously posted projects
                                                    0
                                                    0
project_is_approved
dtype: int64
In [11]:
project_data['teacher_prefix'].fillna(method='ffill',inplace=True)
In [12]:
project data['essay']=project data.project essay 1.map(str)+\
project_data.project_essay_2.map(str)+\
project data.project essay 3.map(str)+\
project_data.project_essay_4.map(str)
In [13]:
project_data.drop(columns=['project_essay_1', 'project_essay_2',
                           'project essay 3', 'project essay 4'], axis=1, inplace=True)
In [14]:
project data.isnull().sum()
Out[14]:
Unnamed: 0
                                                0
                                                0
id
                                                0
teacher id
teacher prefix
                                                0
school state
                                                0
Date
                                                0
project grade category
                                                0
                                                0
project subject categories
project subject subcategories
project_title
                                                0
project_resource_summary
teacher number of previously posted projects
                                                0
project_is_approved
                                                0
                                                0
essay
dtype: int64
In [15]:
print("Null values from Train data :\n")
print(resource data.isnull().sum())
Null values from Train data:
                 0
id
               292
description
                 0
quantity
                 0
price
```

Text Pre-processing

```
In [19]:

def get_sorted_dic(col):
    my_Counter=Counter()
    for word in list(project_data[col]):
        my_Counter.update(word.split())
    count_dict=dict(my_Counter)
    return dict(sorted(count_dict.items(),key=lambda x: x[1]))
```

project_subject_categories

```
In [20]:

clean_categories=processed_list(list(project_data['project_subject_categories']))
project_data['clean_categories']=clean_categories
project_data.drop(['project_subject_categories'],axis=1,inplace=True)
sorted_cat_dict=get_sorted_dic('clean_categories')
```

project_subject_subcategories

```
In [21]:

clean_sub_categories=processed_list(list(project_data['project_subject_subcategories']))
project_data['clean_sub_categories']=clean_sub_categories
project_data.drop(['project_subject_subcategories'],axis=1,inplace=True)
sorted_subcat_dict=get_sorted_dic('clean_sub_categories')
```

essay

```
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"\'r", " are", phrase)

phrase = re.sub(r"\'r", " are", phrase)

phrase = re.sub(r"\'s", " is", phrase)

phrase = re.sub(r"\'d", " would", phrase)

phrase = re.sub(r"\'d", " would", phrase)

phrase = re.sub(r"\'t", " not", phrase)

phrase = re.sub(r"\'t", " not", phrase)

phrase = re.sub(r"\'t", " have", phrase)

phrase = re.sub(r"\'r", " am", phrase)

return phrase
```

In [23]:

```
# https://gist.github.com/sebleier/554280
# we are removing the words from the stop words list: 'no', 'nor', 'not'
stopwords= ['i', 'me', 'my', 'myself', 'we', 'our', 'ours', 'ourselves', 'you', "you're", "you've", \
            "you'll", "you'd", 'your', 'yours', 'yourself', 'yourselves', 'he', 'him', 'his', 'himself', \
            'she', "she's", 'her', 'hers', 'herself', 'it', "it's", 'its', 'itself', 'they', 'them', 'their',\
            'theirs', 'themselves', 'what', 'whoi', 'whom', 'this', 'that', "that'll", 'these', 'those', \
            'am', 'is', 'are', 'was', 'were', 'be', 'been', 'being', 'have', 'has', 'had', 'having', 'do', 'does', \
            'did', 'doing', 'a', 'an', 'the', 'and', 'but', 'if', 'or', 'because', 'as', 'until', 'while', 'of', \setminus
            'at', 'by', 'for', 'with', 'about', 'against', 'between', 'into', 'through', 'during', 'before', 'after',\
            'above', 'below', 'to', 'from', 'up', 'down', 'in', 'out', 'on', 'off', 'over', 'under', 'again', 'further',\
            'then', 'once', 'here', 'there', 'when', 'where', 'why', 'how', 'all', 'any', 'both', 'each', 'few', 'more', \
            'most', 'other', 'some', 'such', 'only', 'own', 'same', 'so', 'than', 'too', 'very', \setminus
            's', 't', 'can', 'will', 'just', 'don', "don't", 'should', "should've", 'now', 'd', 'll', 'm', 'o', 're', \
            've', 'y', 'ain', 'aren', "aren't", 'couldn', "couldn't", 'didn', "didn't", 'doesn', "doesn't", 'hadn', \
            "hadn't", 'hasn', "hasn't", 'haven', "haven't", 'isn', "isn't", 'ma', 'mightn', "mightn't", 'mustn',
            "mustn't", 'needn', "needn't", 'shan', "shan't", 'shouldn', "shouldn't", 'wasn', "wasn't", 'weren', "weren't", \
            'won', "won't", 'wouldn', "wouldn't"]
```

In [24]:

```
def text_processing(dataset, feature_name):
    processed_text = []
    # tqdm is for printing the status bar
    for sentance in tqdm(dataset[feature_name].values):
        sent = decontracted(sentance)
        sent = sent.replace('\\r', '')
        sent = sent.replace('\\"', '')
        sent = sent.replace('\\"', '')
        sent = re.sub('[^A-Za-20-9]+', '', sent)
        sent = ''.join(e for e in sent.split() if e.lower() not in stopwords)
        processed_text.append(sent.lower().strip())
    return processed_text
```

In [25]:

```
project_data['essay'] = text_processing(project_data, 'essay')
100%| 109248/109248 [01:07<00:00, 1612.60it/s]</pre>
```

project_title

```
In [26]:
```

```
project_data['project_title']=text_processing(project_data,'project_title')

100%| 109248/109248 [00:02<00:00, 36756.25it/s]
```

project_resource_summary

In [27]:

```
project_data['project_resource_summary']=text_processing(project_data,'project_resource_summary')
100%| 109248/109248 [00:07<00:00, 14884.87it/s]</pre>
```

project_grade_category

```
In [28]:
```

```
processed_grade=[]

for each_grade in tqdm(project_data['project_grade_category'].values):
    temp=""
    temp=each_grade.lower()
    temp=temp.replace(' ',' ')
    temp=temp.replace(' '-', '_')
    processed_grade.append(temp)

project_data['project_grade_category']=processed_grade

100%| 100948/109248 [00:00<00:00, 928368.18it/s]</pre>
```

In [29]:

```
# Merge the projectdata and pricedata by using id feature

price_data = resource_data.groupby('id').agg({'price':'sum', 'quantity':'sum'}).reset_index()
project_data = pd.merge(project_data, price_data, on='id', how='left')
```

In [30]:

