In [1]:

```
%matplotlib inline
import warnings
warnings.filterwarnings("ignore")
import sqlite3
import pandas as pd
import numpy as np
import nltk
import string
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.feature extraction.text import TfidfTransformer
from sklearn.feature extraction.text import TfidfVectorizer
from sklearn.utils import class weight
from sklearn.feature extraction.text import CountVectorizer
from sklearn.metrics import confusion matrix
from sklearn import metrics
from sklearn.metrics import roc curve, auc
from nltk.stem.porter import PorterStemmer
import re
import string
from nltk.corpus import stopwords
from nltk.stem import PorterStemmer
from nltk.stem.wordnet import WordNetLemmatizer
from gensim.models import Word2Vec
from gensim.models import KeyedVectors
import pickle
from tqdm import tqdm
import os
import datetime
import plotly
import plotly.offline as offline
import plotly.graph objs as go
offline.init_notebook_mode()
from collections import Counter
import tensorflow as tf
from tensorflow import keras
from tensorflow.keras.initializers import RandomNormal
from tensorflow.keras.layers import Conv2D, MaxPooling2D
from tensorflow.keras.preprocessing.text import Tokenizer
from tensorflow.keras.preprocessing.sequence import pad sequences
from tensorflow.keras.layers import Input, Embedding, Dense, Flatten, concate
from tensorflow.keras.models import Model, Sequential
from tensorflow.keras.callbacks import EarlyStopping
```

Splitting data into Train and Test

```
In [2]:
```

```
prepeocessed_data = pd.read_csv('preprocessed_data.csv')
prepeocessed_data.head(2)
```

Out[2]:

Unnamed: 0	Unnamed: 0.1	id	teacher_id	teacher
		•	·	

```
0 8393 p205479 2bf07ba08945e5d8b2a3f269b2b3cfe5
```

```
1 37728 p043609 3f60494c61921b3b43ab61bdde2904df
```

2 rows × 21 columns

```
In [3]:
```

```
y = prepeocessed_data['project_is_approved'].values
X = prepeocessed_data.drop(['project_is_approved'], axis=1)
X.shape
```

Out[3]:

(109248, 20)

In [4]:

```
# train test split
from sklearn.model_selection import train_test_split
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, strat_x_test.shape
```

Out[4]:

```
(21850, 20)
```

In [5]:

```
X_train.head(2)
```

Out[5]:

	Unnamed: 0	Unnamed: 0.1	id	teacher_id te
78894	78894	158986	p205323	c49ee844ee9af3432a8ba3df2ff8d4db
38598	38598	21691	p020686	92e0259c01b6953d0848a84e564b584b

In [6]:

y_train

Out[6]:

array([1, 1, 1, ..., 1, 1, 1], dtype=int64)

In [7]:

```
X_test.head(2)
```

Out[7]:

te	teacher_id	id	Unnamed: 0.1	Unnamed: 0	
	23a5aaa55a7ced73d4ec6609f0d1e016	p178007	25121	33848	33848
	1135d7f31075de665a8feeb3bdae3d49	p228886	11992	40483	40483

In [8]:

y_test

Out[8]:

array([1, 1, 1, ..., 1, 1, 0], dtype=int64)

In [9]:

```
def loadGloveModel(gloveFile):
    print ("Loading Glove Model")
    f = open(gloveFile,'r', encoding="utf8")
    model = {}
    for line in tqdm(f):
        splitLine = line.split()
        word = splitLine[0]
        embedding = np.array([float(val) for val in splitLine[1:]])
        model[word] = embedding
    print ("Done.",len(model)," words loaded!")
    return model
model = loadGloveModel('glove.42B.300d.txt')
812it [00:00, 8059.02it/s]
```

```
8121t [00:00, 8059.021t/s]
Loading Glove Model
1917495it [03:30, 9111.60it/s]
Done. 1917495 words loaded!
```

In [10]:

```
t = Tokenizer()
t.fit_on_texts(X_train['essay'].values)
vocab_size = len(t.word_index) + 1
# integer encode the documents
X_train_encoded_docs = t.texts_to_sequences(X_train['essay'].values)
X_test_encoded_docs = t.texts_to_sequences(X_test['essay'].values)
# pad documents to a max Length of 300 words
max length = 300
X train padded docs = pad sequences(X train encoded docs, maxlen=max length,
embedding_matrix = np.zeros((vocab_size, max_length))
for word, i in tqdm(t.word_index.items()):
   embedding_vector = model.get(word)
   if embedding_vector is not None:
       embedding matrix[i] = embedding vector
print(X train padded docs.shape)
print(X test padded docs.shape)
```

```
100%| 61662/61662 [00:00<00:00, 426393.12it/s]
(87398, 300)
```

(21850, 300)