```
print("Final Feature Names:\n\n", list(project_data.columns))
print("\nSample Data set")
project_data.head()
```

Final Feature Names:

['Unnamed: 0', 'id', 'teacher_id', 'teacher_prefix', 'school_state', 'Date', 'project_grade_category', 'project_title', 'project_resource_summary', 'teacher_number_of_previously_posted_projects', 'project_is_approved', 'essay', 'clean_categories', 'quantity', 'price']

Sample Data set

Out[30]:

	Unn	amed: 0	id	teacher_id	teacher_prefix	school_state	Date	project_grade_category	project_title	project_resource_summary	teacher_number_of_previously_posted_projects	s project_is_approved	essay	clean_categories clea
Ó)	8393	p205479	2bf07ba08945e5d8b2a3f269b2b3cfe5	Mrs.	CA	2016- 04-27 00:27:36	grades_prek_2	engineering steam primary classroom	students need stem kits learn critical science	53	3 1	fortunate enough use fairy tale stem kits clas	Math_Science
1	I	37728	p043609	3f60494c61921b3b43ab61bdde2904df	Ms.	UT (2016- 04-27 00:31:25	grades_3_5	sensory tools focus	students need boogie boards quiet sensory brea	4	1 1	imagine 8 9 years old third grade classroom se	SpecialNeeds
2	2	74477	p189804	4a97f3a390bfe21b99cf5e2b81981c73	Mrs.	CA (2016- 04-27 00:46:53	grades_prek_2	mobile learning mobile listening center	students need mobile listening center able enh	10) 1	class 24 students comes diverse learners stude	Literacy_Language
\$	3 1	00660	p234804	cbc0e38f522143b86d372f8b43d4cff3	Mrs.	GA (2016- 04-27 00:53:00	grades_prek_2	flexible seating flexible learning	students need flexible seating classroom choos	2	2 1	recently read article giving students choice I	AppliedLearning

```
students
essay
crave
Unnamed:
                 id
                                              teacher_id teacher_prefix school_state
                                                                                         Date project_grade_category project_title project_resource_summary teacher_number_of_previously_posted_projects project_is_approved
                                                                                                                                                                                                                                                clean_categories clea
                                                                                         2016-
                                                                                                                        going deep
                                                                                                                                        tudents need copies new
    33679 p137682 06f6e62e17de34fcf81020c77549e1d5
                                                                   Mrs.
                                                                                 WA
                                                                                         04-27
                                                                                                           grades_3_5
                                                                                                                                                                                                                                               Literacy_Language
                                                                                                                           art inner
                                                                                                                                         york times best selle...
                                                                                                                                                                                                                                          eat
                                                                                      01:05:25
                                                                                                                           thinking
                                                                                                                                                                                                                                    obstacles
                                                                                                                                                                                                                                    breakfa...
```

In [31]:

```
y = project_data['project_is_approved'].values
X=project_data.drop(['project_is_approved'], axis=1)
project_data.head(3)
```

Out[31]:

Use Says possess possess possess possess possess possess possess possess posses posses posses primary classroom students need stem kits students need stem kits students need stem kits students need stem kits primary classroom students need stem kits clasm. 1 37728 possess posses possess possess posses		Unnar	med: 0	id		teacher_id	teacher_prefix	school_state	Date	project_grade_category	project_title	project_resource_summary	teacher_number_of_previously_posted_projects	project_is_approved	essay	clean_categories clea
1 37728 p043609 3f60494c61921b3b43ab61bdde2904df Ms. UT 04-27 grades_3_5 tools focus boards quiet sensory brea 2 74477 p189804 4a97f3a390bfe21b99cf5e2b81981c73 Mrs. CA 04-27 grades_prek_2 grades_prek_2 mobile learning center center.	0		8393 _l	p205479	2bf07ba08945e5d8b2	2a3f269b2b3cfe5	Mrs.	CA	04-27		steam primary		53	3 1	fortunate enough use fairy tale stem kits clas	Math_Science
2016- learning students need mobile 2 74477 p189804 4a97f3a390bfe21b99cf5e2b81981c73 Mrs. CA 04-27 grades_prek_2 mobile listening center able enh 00:46:53 listening center able enh Center students need mobile students need mobile listening center able enh	1	3	7728 j	p043609	3f60494c61921b3b43	ab61bdde2904df	Ms.	UT	04-27	grades_3_5	_	boards quiet sensory		1	imagine 8 9 years old third grade classroom se	SpecialNeeds
stude	2		4477 լ	p189804	4a97f3a390bfe21b99	cf5e2b81981c73	Mrs.	CA	04-27		learning mobile	students need mobile	10) 1	class 24 students comes diverse learners stude	Literacy_Language

In [32]:

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.30, stratify=y)
```

In [33]:

```
print("Training data set shape:", X_train.shape)
print("Test data set shape:", X_test.shape)
```

Training data set shape: (76473, 15)
Test data set shape: (32775, 15)

Feature Vectorization

teacher_prefix

In [34]:

```
vectorizer=CountVectorizer()
vectorizer.fit(X_train.teacher_prefix.values)
X_tr_teacher_onehot=vectorizer.transform(X_train.teacher_prefix.values)
X_te_teacher_onehot=vectorizer.transform(X_test.teacher_prefix.values)
```

school_state

In [35]:

vectorizer=CountVectorizer()

```
vectorizer.fit(X_train.school_state.values)
X_tr_school_onehot=vectorizer.transform(X_train.school_state.values)
X_te_school_onehot=vectorizer.transform(X_test.school_state.values)
```

project_grade_category

```
In [36]:

vectorizer=CountVectorizer()
vectorizer.fit(X_train.project_grade_category.values)
X_tr_grade_onehot=vectorizer.transform(X_train.project_grade_category.values)
X_te_grade_onehot=vectorizer.transform(X_test.project_grade_category.values)
```

clean_categories

```
In [37]:

vectorizer=CountVectorizer()
vectorizer.fit(X_train.clean_categories.values)
X_tr_cat_onehot=vectorizer.transform(X_train.clean_categories.values)
X_te_cat_onehot=vectorizer.transform(X_test.clean_categories.values)
```

clean_sub_categories

```
In [38]:

vectorizer=CountVectorizer()
vectorizer.fit(X_train.clean_sub_categories.values)
X_tr_sub_cat_onehot=vectorizer.transform(X_train.clean_sub_categories.values)
X_te_sub_cat_onehot=vectorizer.transform(X_test.clean_sub_categories.values)
```

Normalization

price

```
In [39]:

nrml= Normalizer()
nrml.fit(X_train['price'].values.reshape(1,-1))

X_tr_price_nrml = nrml.transform(X_train.price.values.reshape(1,-1)).reshape(-1,1)
X_te_price_nrml = nrml.transform(X_test.price.values.reshape(1,-1)).reshape(-1,1)
```

teacher_number_of_previously_posted_projects

```
In [40]:

nrml = Normalizer()

nrml.fit(X_train.teacher_number_of_previously_posted_projects.values.reshape(1,-1))

X_tr_teacher_number_nrml = nrml.transform(X_train.teacher_number_of_previously_posted_projects.values.reshape(1,-1)).reshape(-1,1)

X_te_teacher_number_nrml = nrml.transform(X_test.teacher_number_of_previously_posted_projects.values.reshape(1,-1)).reshape(-1,1)
```

```
print("After stacking :")
```

In [42]:

```
print("Training data set shape :", X_tr_vec.shape)
print("Test data set shape :", X_te_vec.shape)
After stacking :
Training data set shape: (76473, 101)
Test data set shape : (32775, 101)
Model Training
Hypertuning Values
In [43]:
alpha_vals=[10**i for i in range(-4,5)]
norm_values=['11','12']
def model=SGDClassifier()
In [44]:
print(" Default Model:\n", def model)
 Default Model:
 SGDClassifier(alpha=0.0001, average=False, class weight=None, epsilon=0.1,
       eta0=0.0, fit_intercept=True, l1_ratio=0.15,
       learning rate='optimal', loss='hinge', max iter=None, n iter=None,
       n jobs=1, penalty='12', power t=0.5, random state=None,
       shuffle=True, tol=None, verbose=0, warm start=False)
In [45]:
# Avoid the bais towards the class which has more number of observations
def model.class weight='balanced'
In [46]:
def kfold_crossvalidate(clf, hyper_vals, X_train, y_train, n_splits):
    #ref code:https://scikit-learn.org/stable/modules/generated/sklearn.model selection.KFold.html
    k fold= KFold(n splits=n splits)
    train auc=[]
    cv auc=[]
    for each_val in tqdm(sorted(hyper_vals)):
        clf.alpha=each val
        tr score=0.0
        cv score=0.0
        for train index, test index in k fold.split(X train):
            X tr, X cv = X train[train index], X train[test index]
            y_tr, y_cv = y_train[train_index], y_train[test_index]
            clf=clf.fit(X_tr,y_tr)
            #used CalibratedClassifierCV to get the probabilty scores
            prob clf=CalibratedClassifierCV(clf, method='sigmoid')
            prob_clf.fit(X_tr,y_tr)
            tr score+=roc auc score(y tr,prob clf.predict proba(X tr)[:,1])
            cv_score+=roc_auc_score(y_cv,prob_clf.predict_proba(X_cv)[:,1])
        tr_score=tr_score/n_splits
        cv_score=cv_score/n_splits
```

train_auc.append(tr_score)
cv auc.append(cv score)

```
return [train_auc,cv_auc]
In [47]:
def cross validate plot(clf, hyper vals, norm values, X train, y train, n splits):
    reg scores={}
    for reg val in norm values:
        clf.penalty=reg val
        reg scores[reg val]=kfold crossvalidate(clf,hyper vals,X train,y train,n splits)
    hyper_vals=np.log10(hyper_vals)
    #Code REf:https://matplotlib.org/3.1.1/gallery/subplots axes and figures/subplots demo.html
    fig, plots= plt.subplots(1, 2, figsize=(14, 5), sharey=True)
    fig.suptitle("AUC PLOT for Train and CV datasets".upper())
    for i in range(2):
        reg val='l'+str(i+1)
        plots[i].plot(hyper vals,reg scores[reg val][0],label="TRAIN AUC")
        plots[i].scatter(hyper vals, reg scores[reg val][0])
        plots[i].plot(hyper_vals,reg_scores[reg_val][1],label="CV AUC")
        plots[i].scatter(hyper_vals, reg_scores[reg_val][1])
        plots[i].set_title('L-{} Regularization'.format(i+1))
        plots[i].set(xlabel='ALPHA values', ylabel='AUC')
        plots[i].label outer()
        plots[i].legend()
    plt.show()
In [48]:
def build best model plot roc(model, norm values, X train data, Y train data, X test data, y test data):
    predictions={}
    for reg val,alpha in norm values.items():
        model.penalty=reg_val
        model.alpha=alpha
        model.fit(X train data,y train data)
        prob clf=CalibratedClassifierCV(model, method='sigmoid')
        prob clf.fit(X train data, y train data)
```

```
In [49]:
```

del y_tr_pred
del y_te_pred

y_tr_pred=prob_clf.predict_proba(X_train_data)[:,1]
y_te_pred=prob_clf.predict_proba(X_test_data)[:,1]

predictions[reg_val]=[y_tr_pred,y_te_pred]

plot roc(predictions, y train data, y test data)

```
def plot_roc(predictions,y_train,y_test):
    #Code REf:https://matplotlib.org/3.1.1/gallery/subplots_axes_and_figures/subplots_demo.html
    fig,plots= plt.subplots(1,2,figsize=(14,5),sharey=True)
    fig.suptitle("ROC Curves".upper())
    cutoof_predictions={}

    for i in range(2):
        reg_val='1'+str(i+1)

        y_train_prob=predictions[reg_val][0]
        y_test_prob=predictions[reg_val][1]

        fpr_tr,tpr_tr,thr_tr=roc_curve(y_train,y_train_prob)
        fpr_te,tpr_te,thr_te=roc_curve(y_test,y_test_prob)
```

```
plots[i].plot(fpr tr,tpr tr,label="AUC score for Train data is : {}".format(np.round(auc(fpr tr,tpr tr),4)))
       plots[i].plot(fpr te,tpr te,label="AUC score is Test data is : {}".format(np.round(auc(fpr te,tpr te),4)))
        plots[i].plot([0,1],[0,1],'k--',label="Random Curve AUC score is :{}".format(0.5))
        plots[i].set title('L-{} Regularization'.format(i+1))
        plots[i].set(xlabel='FPR', ylabel='TPR')
        plots[i].legend()
       plots[i].label outer()
        cutoof thr=thr tr[np.argmax(tpr tr*(1-fpr tr))]
        y train pred=predict with best t(y train prob, cutoof thr)
       y test pred=predict with best t(y test prob, cutoof thr)
        cutoof predictions[reg val]=[y train pred,y test pred]
        cutoof predictions[reg val+'max']=np.round(max(tpr tr*(1-fpr tr)),5)
        cutoof predictions[reg val+'cut off']=np.round(cutoof thr,5)
    plot confusion matrix(cutoof predictions, y train, y test)
    display_accuracy(cutoof_predictions,y_train,y_test)
    del y train prob
    del y_test_prob
    del y train pred
    del y test pred
In [50]:
def predict with best t(pred proba, cut off):
    pred= []
    for i in pred_proba:
       if i>=cut off:
            pred.append(1)
        else:
            pred.append(0)
    return pred
def plot confusion matrix(predictions, y train, y test):
```

In [51]:

```
fig = plt.figure(figsize = (14,7))
fig.suptitle("Confusion Matrix for Train and Test datasets")
ax1 = fig.add subplot(221)
ax2 = fig.add subplot(222)
ax3 = fig.add subplot(223)
ax4 = fig.add subplot(224)
ax1.set title("Train data with L-1")
sns.heatmap(confusion_matrix(y_train,predictions['11'][0]),
                        annot=True, fmt="d", cmap="YlGnBu", ax=ax1,)
ax3.set_title("Test data with L-1")
sns.heatmap(confusion matrix(y test,predictions['11'][1]),
                        annot=True, fmt="d", cmap="YlGnBu", ax=ax3)
ax2.set_title("Train data with L-2")
sns.heatmap(confusion matrix(y train, predictions['12'][0]),
                        annot=True, fmt="d", cmap="YlGnBu", ax=ax2)
ax4.set title("Test data with L-2")
sns.heatmap(confusion matrix(y test,predictions['12'][1]),
                        annot=True, fmt="d", cmap="YlGnBu", ax=ax4)
plt.show()
```