1.3305e-01

```
print(vocab size)
print(embedding_matrix.shape)
print(embedding_matrix[1])
61663
(61663, 300)
[-2.4837e-01 -4.5461e-01 3.9227e-02 -2.8422e-01 -3.1852e-02
2.6355e-01
 -4.6323e+00 1.3890e-02 -5.3928e-01 -8.4454e-02 6.1556e-02 -
4.1552e-01
-1.4599e-01 -5.9321e-01 -2.8738e-02 -3.4991e-02 -2.9698e-01 -
7.9850e-02
  2.7312e-01 2.2040e-01 -8.9859e-02 8.8265e-04 -4.1991e-01 -
1.2536e-01
-5.4629e-02 3.0550e-02 1.9340e-01 -6.3945e-02 2.7405e-02
5.1193e-02
 -3.8656e-01 -1.1085e-01 1.7259e-01 2.9804e-01 -3.5183e-01
1.3150e-01
 -5.4006e-01 -7.6677e-01 -5.5168e-04 1.3076e-01 2.5101e-02
6.2106e-01
 -2.4797e-01 -3.9790e-01 -3.6116e-01 -5.1967e-01 3.0138e-02 -
5.2436e-02
  6.9281e-02 3.5252e-02 -2.1402e-01 2.4836e-01 -1.5693e-01
1.2829e-01
  3.5425e-01 -1.6080e-01 -5.0720e-03 -3.0656e-01 -2.9514e-01 -
1.3554e-01
 -1.4385e-01 -4.0552e-01 5.7233e-01 -2.7670e-01 3.0519e-01
1.5586e-01
  1.6086e-02 -2.2009e-01 4.8589e-01 -4.1384e-01 2.0546e-01
4.0491e-01
  4.1558e-02 -1.3542e-01 2.2544e-01 -2.3629e-01 1.5193e-01 -
1.0859e-02
 -8.2662e-02 -5.5484e-01 -6.1584e-02 -1.1112e-01 -1.1982e-01 -
3.7064e-01
  1.6501e-01 4.4063e-01 -3.3883e-01 -5.7676e-01 5.0847e-01 -
3.5707e-02
-5.9233e-02 3.0748e-02 -2.7689e-01 -7.0433e-02 2.7786e-02 -
5.9336e-01
 -2.8220e+00 -1.0052e-01 6.7168e-01 -1.7046e-01 -2.5902e-01
2.7938e-01
  3.9992e-01 3.7480e-02 -2.6409e-01 -2.6378e-01 2.0645e-01
1.7564e-01
-8.0807e-02 -3.8376e-01 2.6602e-01 3.6214e-01 -9.5112e-02
3.5199e-01
 -8.6994e-01 -1.5747e-01 -2.2550e-01 -6.4948e-02 -2.4845e-01
1.5038e-01
 -3.2951e-01 -2.2285e-01 -2.5509e-02 -2.9725e-01 -3.7715e-01
8.9296e-02
 -3.4399e-02 3.3640e-01 3.5534e-01 3.8253e-01 1.7646e-01
```

```
-3.2743e-01 -4.7115e-01 2.4673e-01 -1.5964e-01 1.8212e-01 -
4.1241e-01
  9.8565e-02 3.8118e-01 3.3043e-01 5.1987e-02 -2.1824e-01
2.2214e-01
 -5.9450e-02 -6.3743e-02 4.3723e-01 1.1068e-01 4.7444e-01
5.6891e-01
  3.1123e-01 -2.0272e-01 8.0078e-02 -4.3905e-01 -1.2246e-01 -
2.5057e-02
 -5.7162e-02 1.4250e-01 9.4468e-02 1.2991e-01 1.0444e-01 -
3.9447e-01
 -2.9337e-01 -2.0466e-01 2.0815e-01 -1.6010e-01 -1.4665e-01
5.4511e-01
  2.9740e-01 -2.2959e-01 -1.7050e-01 -6.2371e-02 -5.0399e-01 -
3.8000e-01
 -3.9528e-01 5.7552e-01 -4.6892e-01 -4.3308e-01 1.5018e-01 -
4.1179e-02
  6.2157e-01 1.9874e-02 -1.1969e-01 -2.5611e-01 2.6602e-01 -
3.7383e-01
  1.2936e-01 -5.0006e-02 -1.1554e-01 -1.7163e-01 -4.2430e-01
1.9844e-01
  5.0611e-01 -1.1093e-01 -1.3939e-01 -5.9377e-01 6.7338e-01
3.8497e-01
  6.2604e-01 -2.0128e-01 3.0058e-01 -1.3946e-01 -1.6186e-01
1.2168e-01
 -1.8410e-02 6.1356e-01 -1.9887e-01 1.9250e-01 8.4372e-03 -
5.0757e-01
  3.5858e-01 -4.9729e-01 -4.4725e-01 2.1423e-02 -2.0769e-01
8.3729e-02
  2.2032e-01 1.4404e-01 1.2590e-03 -4.4309e-01 -1.7242e-01 -
3.5300e-01
 -2.9477e-01 3.2898e-01 -3.1910e+00 3.8910e-01 3.5654e-01
5.2134e-02
  2.0576e-01 -8.8649e-02 1.6398e-01 1.1203e-01 2.8590e-01
2.8940e-01
 -4.4349e-01 9.1036e-01 -3.0902e-01 -1.3985e-01 -3.9499e-01 -
2.7299e-02
 -1.5201e-01 8.4418e-02 -3.7196e-01 4.9827e-02 1.4128e-01 -
1.5126e-01
 -1.6107e-01 4.0226e-03 1.6799e-01 -2.5429e-01 -1.5074e-01 -
5.7409e-01
 -1.5611e-01 6.8407e-02 2.4832e-01 1.6828e-01 7.2764e-02 -
8.6728e-02
  2.1982e-03 1.3593e-01 7.0224e-01 -4.5976e-01 -2.4506e-01 -
3.3874e-01
 -1.0952e-01 2.4698e-01 -5.5919e-01 -3.8866e-01 -1.3372e-01
9.1943e-02
 -1.0543e-01 -3.1319e-01 -2.9952e-01 -2.0611e-01 1.7976e-01
4.5800e-01
 -7.2402e-02 1.6118e-01 -4.1649e-01 -3.0103e-01 2.3234e-01 -
5.0139e-02
  1.0026e-01 3.8974e-01 -6.1342e-02 2.6626e-01 -1.5671e-01
7.5136e-02
 -4.2926e-01 -1.2025e-01 8.2736e-02 -6.2469e-01 4.4267e-02
```

```
6.0673e-01
-1.2458e-01 -1.5443e-01 -1.6339e-01 5.3097e-02 1.5458e-01 -
3.8053e-01]
```