In [52]:

```
def display accuracy(cutoof predictions, y train, y test):
    acc table=PrettyTable()
    acc table.field names = ["Regularizer", "Training Accuracy", "Test Accuracy",
                             "Max value of (TPR*(1-FPR))",'Theshold value']
    acc table.hrules=True
```

```
for i in range(2):
        reg_val='l'+str(i+1)
        acc table.add row([reg val+" - Regularizer".upper(),
                           np.round(accuracy_score(y_train,cutoof_predictions[reg_val][0]),5),
                           np.round(accuracy score(y test, cutoof predictions[reg val][1]),5),
                           cutoof predictions[reg val+'max'],
                           cutoof_predictions[reg_val+'cut_off']])
    print(acc table)
In [53]:
def build summary(model, vectorizer, cv method, opt_parm, reg_scores):
    for reg val,alpha in opt parm.items():
        summry.append([model.upper(), (vectorizer+' '+str(reg val)).upper(),
                   cv method.upper(),"Norm : "+str(reg val) +"& alpha :"+ str(alpha),
                   str(reg scores[reg val][0]).upper(),
                   str(reg_scores[reg_val][1]).upper()])
In [54]:
def print summary(summary):
    summary table = PrettyTable()
    summary table.hrules=True
    summary table.field names=['Model', "Vectorizer", "Method for CV", "Opt Parm", "Train AUC", "Test AUC"]
    for each sum in summary:
        summary table.add row(each sum)
    print(summary_table)
```

TASK-1

Bag Of Words

project_title

```
In [55]:
```

```
vectorizer=CountVectorizer(ngram_range=(1,2),min_df=10)
vectorizer.fit(X_train.project_title.values)
X_tr_title=vectorizer.transform(X_train.project_title.values)
X_te_title=vectorizer.transform(X_test.project_title.values)
```

essay

```
In [56]:
```

```
vectorizer=CountVectorizer(ngram_range=(1,2),min_df=10)
vectorizer.fit(X_train.essay.values)
X_tr_essay=vectorizer.transform(X_train.essay.values)
X_te_essay=vectorizer.transform(X_test.essay.values)
```

project_resource_summary

```
In [57]:
```

```
vectorizer=CountVectorizer(ngram_range=(1,2),min_df=10)
vectorizer.fit(X_train.project_resource_summary.values)
X_tr_resource=vectorizer.transform(X_train.project_resource_summary.values)
X_te_resource=vectorizer.transform(X_test.project_resource_summary.values)
```

```
In [58]:
```

```
X_train_bow=hstack((X_tr_vec, X_tr_title, X_tr_resource, X_tr_essay)).tocsr()
X_test_bow=hstack((X_te_vec, X_te_title, X_te_resource, X_te_essay)).tocsr()
```

```
In [59]:

print("Bag of words:")
print("Training data set shape :", X_train_bow.shape)
print("Test data set shape :", X_test_bow.shape)

Bag of words:
Training data set shape : (76473, 163863)

Test data set shape : (32775, 163863)

In [60]:

# Release the memoxy
del X_tr_title
del X_tr_title
del X_tr_title
del X_tr_esource
del X_tr_esource
del X_tr_esource
del X_tr_essay
del X_tr_essay
```

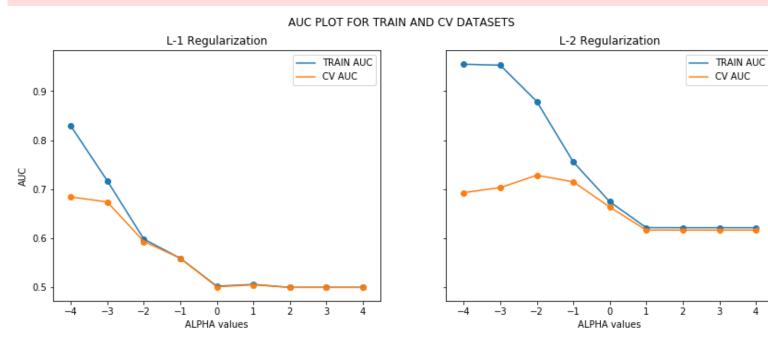
Finding best 'alpha' and build the Model

```
In [61]:
no_folds=7
```

```
In [62]:
```

```
cross_validate_plot(def_model,alpha_vals,norm_values,X_train_bow,y_train,no_folds)

100%| 9/9 [05:13<00:00, 34.81s/it]
100%| 9/9 [03:42<00:00, 24.68s/it]
```

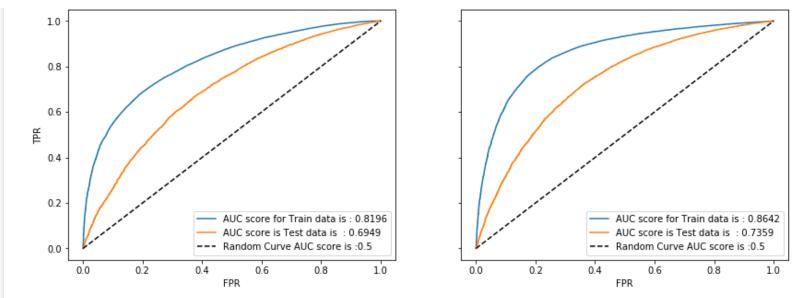


```
In [63]:

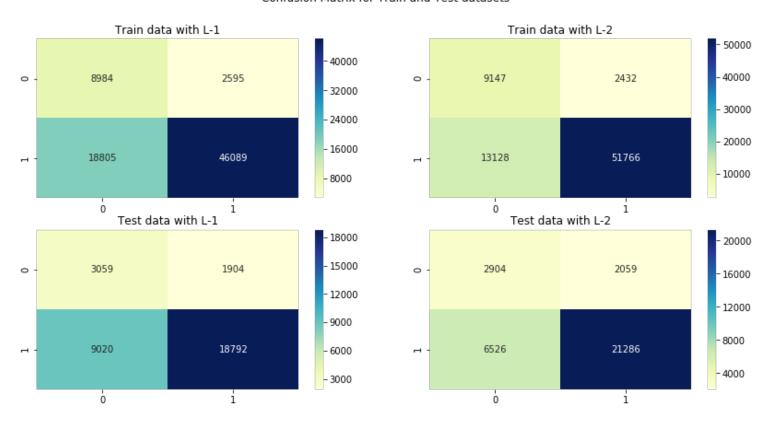
optimal_alpha={}
optimal_alpha['11']=10**-4
optimal_alpha['12']=10**-2
```

In [64]:

build_best_model_plot_roc(def_model,optimal_alpha, X_train_bow, y_train, X_test_bow, y_test)



Confusion Matrix for Train and Test datasets



Regularizer	Training Accuracy	Test Accuracy	Max value of (TPR*(1-FPR))	Theshold value
11 - REGULARIZER	0.72016	0.6667	0.55105	0.83734
12 - REGULARIZER	0.79653	0.73806	0.63016	0.8145

In [65]:

```
del X_train_bow
del X_test_bow
```

In [66]:

```
summry=[]
reg_scores={}
```

In [67]:

```
reg_scores['11']=[.8196,.6949]
reg_scores['12']=[.8642,.7359]
build_summary('SGDClassifier', "Bow", "K-FOLD", optimal_alpha, reg_scores)
```

TF-IDF

project_title

```
In [68]:

tf_idf_vectorizer=TfidfVectorizer(ngram_range=(1,2),min_df=10)

tf_idf_vectorizer.fit(X_train.project_title.values)

X_tr_title=tf_idf_vectorizer.transform(X_train.project_title.values)

X_te_title=tf_idf_vectorizer.transform(X_test.project_title.values)
```

essay

```
In [69]:
```

```
tf_idf_vectorizer=TfidfVectorizer(ngram_range=(1,2),min_df=10)
tf_idf_vectorizer.fit(X_train.essay.values)
X_tr_essay=tf_idf_vectorizer.transform(X_train.essay.values)
X_te_essay=tf_idf_vectorizer.transform(X_test.essay.values)
```

project_resource_summary

```
In [70]:
```

```
tf_idf_vectorizer=TfidfVectorizer(ngram_range=(1,2),min_df=10)

tf_idf_vectorizer.fit(X_train.project_resource_summary.values)

X_tr_resource=tf_idf_vectorizer.transform(X_train.project_resource_summary.values)

X_te_resource=tf_idf_vectorizer.transform(X_test.project_resource_summary.values)
```

In [71]:

```
X_train_tfidf=hstack((X_tr_vec, X_tr_title, X_tr_resource, X_tr_essay)).tocsr()
X_test_tfidf=hstack((X_te_vec, X_te_title, X_te_resource, X_te_essay)).tocsr()
```

In [72]:

```
print("TF-IDF:")
print("Training data set shape :", X_train_tfidf.shape)
print("Test data set shape :", X_test_tfidf.shape)

TF-IDF:
```

Training data set shape : (76473, 163863)
Test data set shape : (32775, 163863)

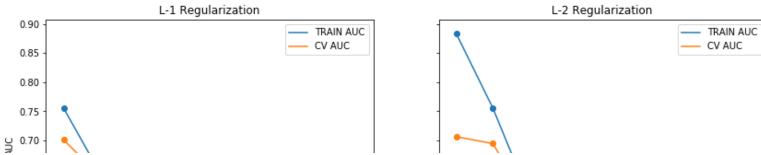
Finding right "Alpha" and build model

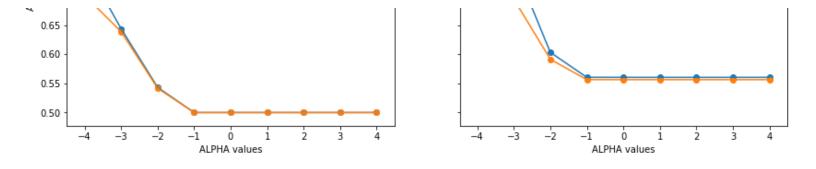
In [73]:

```
cross_validate_plot(def_model,alpha_vals,norm_values,X_train_tfidf,y_train,no_folds)

100%| 9/9 [05:01<00:00, 33.51s/it]
100%| 9/9 [03:45<00:00, 25.01s/it]
```

AUC PLOT FOR TRAIN AND CV DATASETS





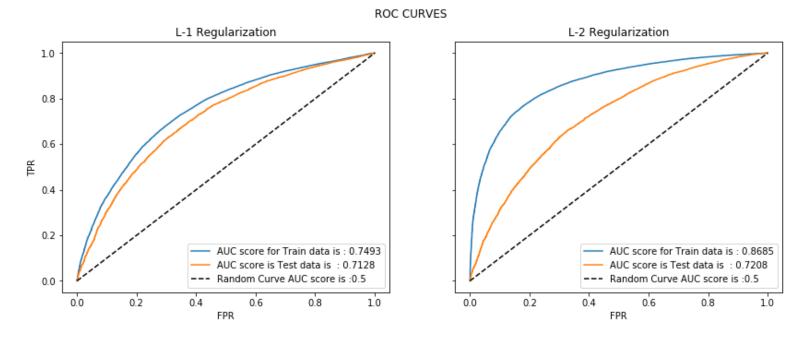
In [74]:

optimal_alpha['11']=10**-4
optimal_alpha['12']=10**-4

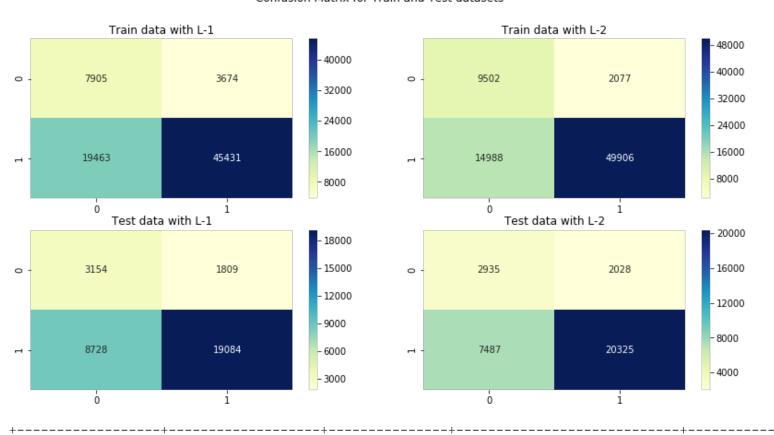
Regularizer

In [75]:

build_best_model_plot_roc(def_model,optimal_alpha,X_train_tfidf,y_train,X_test_tfidf,y_test)



Confusion Matrix for Train and Test datasets

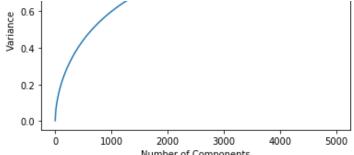


| Training Accuracy | Test Accuracy | Max value of (TPR*(1-FPR)) | Theshold value |