1.4 Encoding Categorical and Numerical features

In [15]:

```
from sklearn.preprocessing import LabelEncoder
import numpy as np
class LabelEncoderExt(object):
   def __init__(self):
       It differs from LabelEncoder by handling new classes and providing a
       Unknown will be added in fit and transform will take care of new item
        self.label_encoder = LabelEncoder()
       # self.classes = self.label encoder.classes
   def fit(self, data_list):
       This will fit the encoder for all the unique values and introduce unk
        :param data list: A list of string
        :return: self
        self.label_encoder = self.label_encoder.fit(list(data list) + ['Unknown]
        self.classes = self.label encoder.classes
        return self
   def transform(self, data_list):
       This will transform the data list to id list where the new values get
        :param data_list:
        :return:
       new_data_list = list(data_list)
        for unique item in np.unique(data list):
            if unique item not in self.label encoder.classes :
                new data list = ['Unknown' if x==unique item else x for x in
        return self.label encoder.transform(new data list)
```

1.4.1 encoding categorical features: clean_categories

In [16]:

```
from sklearn.preprocessing import LabelEncoder
vectorizer_cat = LabelEncoderExt()
vectorizer_cat.fit(X_train['clean_categories'].values) # fit has to happen or
X_train_cc_ohe = vectorizer_cat.transform(X_train['clean_categories'].values)
X_test_cc_ohe = vectorizer_cat.transform(X_test['clean_categories'].values)
print("After vectorizations")
print(X_train_cc_ohe.shape, y_train.shape)
print(X_test_cc_ohe.shape, y_test.shape)
After vectorizations
```

```
After vectorizations (87398,) (87398,) (21850,)
```

1.4.2 encoding categorical features: clean_subcategories

In [19]:

```
vectorizer_subcat = LabelEncoderExt()
vectorizer_subcat.fit(X_train['clean_subcategories'].values) # fit has to hap
X_train_csc_ohe = vectorizer_subcat.transform(X_train['clean_subcategories'].
X_test_csc_ohe = vectorizer_subcat.transform(X_test['clean_subcategories'].values)
print("After vectorizations")
print(X_train_csc_ohe.shape, y_train.shape)
print(X_test_csc_ohe.shape, y_test.shape)
```

```
After vectorizations (87398,) (87398,) (21850,) (21850,)
```

1.4.3 encoding categorical features: school_state

In [20]:

```
vectorizer_school_state = LabelEncoderExt()
vectorizer_school_state.fit(X_train['school_state'].values)

X_train_state_ohe = vectorizer_school_state.transform(X_train['school_state']
X_test_state_ohe = vectorizer_school_state.transform(X_test['school_state'].v

print("After vectorizations")
print(X_train_state_ohe.shape, y_train.shape)
print(X_test_state_ohe.shape, y_test.shape)

After vectorizations
```

```
After vectorizations (87398,) (87398,) (21850,) (21850,)
```