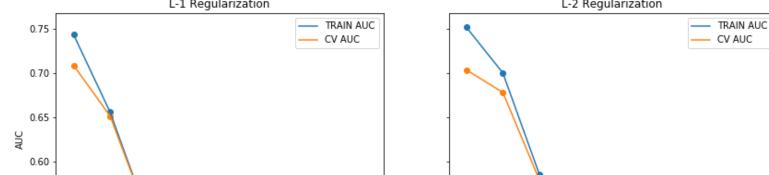
```
| 12 - REGULARIZER | 0.77685 | 0.70969 | 0.63109 | 0.82782
In [76]:
reg scores['11']=[.7493,.7128]
reg scores['12']=[.8685,.7208]
build summary('SGD Classifier', "tf idf", "cross validate", optimal alpha, reg scores)
Task-2(Truncated SVD on TF-IDF Vectorizer)
In [77]:
# Due to memory issues only used 5000 features from dataset
tf idf vectorizer=TfidfVectorizer(ngram range=(1,2),min df=10,max features=5000)
tf idf vectorizer.fit(X train.essay.values)
X tr essay=tf idf vectorizer.transform(X train.essay.values)
X_te_essay=tf_idf_vectorizer.transform(X_test.essay.values)
In [78]:
print("Before Reducing Dimenionatiy :\n")
print("Essay feature shape of Train data:", X tr essay.shape)
print("Essay feature shape of Test data: :",X te essay.shape)
Before Reducing Dimenionatiy:
Essay feature shape of Train data: (76473, 5000)
Essay feature shape of Test data: : (32775, 5000)
In [79]:
# Code ref:https://chrisalbon.com/machine learning/feature engineering/select best number of components in tsvd/
svd = TruncatedSVD(n components=X tr essay.shape[1]-1,random state=103)
svd.fit(X tr essay)
TruncatedSVD(algorithm='randomized', n components=4999, n iter=5,
       random state=103, tol=0.0)
In [80]:
comp to variance={}
cum var ratio=0.0
for comp, cur var in enumerate(svd.explained variance ratio):
    cum var ratio+=cur var
    comp to variance[comp+1]=cum var ratio
del cum var ratio
In [81]:
plt.plot(list(comp to variance.keys()), list(comp to variance.values()), label="% Of Variance preserved")
plt.xlabel("Number of Components")
plt.ylabel("Variance")
plt.legend()
plt.title("SVD")
plt.show()

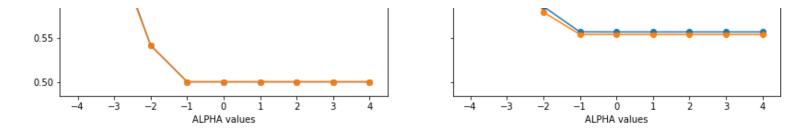
    % Of Variance preserved
```

0.8



```
Number of Components
Decision through Observation:
With the 3200 components we preserved 90% of the information .so we can choose 3200 componets as our newnumber of dimensions instead of 5000 features. we are reduced by 1800 features
and 36% of the featureset
In [82]:
svd = TruncatedSVD(n components=3200, random state=103)
svd.fit(X tr essay)
X_tr_essay=svd.transform(X_tr_essay)
X te essay=svd.transform(X te essay)
In [83]:
print("After Reducing Dimenionatiy :\n")
print("Essay feature shape of Train data:", X tr essay.shape)
print("Essay feature shape of Test data: :",X te essay.shape)
After Reducing Dimenionatiy:
Essay feature shape of Train data: (76473, 3200)
Essay feature shape of Test data: : (32775, 3200)
In [84]:
X train tfidf=hstack((X tr vec, X tr essay)).tocsr()
X test tfidf=hstack((X te vec, X te essay)).tocsr()
In [85]:
print("Truncated SVD:")
print("Training data set shape :", X train tfidf.shape)
print("Test data set shape :", X test tfidf.shape)
Truncated SVD:
Training data set shape: (76473, 3301)
Test data set shape : (32775, 3301)
In [86]:
cross_validate_plot(def_model,alpha_vals,norm_values,X_train_tfidf,y_train,no_folds)
100%|
                 9/9 [57:09<00:00, 381.08s/it]
                 9/9 [26:44<00:00, 178.30s/it]
100%|
                                    AUC PLOT FOR TRAIN AND CV DATASETS
                    L-1 Regularization
                                                                      L-2 Regularization
                                                                                       TRAIN AUC
                                     TRAIN AUC
  0.75
                                     CV AUC
                                                                                       CV AUC
  0.70
```





In [87]:

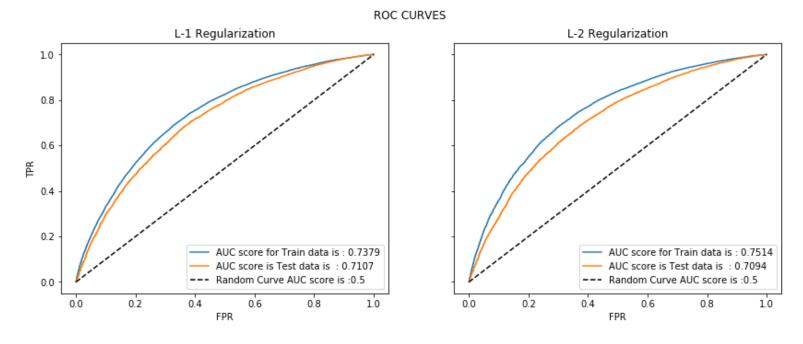
optimal_alpha['11']=10**-4
optimal_alpha['12']=10**-4

| 11 - REGULARIZER |

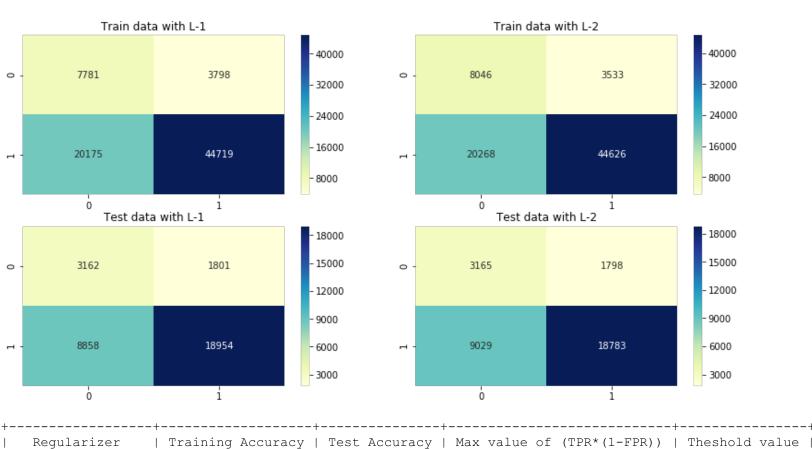
0.68652

In [88]:

build_best_model_plot_roc(def_model,optimal_alpha, X_train_tfidf, y_train, X_test_tfidf, y_test)



Confusion Matrix for Train and Test datasets



0.67478

0.46308

0.83383

```
| 12 - REGULARIZER | 0.68877 | 0.66966 |
                                                             0.47785
In [89]:
reg scores['11']=[.7379,.7107]
reg scores['12']=[.7514,.7094]
build summary('SGDClassifier', "TruncatedSVD(tf-idf)", "KFold", optimal alpha, reg scores)
In [90]:
# Release the memory
del X tr title
del X te title
del X tr resource
del X te resource
del X tr essay
del X te essay
In [91]:
del X train tfidf
del X test tfidf
Avg W2V
In [92]:
# stronging variables into pickle files python:
#http://www.jessicayung.com/how-to-use-pickle-to-save-and-load-variables-in-python/
# make sure you have the glove_vectors file
with open('../glove vectors', 'rb') as f:
   model = pickle.load(f)
   glove words = set(model.keys())
In [93]:
# average Word2Vec
```

```
def avg_w2vec(glove_words, feature_values):
    # compute average word2vec for each review.
    avg_w2v_vec = []; # the avg-w2v for each sentence/review is stored in this list
    for sent in tqdm(feature values): # for each review/sentence
        sent vec = np.zeros(300) # as word vectors are of zero length 300, you might need to
        #change this to 300 if you use google's w2v
        cnt words =0; # num of words with a valid vector in the sentence/review
        for word in sent.split(): # for each word in a review/sentence
            if word in glove words:
                sent vec += model[word]
                cnt words += 1
       if cnt words != 0:
            sent vec /= cnt words
        avg w2v vec.append(sent vec)
   print(len(avg w2v vec))
   print(len(avg_w2v_vec[0]))
   return avg_w2v_vec
```

project_title

```
In [94]:
```

X_tr_title=avg_w2vec(glove_words,X_train.project_title.values)

```
| 76473/76473 [00:01<00:00, 67036.79it/s]
76473
300
In [95]:
X te title=avg w2vec(glove words, X test.project title.values)
              | 32775/32775 [00:00<00:00, 66997.97it/s]
100%|
32775
300
essay
In [96]:
X_tr_essay=avg_w2vec(glove_words,X_train.essay.values)
100%|
               | 76473/76473 [00:19<00:00, 3886.88it/s]
76473
300
In [97]:
X_te_essay=avg_w2vec(glove_words,X_test.essay.values)
               | 32775/32775 [00:08<00:00, 3875.60it/s]
100%|
32775
300
project_resource_summary
In [98]:
X_tr_resource=avg_w2vec(glove_words, X_train.project_resource_summary.values)
100%|
              | 76473/76473 [00:02<00:00, 31671.15it/s]
76473
300
In [99]:
X_te_resource=avg_w2vec(glove_words, X_test.project_resource_summary.values)
              | 32775/32775 [00:01<00:00, 29909.13it/s]
100%|
32775
300
In [100]:
X_train_awv=hstack((X_tr_vec, X_tr_title, X_tr_essay, X_tr_resource)).tocsr()
X_test_awv=hstack((X_te_vec, X_te_title, X_te_essay, X_te_resource)).tocsr()
In [101]:
print("Average Word 2 vector:")
print("Training data set shape :", X train awv.shape)
print("Test data set shape :", X_test_awv.shape)
```

1 0 1

```
In [102]:

# Release the memory
del X_tr_title
del X_tr_title

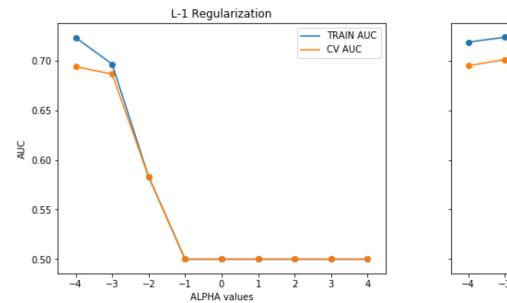
del X_tr_resource
del X_tr_resource
del X_tr_essay
del X_tr_essay

Find the right 'alpha' and build the Classifier

In [103]:

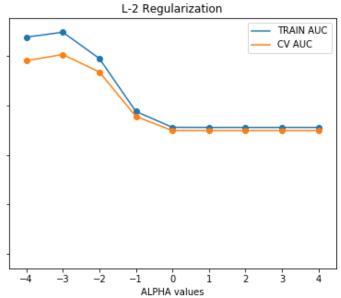
cross_validate_plot(def_model,alpha_vals,norm_values,X_train_awv,y_train,no_folds)
```

AUC PLOT FOR TRAIN AND CV DATASETS



9/9 [15:33<00:00, 103.77s/it]

9/9 [07:21<00:00, 49.02s/it]



In [104]:

100%| 100%|

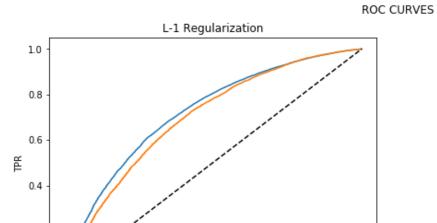
```
optimal_alpha['11']=10**-4
optimal alpha['12']=10**-3
```

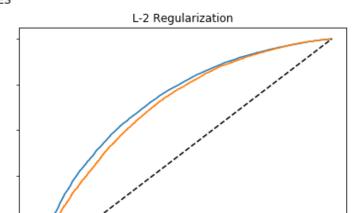
Average word / vector:

Training data set shape : (76473, 1001)
Test data set shape : (32775, 1001)

In [105]:

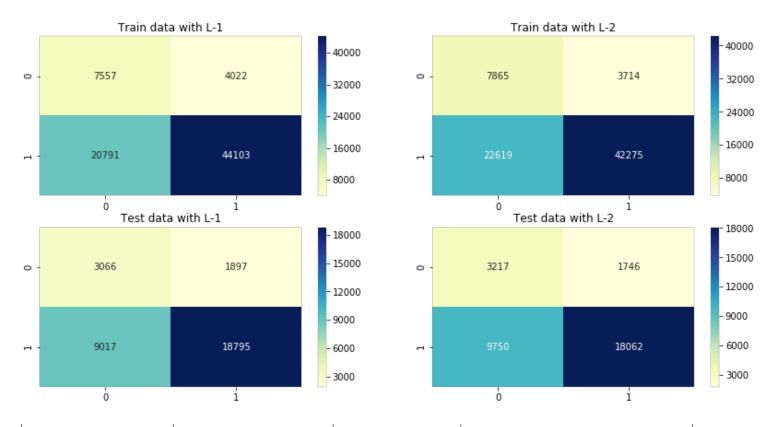
build_best_model_plot_roc(def_model,optimal_alpha, X_train_awv, y_train, X_test_awv, y_test)







Confusion Matrix for Train and Test datasets



Regularizer	Training Accuracy	Test Accuracy	Max value of (TPR*(1-FPR))	Theshold value
11 - REGULARIZER	0.67553	0.667	0.44355	0.8384
12 - REGULARIZER	0.65566	0.64924	0.44249	0.84747

In [106]:

```
reg_scores['11']=[.7217,.6981]
reg_scores['12']=[.7236,.7024]
build_summary('SGDClassifier',"Avg w2v", "K-Fold",optimal_alpha,reg_scores)
```

In [107]:

```
del X_train_awv
del X_test_awv
```

TF-IDF AW2V

In [108]:

essay

In [109]:

```
tfidf model = TfidfVectorizer()
tfidf model.fit(X train.essay.values)
dictionary = dict(zip(tfidf_model.get_feature_names(), list(tfidf_model.idf_)))
tfidf words = set(tfidf model.get feature names())
In [110]:
X_tr_essay= tfidf_avgw2v(glove_words,tfidf_words,X_train.essay.values)
100%|
               | 76473/76473 [02:21<00:00, 540.66it/s]
76473
300
In [111]:
X_te_essay= tfidf_avgw2v(glove_words,tfidf_words,X_test.essay.values)
              | 32775/32775 [00:57<00:00, 569.90it/s]
100%|
32775
300
```

project_title

In [112]:

```
tfidf_model = TfidfVectorizer()
tfidf_model.fit(X_train.project_title.values)
dictionary = dict(zip(tfidf_model.get_feature_names(), list(tfidf_model.idf_)))
tfidf_words = set(tfidf_model.get_feature_names())
```

```
In [113]:

X_tr_title=tfidf_avgw2v(glove_words,tfidf_words,X_train.project_title.values)

100%| 76473/76473 [00:02<00:00, 28417.93it/s]

76473
300</pre>
```

```
In [114]:

X_te_title=tfidf_avgw2v(glove_words,tfidf_words,X_test.project_title.values)

100%| 32775/32775 [00:01<00:00, 29532.03it/s]</pre>
```