1.4.4 encoding categorical features: teacher_prefix

In [21]:

```
vectorizer_prefix = LabelEncoderExt()
vectorizer_prefix.fit(X_train['teacher_prefix'].values)

X_train_teacher_ohe = vectorizer_prefix.transform(X_train['teacher_prefix'].values)

X_test_teacher_ohe = vectorizer_prefix.transform(X_train['teacher_prefix'].values)

print("After vectorizations")
print(X_train_teacher_ohe.shape, y_train.shape)
print(X_test_teacher_ohe.shape, y_test.shape)
```

```
After vectorizations (87398,) (87398,) (21850,) (21850,)
```

1.4.5 encoding categorical features: project_grade_category

```
In [22]:
```

```
vectorizer_grade = LabelEncoderExt()
vectorizer_grade.fit(X_train['project_grade_category'].values)

X_train_grade_ohe = vectorizer_grade.transform(X_train['project_grade_categor X_test_grade_ohe = vectorizer_grade.transform(X_test['project_grade_category'])
print("After vectorizations")
print(X_train_grade_ohe.shape, y_train.shape)
print(X_test_grade_ohe.shape, y_test.shape)

After vectorizations
(87398,) (87398,)
```

1.4.6 encoding numerical features: price

In [24]:

(21850,) (21850,)

```
from sklearn.preprocessing import MinMaxScaler

scaler = MinMaxScaler()
scaler.fit(X_train['price'].values.reshape(-1,1))
X_train_price=scaler.transform(X_train['price'].values.reshape(-1,1))
X_test_price=scaler.transform(X_test['price'].values.reshape(-1,1))

print(X_train_price.shape)
print(X_test_price.shape)
print(X_train_price)
print(X_test_price)
```

```
(87398, 1)

(21850, 1)

[[0.00684614]

[0.05806564]

[0.02537021]

...

[0.02702549]

[0.04971725]

[0.04152389]]

[[0.00693015]

[0.1253258]

[0.1253258]

[0.02181162]

...

[0.07185593]

[0.02632337]

[0.01072578]]
```

1.4.7 encoding numerical features: teacher_number_of_previously_posted_projects

In [25]:

```
scaler = MinMaxScaler()
scaler.fit(X_train['teacher_number_of_previously_posted_projects'].values.res
X_train_pre_proj=scaler.transform(X_train['teacher_number_of_previously_posted
X_test_pre_proj=scaler.transform(X_test['teacher_number_of_previously_posted]

print(X_train_pre_proj.shape)
print(X_test_pre_proj.shape)

(87398, 1)
(21850, 1)
```

1.4.8 encoding numerical features: quantity

In [26]:

```
scaler = MinMaxScaler()
scaler.fit(X_train['quantity'].values.reshape(-1,1))
X_train_quantity=scaler.transform(X_train['quantity'].values.reshape(-1,1))
X_test_quantity=scaler.transform(X_test['quantity'].values.reshape(-1,1))

print(X_train_quantity.shape)

print(X_test_quantity.shape)

(87398, 1)
(21850, 1)
```

In [27]:

```
X_train_numerals=np.concatenate((X_train_pre_proj,X_train_quantity,X_train_pr
X_test_numerals=np.concatenate((X_test_pre_proj,X_test_quantity,X_test_price)
print(X_train_numerals.shape)
print(X_test_numerals.shape)
```

```
(87398, 3)
(21850, 3)
```

```
In [28]:
```

```
print('vocabulary size is ',vocab_size)
```

vocabulary size is 61663

```
In [1]:
```

```
These cells are to load the data if the system crashes during training
import warnings
warnings.filterwarnings("ignore")
import sqlite3
import pandas as pd
import numpy as np
import nltk
import string
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.feature extraction.text import TfidfTransformer
from sklearn.feature extraction.text import TfidfVectorizer
from sklearn.utils import class weight
from sklearn.feature extraction.text import CountVectorizer
from sklearn.metrics import confusion matrix
from sklearn import metrics
from sklearn.metrics import roc curve, auc
from nltk.stem.porter import PorterStemmer
import re
import string
from nltk.corpus import stopwords
from nltk.stem import PorterStemmer
from nltk.stem.wordnet import WordNetLemmatizer
from gensim.models import Word2Vec
from gensim.models import KeyedVectors
import pickle
from tadm import tadm
import os
import datetime
import plotly
import plotly.offline as offline
import plotly.graph objs as go
offline.init notebook mode()
from collections import Counter
import tensorflow as tf
from tensorflow import keras
from tensorflow.keras.initializers import RandomNormal
from tensorflow.keras.layers import Conv2D, MaxPooling2D
from tensorflow.keras.preprocessing.text import Tokenizer
from tensorflow.keras.preprocessing.sequence import pad sequences
from tensorflow.keras.layers import Input, Embedding, Dense, Flatten, concate
from tensorflow.keras.models import Model, Sequential
```

```
from tensorflow.keras.callbacks import EarlyStopping, ModelCheckpoint, Reduce
vocab_size = 61663
```

In [29]:

```
#Saving processed data
from sklearn.externals import joblib
joblib.dump(embedding matrix, 'embedding matrix.pkl')
joblib.dump(X_train_padded_docs, 'X_train_padded_docs.pkl')
joblib.dump(X_train_cc_ohe, 'X_train_cc_ohe.pkl')
joblib.dump(X train csc ohe, 'X train csc ohe.pkl')
joblib.dump(X_train_teacher_ohe, 'X_train_teacher_ohe.pkl')
joblib.dump(X_train_grade_ohe, 'X_train_grade_ohe.pkl')
joblib.dump(X_train_state_ohe, 'X_train_state_ohe.pkl')
joblib.dump(X_train_numerals, 'X_train_numerals.pkl')
joblib.dump(y_train, 'y_train.pkl')
joblib.dump(X_test_padded_docs, 'X_test_padded_docs.pkl')
joblib.dump(X_test_cc_ohe, 'X_test_cc_ohe.pkl')
joblib.dump(X_test_csc_ohe, 'X_test_csc_ohe.pkl')
joblib.dump(X_test_teacher_ohe, 'X_test_teacher_ohe.pkl')
joblib.dump(X_test_grade_ohe, 'X_test_grade_ohe.pkl')
joblib.dump(X_test_state_ohe, 'X_test_state_ohe.pkl')
joblib.dump(X test numerals, 'X test numerals.pkl')
joblib.dump(y_test, 'y_test.pkl')
```

Out[29]:

```
['y_test.pkl']
```

In [2]:

```
#Loading processed data
from sklearn.externals import joblib
embedding matrix = joblib.load('embedding matrix.pkl')
X_train_padded_docs = joblib.load('X_train_padded_docs.pkl')
X train cc ohe = joblib.load('X train cc ohe.pkl')
X train csc ohe = joblib.load('X train csc ohe.pkl')
X_train_teacher_ohe = joblib.load('X_train_teacher_ohe.pkl')
X train grade ohe = joblib.load('X train grade ohe.pkl')
X_train_state_ohe = joblib.load('X_train_state_ohe.pkl')
X_train_numerals = joblib.load('X_train_numerals.pkl')
y train = joblib.load('y train.pkl')
X_test_padded_docs = joblib.load('X_test_padded_docs.pkl')
X_test_cc_ohe = joblib.load('X_test_cc_ohe.pkl')
X test csc ohe = joblib.load('X test csc ohe.pkl')
X_test_teacher_ohe = joblib.load('X_test_teacher_ohe.pkl')
X_test_grade_ohe = joblib.load('X_test_grade_ohe.pkl')
X test state ohe = joblib.load('X test state ohe.pkl')
X_test_numerals = joblib.load('X_test_numerals.pkl')
y test = joblib.load('y test.pkl')
```

In [3]:

Model's architecture

```
# Essay Layers
essay_input = Input(shape=(len(X_train_padded_docs[0]),), name='essay input')
essay input 1 = Embedding(input dim=vocab size,output dim=300, input length=1
                              weights=[embedding matrix], trainable=False)(es
essay input 1 = CuDNNLSTM(units = 64,
                              kernel initializer= 'he normal',
                              return_sequences=True)(essay_input_1)
essay input 1 = Flatten()(essay_input_1)
# Category Layers
categories input = Input(shape=(1,), name='categories input')
categories_input_1= Embedding(input_dim=len(set(X_train_cc_ohe)), output_dim
categories_input_1 = Flatten()(categories_input 1)
# Sub Category Layers
sub_categories_input = Input(shape=(1,), name='sub_categories_input')
sub categories input 1 = Embedding(input dim=len(set(X train csc ohe)), output
sub categories input 1 = Flatten()(sub categories input 1)
# Grade Layers
proj grade input = Input(shape=(1,), name='proj grade input')
proj grade input 1 = Embedding(input dim=len(set(X train grade ohe)), output
proj_grade_input_1 = Flatten()(proj_grade_input_1)
# School Layers
school_state_input = Input(shape=(1,), name='school_state_input')
school state input 1 = Embedding(input dim=len(set(X train state ohe)), output
school state input 1 = Flatten()(school state input 1)
# Teacher Prefix Layers
tch_input = Input(shape=(1,), name='tch_input')
tch input 1= Embedding(input dim=len(set(X train teacher ohe)), output dim =
tch_input_1 = Flatten()(tch_input_1)
# Numerical Layers
numeral_input = Input(shape=(X_train_numerals.shape[1],),name='numeral_input
numeral input 1 = Dense(units = 64,
                            activation = 'relu',
                            kernel_initializer = 'he_normal')(numeral_input)
# Cconcatinating all the above Layers
x = concatenate([essay_input_1, categories_input_1, sub_categories_input_1,
```

```
proj_grade_input_1, school_state_input_1,
                     tch input 1, numeral input 1])
# Dense Layers
output = Dense(units = 624,
                  activation = 'relu',
                            kernel initializer = 'he normal')(x)
output = Dropout(0.7)(output)
output = BatchNormalization()(output)
output = Dropout(0.7)(output)
output = BatchNormalization()(output)
output = Dense(units = 512,
                  activation = 'relu',
                            kernel initializer = 'he normal')(output)
output = Dropout(0.8)(output)
output = BatchNormalization()(output)
output = Dense(units = 512,
                  activation = 'relu',
                            kernel initializer = 'he normal')(output)
output = Dropout(0.45)(output)
output =Dense(1, activation='sigmoid', name='output')(output)
model 1 = Model(inputs = [essay input, categories input, sub categories input
                            proj_grade_input, school_state_input, tch input,
                  outputs=output)
model_1.compile(loss = 'binary_crossentropy',
                  optimizer = 'adam',
                  metrics = ['accuracy', auc])
WARNING:tensorflow:From c:\users\addu\appdata\local\program
```

s\python\python37\lib\site-packages\tensorflow\python\keras \initializers.py:119: calling RandomUniform. init (from tensorflow.python.ops.init ops) with dtype is deprecated an d will be removed in a future version. Instructions for updating: Call initializer instance with the dtype argument instead o f passing it to the constructor WARNING:tensorflow:Large dropout rate: 0.7 (>0.5). In Tenso rFlow 2.x, dropout() uses dropout rate instead of keep_pro b. Please ensure that this is intended. WARNING:tensorflow:Large dropout rate: 0.7 (>0.5). In Tenso rFlow 2.x, dropout() uses dropout rate instead of keep pro b. Please ensure that this is intended. WARNING:tensorflow:Large dropout rate: 0.8 (>0.5). In Tenso rFlow 2.x, dropout() uses dropout rate instead of keep pro b. Please ensure that this is intended. WARNING:tensorflow:From c:\users\addu\appdata\local\program s\python\python37\lib\site-packages\tensorflow\python\ops\i

Training the model with train data

In [5]:

```
log_dir="logs\\model_1_log" + datetime.datetime.now().strftime("%Y%m%d-%H%M%S
tensorboard = TensorBoard(log_dir=log_dir)
mcp_save = ModelCheckpoint('.model1_best_weights.hdf5', save_best_only=True,
callbacks_list = [mcp_save,tensorboard]
```

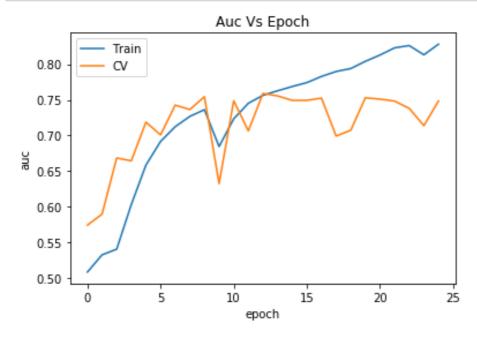
In [6]:

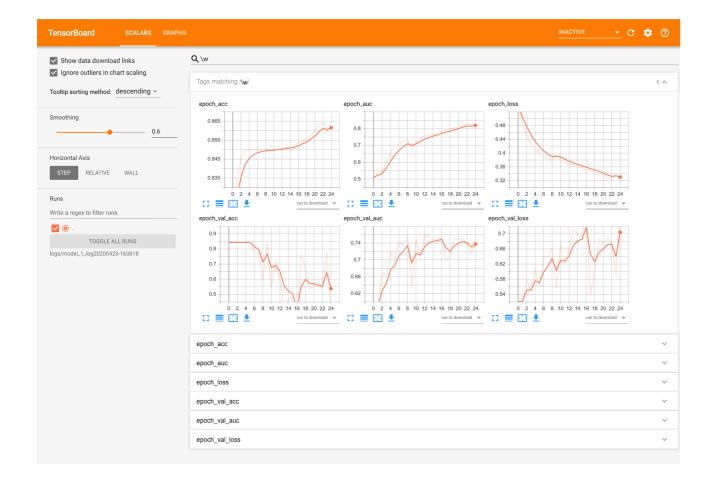
```
Epoch 11/25
69918/69918 - 4s - loss: 0.3842 - acc: 0.8500 - auc: 0.7233
- val loss: 0.6702 - val acc: 0.5412 - val auc: 0.7487
Epoch 12/25
69918/69918 - 4s - loss: 0.3735 - acc: 0.8503 - auc: 0.7451
- val_loss: 0.6248 - val_acc: 0.7154 - val_auc: 0.7064
Epoch 13/25
69918/69918 - 5s - loss: 0.3670 - acc: 0.8506 - auc: 0.7563
- val loss: 0.6633 - val acc: 0.5947 - val auc: 0.7590
Epoch 14/25
69918/69918 - 4s - loss: 0.3643 - acc: 0.8513 - auc: 0.7627
- val_loss: 0.7102 - val_acc: 0.4108 - val_auc: 0.7555
Epoch 15/25
69918/69918 - 4s - loss: 0.3609 - acc: 0.8512 - auc: 0.7687
- val_loss: 0.6991 - val_acc: 0.4696 - val_auc: 0.7492
Epoch 16/25
69918/69918 - 4s - loss: 0.3574 - acc: 0.8517 - auc: 0.7741
- val_loss: 0.6925 - val_acc: 0.4758 - val_auc: 0.7492
Fnoch 17/25
```

Auc Vs Epoch

In [7]:

```
# summarize history for accuracy
plt.plot(history.history['auc'])
plt.plot(history.history['val_auc'])
plt.title('Auc Vs Epoch')
plt.ylabel('auc')
plt.xlabel('epoch')
plt.legend(['Train', 'CV'], loc='right bottom')
plt.show()
```





Testing the model with test data

```
# Essay Layers
essay_input = Input(shape=(len(X_train_padded_docs[0]),), name='essay input')
essay input 1 = Embedding(input dim=vocab size,output dim=300, input length=1
                              weights=[embedding matrix], trainable=False)(es
essay input 1 = CuDNNLSTM(units = 64,
                              kernel initializer= 'he normal',
                              return_sequences=True)(essay_input_1)
essay input 1 = Flatten()(essay input 1)
# Category Layers
categories input = Input(shape=(1,), name='categories input')
categories_input_1= Embedding(input_dim=len(set(X_train_cc_ohe)), output_dim
categories_input_1 = Flatten()(categories_input 1)
# Sub Category Layers
sub_categories_input = Input(shape=(1,), name='sub_categories_input')
sub categories input 1 = Embedding(input dim=len(set(X train csc ohe)), output
sub categories input 1 = Flatten()(sub categories input 1)
# Grade Layers
proj grade input = Input(shape=(1,), name='proj grade input')
proj grade input 1 = Embedding(input dim=len(set(X train grade ohe)), output
proj_grade_input_1 = Flatten()(proj_grade_input_1)
# School Layers
school_state_input = Input(shape=(1,), name='school_state_input')
school state input 1 = Embedding(input dim=len(set(X train state ohe)), output
school state input 1 = Flatten()(school state input 1)
# Teacher Prefix Layers
tch_input = Input(shape=(1,), name='tch_input')
tch input 1= Embedding(input dim=len(set(X train teacher ohe)), output dim =
tch_input_1 = Flatten()(tch_input_1)
# Numerical Layers
numeral_input = Input(shape=(X_train_numerals.shape[1],),name='numeral_input
numeral input 1 = Dense(units = 64,
                            activation = 'relu',
                            kernel_initializer = 'he_normal')(numeral_input)
# Cconcatinating all the above Layers
x = concatenate([essay_input_1, categories_input_1, sub_categories_input_1,
```

```
proj_grade_input_1, school_state_input_1,
                     tch input 1, numeral input 1])
# Dense Layers
output = Dense(units = 624,
                  activation = 'relu',
                            kernel_initializer = 'he_normal')(x)
output = Dropout(0.7)(output)
output = BatchNormalization()(output)
output = Dropout(0.7)(output)
output = BatchNormalization()(output)
output = Dense(units = 512,
                  activation = 'relu',
                            kernel initializer = 'he normal')(output)
output = Dropout(0.8)(output)
output = BatchNormalization()(output)
output = Dense(units = 512,
                  activation = 'relu',
                            kernel initializer = 'he normal')(output)
output = Dropout(0.45)(output)
output =Dense(1, activation='sigmoid', name='output')(output)
model 1 = Model(inputs = [essay input, categories input, sub categories input
                            proj_grade_input, school_state_input, tch input,
                  outputs=output)
model_1.compile(loss = 'binary_crossentropy',
                  optimizer = 'adam',
                  metrics = ['accuracy', auc])
```

WARNING:tensorflow:Large dropout rate: 0.7 (>0.5). In TensorFl ow 2.x, dropout() uses dropout rate instead of keep_prob. Plea se ensure that this is intended.

WARNING:tensorflow:Large dropout rate: 0.7 (>0.5). In TensorFl ow 2.x, dropout() uses dropout rate instead of keep_prob. Plea se ensure that this is intended.

In [9]:

```
#loading the best weights obtained from traing model on train data
model_1.load_weights(".model1_best_weights.hdf5")
```

In [10]:

Auc score of model on unseen Test data

```
In [11]:
```

```
print('Test Auc obtained is ',score)
```

Test Auc obtained is 0.75348437

In [12]:

model_1.summary()		
Model: "model_1"		
Layer (type) Connected to	Output Shape	Param #
essay_input (InputLayer)		0
embedding_6 (Embedding) essay_input[0][0]	(None, 300, 300)	18498900
categories_input (InputLayer)	[(None, 1)]	0
sub_categories_input (InputLaye	[(None, 1)]	0
proj_grade_input (InputLayer)	[(None, 1)]	0
school_state_input (InputLayer)	[(None, 1)]	0
tch_input (InputLayer)	[(None, 1)]	0
<pre>cu_dnnlstm_1 (CuDNNLSTM) embedding_6[0][0]</pre>	(None, 300, 64)	93696
embedding_7 (Embedding) categories_input[0][0]	(None, 1, 16)	816
<pre>embedding_8 (Embedding) sub_categories_input[0][0]</pre>	(None, 1, 64)	25216
embedding_9 (Embedding) proj_grade_input[0][0]	(None, 1, 64)	256
embedding_10 (Embedding) school_state_input[0][0]	(None, 1, 64)	3264

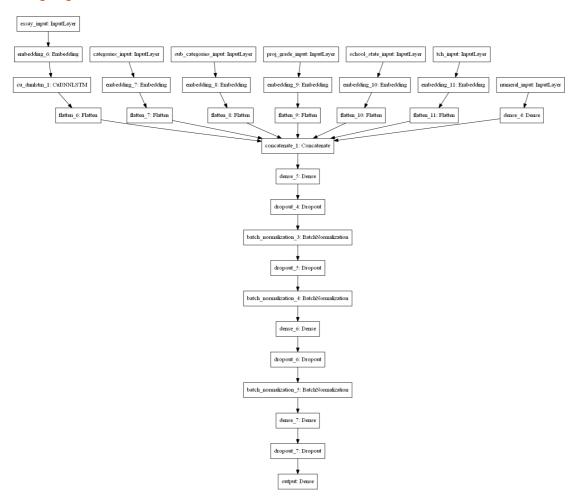
embedding_11 (Embedding) tch_input[0][0]	(None,	1, 64)	320
numeral_input (InputLayer)	[(None	, 3)]	0
flatten_6 (Flatten) cu_dnnlstm_1[0][0]	(None,	19200)	0
flatten_7 (Flatten) embedding_7[0][0]	(None,	16)	0
flatten_8 (Flatten) embedding_8[0][0]	(None,	64)	0
flatten_9 (Flatten) embedding_9[0][0]	(None,	64)	0
flatten_10 (Flatten) embedding_10[0][0]	(None,	64)	0
flatten_11 (Flatten) embedding_11[0][0]	(None,	64)	0
dense_4 (Dense) numeral_input[0][0]	(None,	64)	256
<pre>concatenate_1 (Concatenate) flatten_6[0][0]</pre>	(None,	19536)	0
flatten_7[0][0]			
flatten_8[0][0]			
flatten_9[0][0]			
flatten_10[0][0]			
flatten_11[0][0]			
dense_4[0][0]			
dense_5 (Dense)	(None,	624)	12191088

dropout_4 (Dropout) dense_5[0][0]	(None,	624)	0
batch_normalization_3 (BatchNor dropout_4[0][0]	(None,	624)	2496
dropout_5 (Dropout) batch_normalization_3[0][0]	(None,	624)	0
batch_normalization_4 (BatchNor dropout_5[0][0]	(None,	624)	2496
dense_6 (Dense) batch_normalization_4[0][0]	(None,	512)	320000
dropout_6 (Dropout) dense_6[0][0]	(None,	512)	0
batch_normalization_5 (BatchNor dropout_6[0][0]	(None,	512)	2048
dense_7 (Dense) batch_normalization_5[0][0]	(None,	512)	262656
dropout_7 (Dropout) dense_7[0][0]	(None,	512)	0
output (Dense) dropout_7[0][0]	(None,	·	513
Total params: 31,404,021 Trainable params: 12,901,601 Non-trainable params: 18,502,420	====		

In [13]:

```
#drawing models
tf.keras.utils.plot_model(
    model_1,
    show_shapes=False,
    show_layer_names=True,
    rankdir='TB'
)
```

Out[13]:



Summary

- Created tensorflow model using text, categorical and numerical layers.
- Performed hyper parameter tuning manually on number of layers, activation functions and optimizers.
- Trained the network using the best obtained hyper parameters.
- Tested the model using Test data and obtained test aucroc score of 0.7534.
- Printed the summary of the model along with it's image.