32775 300

```
project_resource_summary
In [115]:
tfidf model = TfidfVectorizer()
tfidf model.fit(X train.project resource summary.values)
dictionary = dict(zip(tfidf model.get feature names(), list(tfidf model.idf )))
tfidf words = set(tfidf model.get feature names())
In [116]:
X tr resource=tfidf avgw2v(glove words,tfidf words,X train.project resource summary.values)
              | 76473/76473 [00:07<00:00, 10585.43it/s]
76473
300
In [117]:
X te resource=tfidf avgw2v(glove words,tfidf words,X test.project resource summary.values)
            | 32775/32775 [00:03<00:00, 10453.31it/s]
32775
300
In [118]:
X train tfidfawv=hstack((X tr vec, X tr title, X tr essay, X tr resource)).tocsr()
X test tfidfawv=hstack((X te vec, X te title, X te essay, X te resource)).tocsr()
In [119]:
print("Average Word 2 vector:")
print("Training data set shape :", X train tfidfawv.shape)
print("Test data set shape :", X test tfidfawv.shape)
Average Word 2 vector:
Training data set shape: (76473, 1001)
Test data set shape : (32775, 1001)
In [120]:
# Release the memory
del X tr title
del X te title
del X tr resource
del X te resource
del X tr essay
del X te essay
Find the right 'alpha' and build the Classifier
In [121]:
cross validate plot(def model, alpha vals, norm values, X train tfidfawv, y train, no folds)
100%|
                 9/9 [16:02<00:00, 106.97s/it]
```

100%|

9/9 [07:49<00:00, 52.13s/it]

L-1 Regularization

AUC PLOT FOR TRAIN AND CV DATASETS

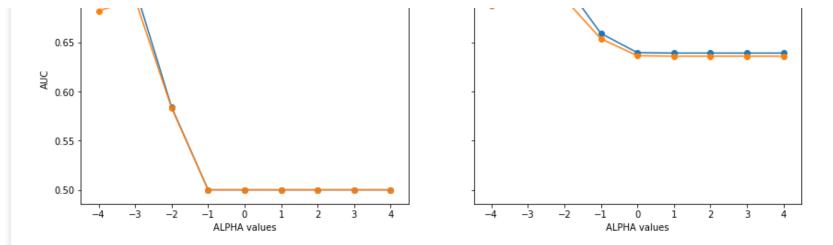
TRAIN AUC

CV AUC

L-2 Regularization

- TRAIN AUC

CV AUC

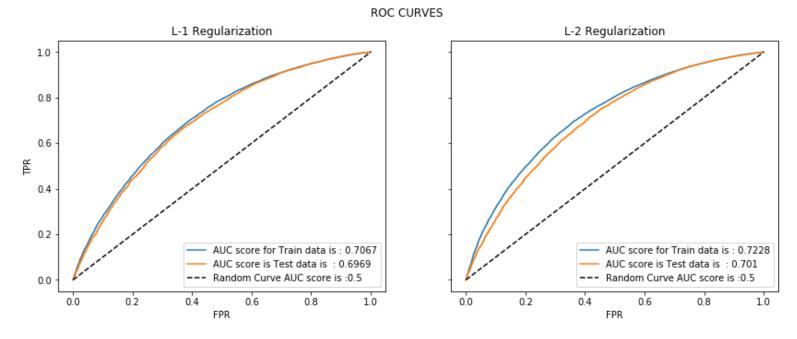


In [122]:

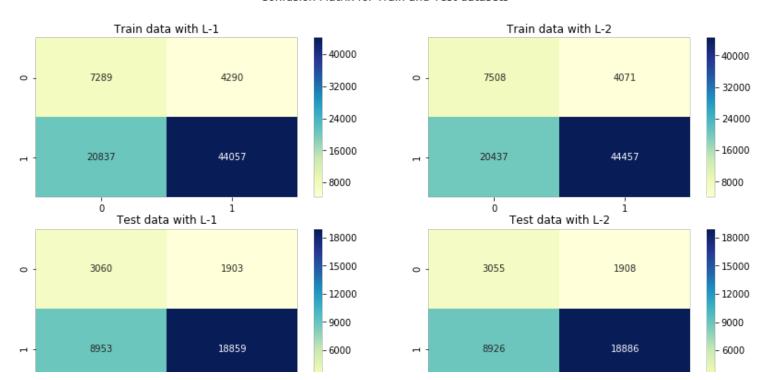
```
optimal_alpha['11']=10**-3
optimal_alpha['12']=10**-3
```

In [123]:

build_best_model_plot_roc(def_model,optimal_alpha, X_train_tfidfawv,y_train, X_test_tfidfawv,y_test)



Confusion Matrix for Train and Test datasets



In [124]:

```
reg_scores['11']=[.7067,.6969]
reg_scores['12']=[.7228,.701]
build_summary('SGDClassifier',"tf_idf avg w2v", "K-Fold",optimal_alpha,reg_scores)
```

In [125]:

```
del X_train_tfidfawv
del X test tfidfawv
```

print summary(summry)

Summary

In [149]:

```
Vectorizer | Method for CV | Opt Parm | Train AUC | Test AUC |
Model |
+-----+
| SGDCLASSIFIER | BOW L2 | K-FOLD | NORM : L2 alpha : 0.01 | 0.8642 | 0.7359 |
| SGDCLASSIFIER | BOW L1 | K-FOLD | NORM : L1& ALPHA :0.0001 | 0.8196 | 0.6949 |
| SGD CLASSIFIER | TF IDF L2 | K-FOLD | NORM : L2& ALPHA :0.0001 | 0.8685 | 0.7208 |
| SGD CLASSIFIER | TF IDF L1 | K-FOLD | NORM : L1& ALPHA :0.0001 | 0.7493 | 0.7128 |
| SGDCLASSIFIER | TRUNCATEDSVD(TF-IDF) L2 | K-FOLD | NORM : L2& ALPHA :0.0001 | 0.7514 | 0.7094 |
+-----+
| SGDCLASSIFIER | TRUNCATEDSVD(TF-IDF) L1 | K-FOLD | NORM : L1& ALPHA :0.0001 | 0.7379 | 0.7107 |
+----+
| SGDCLASSIFIER | AVG W2V L2 | K-FOLD | NORM : L2& ALPHA :0.001 | 0.7236 | 0.7024 |
+----+
| SGDCLASSIFIER | AVG W2V L1 | K-FOLD | NORM : L1& ALPHA :0.0001 | 0.7217 | 0.6981 |
+----+
| SGDCLASSIFIER | TF IDF AVG W2V L2 | K-FOLD | NORM : L2& ALPHA :0.001 | 0.7228 | 0.701 |
+-----+
| SGDCLASSIFIER | TF IDF AVG W2V L1 | K-FOLD | NORM : L1& ALPHA :0.001 | 0.7067 | 0.6969 |
```

In []: