Importing Libraries

```
In [1]:
```

```
import pandas as pd
import numpy as np
from time import time
from tensorflow.python.keras.callbacks import TensorBoard
import re

#from google.colab import drive
#drive.mount('/content/gdrive')
# Dataset is now stored in a Pandas Dataframe

C:\ProgramData\Anaconda3\lib\site-packages\h5py\_init__.py:36: FutureWarning: Conversion of the s
econd argument of issubdtype from `float` to `np.floating` is deprecated. In future, it will be
treated as `np.float64 == np.dtype(float).type`.
    from ._conv import register_converters as _register_converters
```

Importing Data

```
In [2]:
```

```
data=pd.read_csv('DonorsChoose_LSTM.csv')
```

In [60]:

```
data[101926:101927]
```

Out[60]:

	teacher_prefix	school_state	project_grade_category	teacher_number_of_previously_posted_projects	project_is_
101926	Ms.	FL	Grades 9-12	2	1
41		·	188		

In [11]:

```
data['total_text']=data['clean_essay'].map(str) + data['clean_title'].map(str)
```

In [12]:

```
from sklearn.model_selection import train_test_split
from sklearn.metrics import roc_auc_score
train, test = train_test_split(data, random_state=123, shuffle=True, test_size=0.1)
print("Training data shape:", train.shape)
print("Test data shape:", test.shape)
```

Training data shape: (98323, 13) Test data shape: (10925, 13)

In [13]:

```
train=train.reset_index()
test=test.reset_index()
```

In [14]:

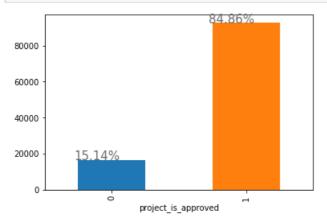
```
train.head(2)
```

Out[14]:

	index	teacher_prefix	school_state	project_grade_category	teacher_number_of_previously_posted_projects	project_
0	101926	Ms.	FL	Grades 9-12	2	1
1	83948	Mrs.	CA	Grades 3-5	0	1

Distribution of positive and negative data points

```
In [65]:
```



Importing pre trained Glove vectors

```
In [3]:
```

```
import io
import pickle
with io.open('glove_vectors.txt', 'rb') as f:
    glove_model = pickle.load(f)
    glove_words = set(glove_model.keys())
```

Preprocessing with Keras tokenizer

token.fit on texts(train['clean subcategories'])

subcategory_size = len(token.word_index) + 1

```
In [8]:
```

```
from keras.preprocessing.text import Tokenizer

Using TensorFlow backend.
```

In [9]: token = Tokenizer() token.fit_on_texts(train['total_text']) train['total text']=token.texts to sequences(train['total text']) test['total_text'] = token.texts_to_sequences(test['total_text']) text size = len(token.word index) + 1# create a weight matrix for words in training docs --code copied from https://machinelearningmastery.com/use-word-embedding-layers-deep-learning-keras/ embedding matrix = np.zeros((text size, 300)) for word, i in token.word index.items(): embedding vector = glove model.get(word) if embedding vector is not None: embedding matrix[i] = embedding vector token = Tokenizer() token.fit_on_texts(train['clean_categories']) train['clean_categories']=token.texts_to_sequences(train['clean_categories']) test['clean_categories']=token.texts_to_sequences(test['clean_categories']) category size = len(token.word index) + 1 token = Tokenizer() token.fit on texts(train['teacher prefix']) train['teacher prefix']=token.texts to sequences(train['teacher prefix']) test['teacher prefix']=token.texts to sequences(test['teacher prefix']) prefix size = len(token.word index) + 1 token = Tokenizer() token.fit on texts(train['school state']) train['school state']=token.texts to sequences(train['school state']) test['school state'] = token.texts to sequences(test['school state']) state_size = len(token.word_index) + 1 token = Tokenizer() token.fit on texts(train['project grade category']) train['project_grade_category']=token.texts_to_sequences(train['project_grade_category']) test['project_grade_category']=token.texts_to_sequences(test['project_grade_category']) grade size = len(token.word index) + 1 token = Tokenizer()

In [74]: embedding_matrix.shape Out[74]: (67062, 300) In [75]: train.drop(['clean essay','clean title','index'] , axis=1, inplace=True)

train['clean_subcategories']=token.texts_to_sequences(train['clean_subcategories'])
test['clean_subcategories']=token.texts_to_sequences(test['clean_subcategories'])

test.drop(['clean_essay','clean_title','index'] , axis=1, inplace=True)

In [10]: train.tail(2)

Out[10]:

	index	teacher_prefix	school_state	project_grade_category	teacher_number_of_previously_posted_projects	proje			
98321	28030	[2]	[2]	[1, 2, 3]	1	1			
98322	15725	[2]	[18]	[1, 2, 3]	1	0			
4									

Getting data in the form of Dictionary which then will be given as a input to Deep Learning Models

In [13]:

```
from keras.preprocessing.sequence import pad_sequences
from numpy import array
```

In [78]:

```
X_train={}

X_train["posted_projects"]= array(train["teacher_number_of_previously_posted_projects"]).reshape(len(train),1)

X_train["price"]= array(train["price"]).reshape(len(train),1)

X_train["quantity"]= array(train["quantity"]).reshape(len(train),1)

X_train["Is_digit_present"]= array(train["Is_digit_present"]).reshape(len(train),1)

X_train["teacher_prefix"] = pad_sequences(train["teacher_prefix"], maxlen=1)

X_train["school_state"] = pad_sequences(train["school_state"], maxlen=1)

X_train["project_grade_category"] = pad_sequences(train["project_grade_category"], maxlen=3)

X_train["total_text"] = pad_sequences(train["total_text"], maxlen=300)

X_train["clean_categories"] = pad_sequences(train["clean_categories"], maxlen=4)

X_train["clean_subcategories"] = pad_sequences(train["clean_subcategories"], maxlen=4)

X_train["output"]= array(train["project_is_approved"])
```

In [79]:

```
X_test={}

X_test["posted_projects"]= array(test["teacher_number_of_previously_posted_projects"]).reshape(len (test),1)

X_test["price"]= array(test["price"]).reshape(len(test),1)

X_test["quantity"]= array(test["quantity"]).reshape(len(test),1)

X_test["Is_digit_present"]= array(test["Is_digit_present"]).reshape(len(test),1)

X_test["teacher_prefix"]= pad_sequences(test["teacher_prefix"], maxlen=1)

X_test["school_state"]= pad_sequences(test["school_state"], maxlen=1)

X_test["project_grade_category"]= pad_sequences(test["project_grade_category"], maxlen=3)

X_test["total_text"]= pad_sequences(test["total_text"], maxlen=300)

X_test["clean_categories"]= pad_sequences(test["clean_categories"], maxlen=4)

X_test["clean_subcategories"]= pad_sequences(test["clean_subcategories"], maxlen=4)
```

```
X test["output"] = array(test["project is approved"])
```

In [14]:

```
from keras.initializers import he normal
from keras.models import Model
from keras.optimizers import Adam
from keras.layers import Input, BatchNormalization, Embedding, LSTM, Flatten, concatenate, Dense, D
ropout
#import tensorflow as tf
#sess = tf.Session()
#from keras import backend as K
#K.set session(sess)
```

In [80]:

```
import platform
print(platform.python version())
```

3.6.8

Model 1

In [107]:

```
# Input layers
previously posted projects = Input(shape=(1,), name="posted projects")
price = Input(shape=(1,), name="price")
digit_present = Input(shape=(1,), name="Is_digit_present")
quantity = Input(shape=(1,), name="quantity")
school state = Input(shape=(1,), name="school state")
teacher prefix = Input(shape=(1,), name="teacher prefix")
project grade= Input(shape=(3,), name="project grade category")
total text = Input(shape=(300,), name="total text")
clean categories = Input(shape=(4,), name="clean categories")
clean subcategories = Input(shape=(4,), name="clean subcategories")
             # Batch normalization layer
\#previously\_posted\_projects\_bn = BatchNormalization (axis=1, momentum=0.99, epsilon=0.001, center=Table (axis=1, momentum=0.99, epsilon=0.99, epsilon=0.99, epsilon=0.99, ep
rue, scale=True, beta_initializer='zeros', gamma_initializer='ones',
moving mean initializer='zeros', moving variance initializer='ones') (previously posted projects)
#price_bn = BatchNormalization(axis=1, momentum=0.99, epsilon=0.001, center=True, scale=True, beta
 initializer='zeros', gamma initializer='ones', moving mean initializer='zeros',
moving variance initializer='ones') (price)
\#quantity_bn = BatchNormalization(axis=1, momentum=0.99, epsilon=0.001, center=True, scale=True, but is a second of the secon
eta initializer='zeros', gamma initializer='ones', moving mean initializer='zeros',
moving variance initializer='ones') (quantity)
            # Embedding layers
emb text layer = Embedding(text size, 300, weights=[embedding matrix], trainable=False)
emb_category_layer = Embedding(category_size, 4)
emb_subcategory_layer = Embedding(subcategory_size, 4)
emb_state_layer = Embedding(state_size, 8)
emb prefix layer = Embedding(prefix size, 2)
emb grade layer = Embedding(grade size, 2)
            # Giving Input to Embedding layers
emb text = emb text layer(total text)
emb_category =emb_category_layer(clean_categories)
emb subcategory =emb subcategory layer(clean subcategories)
emb state =emb state layer(school state)
emb prefix =emb prefix layer(teacher prefix)
emb_grade =emb_grade_layer(project_grade)
```

```
# LSTM layers
lstm_text = LSTM(12, activation="relu", return_sequences=True) (emb_text)
    # Flatten layers
flatten text =Flatten()(lstm text)
flatten category =Flatten()(emb category)
flatten_subcategory =keras.layers.Flatten()(emb_subcategory)
flatten state =Flatten()(emb state)
flatten_prefix =Flatten()(emb_prefix)
flatten grade =Flatten()(emb grade)
    # concatenation of all numeric layers
numeric= concatenate([previously posted projects,
                                      price,
                                      digit present,
                                       quantity])
    # Dense layer
dense numeric =Dense(4, activation='relu', kernel initializer=he normal(seed=5))(numeric)
    # Merge all layers into one
x = concatenate([dense numeric,
                                 flatten text,
                                 flatten category,
                                 flatten subcategory,
                                 flatten state,
                                  flatten prefix,
                                  flatten grade])
dense x =Dense(8, activation='relu', kernel initializer=he normal(seed=3))(x)
dense_x_bn= BatchNormalization(axis=1, momentum=0.99, epsilon=0.001, center=True, scale=True, beta_
initializer='zeros', gamma initializer='ones', moving mean initializer='zeros',
moving variance initializer='ones') (dense x)
\#drop \ x = Dropout(0.5) (dense \ x \ bn)
dense2 x =Dense(4, activation='relu', kernel initializer=he normal(seed=1)) (dense x bn)
\#drop2 \ x = Dropout(0.5)(dense2 \ x)
dense3 x =Dense(2, activation='relu', kernel initializer=he normal(seed=2))(dense2 x)
    # Dense layers
    \#x = keras.layers.Dense(128, activation="relu")(x)
    # Output layers
output = Dense(1, activation="sigmoid", name='final_output')(dense3_x)
model = Model(inputs=[previously posted projects,price,digit present,quantity,school state,teacher
prefix,project grade,total text,clean categories,clean subcategories], outputs=[output])
model.summary()
```

Layer (type)	Output	Shape	Param #	Connected to
total_text (InputLayer)	(None,	300)	0	
posted_projects (InputLayer)	(None,	1)	0	
price (InputLayer)	(None,	1)	0	
Is_digit_present (InputLayer)	(None,	1)	0	
quantity (InputLayer)	(None,	1)	0	
embedding_91 (Embedding)	(None,	300, 300)	20118600	total_text[0][0]
clean_categories (InputLayer)	(None,	4)	0	
clean_subcategories (InputLayer	(None,	4)	0	
school state (InputLayer)	(None,	1)	0	

teacher_prefix (InputLayer)	(None,	1)	0	
project_grade_category (InputLa	(None,	3)	0	
concatenate_31 (Concatenate)	(None,	4)	0	<pre>posted_projects[0][0] price[0][0] Is_digit_present[0][0] quantity[0][0]</pre>
lstm_16 (LSTM)	(None,	300, 12)	15024	embedding_91[0][0]
embedding_92 (Embedding)	(None,	4, 4)	64	clean_categories[0][0]
embedding_93 (Embedding)	(None,	4, 4)	152	clean_subcategories[0][0]
embedding_94 (Embedding)	(None,	1, 8)	416	school_state[0][0]
embedding_95 (Embedding)	(None,	1, 2)	12	teacher_prefix[0][0]
embedding_96 (Embedding)	(None,	3, 2)	20	<pre>project_grade_category[0][0]</pre>
dense_61 (Dense)	(None,	4)	20	concatenate_31[0][0]
flatten_91 (Flatten)	(None,	3600)	0	lstm_16[0][0]
flatten_92 (Flatten)	(None,	16)	0	embedding_92[0][0]
flatten_93 (Flatten)	(None,	16)	0	embedding_93[0][0]
flatten_94 (Flatten)	(None,	8)	0	embedding_94[0][0]
flatten_95 (Flatten)	(None,	2)	0	embedding_95[0][0]
flatten_96 (Flatten)	(None,	6)	0	embedding_96[0][0]
concatenate_32 (Concatenate)	(None,	3652)	0	dense_61[0][0] flatten_91[0][0] flatten_92[0][0] flatten_93[0][0] flatten_94[0][0] flatten_95[0][0] flatten_96[0][0]
dense_62 (Dense)	(None,	8)	29224	concatenate_32[0][0]
batch_normalization_44 (BatchNo	(None,	8)	32	dense_62[0][0]
dense_63 (Dense)	(None,	4)	36	batch_normalization_44[0][0]
dense_64 (Dense)	(None,	2)	10	dense_63[0][0]
final_output (Dense)	(None,	1)	3	dense_64[0][0]

Total params: 20,163,613 Trainable params: 44,997

Non-trainable params: 20,118,616

AUC metric Function

In [13]:

```
import tensorflow as tf

def roc_auc(y_true, y_pred):
    auc = tf.metrics.auc(y_true, y_pred,weights=None,num_thresholds=200)[1]
    keras.backend.get_session().run(tf.local_variables_initializer()) # use to reset the local
    variables created by tf.metrics.auc
    return auc
```

```
tb =TensorBoard(log dir="logs/{}".format(time()))
model.compile(optimizer = Adam(lr=1e-2),
                loss = {'final output': 'binary crossentropy'},
                metrics = [roc auc])
model.fit({"posted projects":X train['posted projects'], "price":X train['price'],
            "Is digit present":X train['Is digit present'], "quantity":X train['quantity'],
            "school state":X train['school state'], "teacher prefix":X train['teacher prefix'],
       "project_grade_category":X_train['project_grade_category'], "total_text":X_train['total_tex
t'],
       "clean categories":X train['clean categories'], "clean subcategories":X train['clean subcat
egories']}
            "final output":X train['output']},
               batch size=2500,
               epochs=40,
            validation data=(("posted projects":X test['posted projects'], "price":X test['price'
],
            "Is_digit_present":X_test['Is_digit_present'], "quantity":X_test['quantity'],
            "school_state":X_test['school_state'], "teacher_prefix":X_test['teacher_prefix'],
       "project_grade_category":X_test['project_grade_category'], "total_text":X_test['total_text'
],
       "clean categories":X test['clean categories'], "clean subcategories":X test['clean subcateg
ories']}
            "final output":X test['output']}),
              callbacks=[tb])
4
                                                                                     •
Train on 98323 samples, validate on 10925 samples
Epoch 1/40
98323/98323 [============== ] - 46s 465us/step - loss: 0.5809 - roc auc: 0.4886 - v
al loss: 0.4358 - val roc auc: 0.4888
Epoch 2/40
al loss: 0.4318 - val roc auc: 0.5271
Epoch 3/40
98323/98323 [============= ] - 38s 386us/step - loss: 0.4061 - roc auc: 0.5437 - v
al loss: 0.4327 - val roc auc: 0.5590
Epoch 4/40
98323/98323 [=============] - 38s 387us/step - loss: 0.4004 - roc auc: 0.5727 - v
al_loss: 0.4323 - val_roc_auc: 0.5832
Epoch 5/40
98323/98323 [==================== ] - 38s 389us/step - loss: 0.3962 - roc auc: 0.5928 - v
al loss: 0.4339 - val_roc_auc: 0.6011
Epoch 6/40
98323/98323 [============= ] - 38s 387us/step - loss: 0.3941 - roc auc: 0.6087 - v
al_loss: 0.4346 - val_roc_auc: 0.6142
Epoch 7/40
98323/98323 [============== ] - 38s 386us/step - loss: 0.3900 - roc auc: 0.6202 - v
al loss: 0.4356 - val roc auc: 0.6252
Epoch 8/40
98323/98323 [============= ] - 38s 383us/step - loss: 0.3875 - roc auc: 0.6303 - v
al loss: 0.4365 - val roc auc: 0.6344
Epoch 9/40
98323/98323 [============= ] - 38s 390us/step - loss: 0.3860 - roc auc: 0.6386 - v
al_loss: 0.4400 - val_roc_auc: 0.6416
Epoch 10/40
al loss: 0.4401 - val roc auc: 0.6480
Epoch 11/40
98323/98323 [============== ] - 38s 388us/step - loss: 0.3841 - roc auc: 0.6508 - v
al loss: 0.4418 - val roc auc: 0.6531
Epoch 12/40
98323/98323 [============== ] - 38s 388us/step - loss: 0.3827 - roc auc: 0.6557 - v
al loss: 0.4435 - val roc auc: 0.6574
Epoch 13/40
98323/98323 [============== ] - 38s 385us/step - loss: 0.3804 - roc auc: 0.6596 - v
al_loss: 0.4436 - val_roc_auc: 0.6615
Epoch 14/40
98323/98323 [==============] - 38s 386us/step - loss: 0.3779 - roc auc: 0.6637 - v
al loss. 0 4459 - val roc auc. 0 6655
```

```
at 1000. 0.7700 vat 100 auc. 0.0000
Epoch 15/40
al_loss: 0.4506 - val_roc_auc: 0.6689
Epoch 16/40
98323/98323 [============== ] - 38s 386us/step - loss: 0.3783 - roc auc: 0.6705 - v
al loss: 0.4488 - val roc auc: 0.6718
Epoch 17/40
al loss: 0.4467 - val roc auc: 0.6743
Epoch 18/40
98323/98323 [============= ] - 38s 387us/step - loss: 0.3760 - roc auc: 0.6756 - v
al loss: 0.4478 - val roc auc: 0.6768
Epoch 19/40
98323/98323 [============== ] - 38s 388us/step - loss: 0.3748 - roc auc: 0.6782 - v
al loss: 0.4462 - val roc auc: 0.6792
Epoch 20/40
al_loss: 0.4488 - val_roc_auc: 0.6814
Epoch 21/40
al loss: 0.4508 - val roc auc: 0.6835
Epoch 22/40
98323/98323 [============== ] - 38s 384us/step - loss: 0.3737 - roc auc: 0.6845 - v
al loss: 0.4520 - val roc auc: 0.6853
Epoch 23/40
98323/98323 [============== ] - 38s 386us/step - loss: 0.3712 - roc auc: 0.6864 - v
al loss: 0.4598 - val roc auc: 0.6871
Epoch 24/40
al loss: 0.4632 - val roc auc: 0.6886
Epoch 25/40
al loss: 0.4626 - val roc auc: 0.6900
Epoch 26/40
98323/98323 [============== ] - 39s 399us/step - loss: 0.3697 - roc auc: 0.6909 - v
al loss: 0.4558 - val roc auc: 0.6915
Epoch 27/40
98323/98323 [============= ] - 43s 438us/step - loss: 0.3706 - roc auc: 0.6923 - v
al loss: 0.4581 - val roc auc: 0.6928
Epoch 28/40
98323/98323 [============== ] - 38s 387us/step - loss: 0.3681 - roc auc: 0.6936 - v
al loss: 0.4646 - val roc auc: 0.6942
Epoch 29/40
98323/98323 [============= ] - 39s 392us/step - loss: 0.3694 - roc auc: 0.6948 - v
al_loss: 0.4679 - val_roc_auc: 0.6953
Epoch 30/40
98323/98323 [============== ] - 38s 386us/step - loss: 0.3663 - roc auc: 0.6960 - v
al_loss: 0.4630 - val_roc_auc: 0.6966
Epoch 31/40
al_loss: 0.4686 - val_roc_auc: 0.6977
Epoch 32/40
al loss: 0.4718 - val roc auc: 0.6988
Epoch 33/40
98323/98323 [============== ] - 38s 383us/step - loss: 0.3643 - roc auc: 0.6995 - v
al loss: 0.4767 - val roc auc: 0.6999
Epoch 34/40
98323/98323 [============== ] - 38s 386us/step - loss: 0.3663 - roc auc: 0.7004 - v
al_loss: 0.4808 - val_roc_auc: 0.7007
Epoch 35/40
98323/98323 [============= ] - 38s 383us/step - loss: 0.3663 - roc auc: 0.7012 - v
al loss: 0.4775 - val roc auc: 0.7015
Epoch 36/40
98323/98323 [============= ] - 38s 391us/step - loss: 0.3663 - roc auc: 0.7020 - v
al loss: 0.4655 - val roc auc: 0.7023
Epoch 37/40
98323/98323 [============= ] - 38s 386us/step - loss: 0.3644 - roc auc: 0.7029 - v
al loss: 0.4799 - val roc auc: 0.7032
Epoch 38/40
al loss: 0.4824 - val roc auc: 0.7039
Epoch 39/40
98323/98323 [============= ] - 38s 389us/step - loss: 0.3629 - roc auc: 0.7044 - v
al_loss: 0.4767 - val_roc_auc: 0.7047
Epoch 40/40
```

```
al_loss: 0.4801 - val_roc_auc: 0.7055

Out[108]:
<keras.callbacks.History at 0x198b5b67828>
```

Saving my neural network model to JSON

```
In [109]:
```

```
from keras.models import model_from_json

# serialize model to JSON
model_json = model.to_json()
with open("model.json", "w") as json_file:
    json_file.write(model_json)
# serialize weights to HDF5
model.save_weights("model.h5")
```

Getting TF-IDF values of words in text data

```
In [16]:
```

```
from sklearn.feature_extraction.text import TfidfVectorizer
from sklearn.feature_extraction.text import CountVectorizer
# merge texts
text = list(train['total_text'])

tfidf = TfidfVectorizer()
tfidf.fit_transform(text)

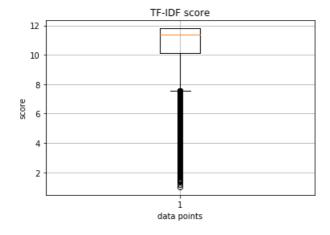
# dict key:word and value:tf-idf score
word2tfidf = dict(zip(tfidf.get_feature_names(), tfidf.idf_))
```

```
In [17]:
```

```
score=[]
for key in word2tfidf.keys():
    score.append(word2tfidf[key])
score=np.asarray(score).reshape(-1)
```

In [19]:

```
import matplotlib.pyplot as plt
plt.boxplot([score])
plt.title('TF-IDF score')
plt.xlabel('data points')
plt.ylabel('score')
plt.grid()
plt.show()
```



Removing words which have very high or very low TF-IDF values

```
In [20]:
remove_words=[]
for i in range (109248):
    total words=(data['total text'][i])
    wordList = re.sub("[^\w]", "", total_words).split()
    for j in wordList:
        try:
            idf = word2tfidf[i]
        except:
            idf = 0
        if(idf>10 and idf <12):</pre>
            continue;
        else:
            remove words.append(j)
In [21]:
remove words=list(set(remove words))
In [27]:
```

```
#code copied from ---https://stackoverflow.com/questions/45447848/check-for-words-from-list-and-re
move-those-words-in-pandas-dataframe-column
remove= r'\b(?:{})\b'.format('|'.join(remove_words))
train['total_text'] = train['total_text'].str.replace(remove, '')
test['total_text'] = test['total_text'].str.replace(remove, '')
```

```
In [35]:

token = Tokenizer()

token.fit_on_texts(train['total_text'])
train['total_text']=token.texts_to_sequences(train['total_text'])
test['total_text']=token.texts_to_sequences(test['total_text'])
text_size = len(token.word_index) + 1
```

Getting data in the form of Dictionary which then will be given as a input to Deep Learning Models

```
In [37]:

X_train={}

X_train["posted_projects"]= array(train["teacher_number_of_previously_posted_projects"]).reshape(len(train),1)

X_train["price"]= array(train["price"]).reshape(len(train),1)

X_train["quantity"]= array(train["quantity"]).reshape(len(train),1)

X_train["Is_digit_present"]= array(train["Is_digit_present"]).reshape(len(train),1)

X_train["teacher_prefix"]= pad_sequences(train["teacher_prefix"], maxlen=1)

X_train["school_state"]= pad_sequences(train["school_state"], maxlen=1)

X_train["project_grade_category"]= pad_sequences(train["project_grade_category"], maxlen=3)

X_train["total_text"]= pad_sequences(train["total_text"], maxlen=300)

X_train["clean_categories"]= pad_sequences(train["clean_categories"], maxlen=4)

X_train["clean_subcategories"]= pad_sequences(train["clean_subcategories"], maxlen=4)

X_train["output"]= array(train["project_is_approved"])
```

```
In [38]:
```

```
X_test={}

X_test["posted_projects"]= array(test["teacher_number_of_previously_posted_projects"]).reshape(len (test),1)
X_test["price"]= array(test["price"]).reshape(len(test),1)
X_test["quantity"]= array(test["quantity"]).reshape(len(test),1)
X_test["Is_digit_present"]= array(test["Is_digit_present"]).reshape(len(test),1)

X_test["teacher_prefix"]= pad_sequences(test["teacher_prefix"], maxlen=1)
X_test["school_state"]= pad_sequences(test["school_state"], maxlen=1)
X_test["project_grade_category"]= pad_sequences(test["project_grade_category"], maxlen=3)

X_test["total_text"]= pad_sequences(test["total_text"], maxlen=300)
X_test["clean_categories"]= pad_sequences(test["clean_categories"], maxlen=4)
X_test["clean_subcategories"]= pad_sequences(test["clean_subcategories"], maxlen=4)
X_test["output"]= array(test["project_is_approved"])
```

In [40]:

```
import keras
   # Input layers
previously_posted_projects = Input(shape=(1,), name="posted projects")
price = Input(shape=(1,), name="price")
digit_present = Input(shape=(1,), name="Is_digit_present")
quantity = Input(shape=(1,), name="quantity")
school state = Input(shape=(1,), name="school state")
teacher prefix = Input(shape=(1,), name="teacher prefix")
project grade= Input(shape=(3,), name="project grade category")
total_text = Input(shape=(300,), name="total_text")
clean_categories = Input(shape=(4,), name="clean_categories")
clean_subcategories = Input(shape=(4,), name="clean_subcategories")
    # Batch normalization layer
#previously_posted_projects_bn = BatchNormalization(axis=1, momentum=0.99, epsilon=0.001, center=T
rue, scale=True, beta_initializer='zeros', gamma_initializer='ones',
moving mean initializer='zeros', moving variance initializer='ones') (previously posted projects)
#price bn = BatchNormalization(axis=1, momentum=0.99, epsilon=0.001, center=True, scale=True, beta
initializer='zeros', gamma initializer='ones', moving mean initializer='zeros',
moving variance initializer='ones') (price)
#quantity bn = BatchNormalization(axis=1, momentum=0.99, epsilon=0.001, center=True, scale=True, b
eta initializer='zeros', gamma initializer='ones', moving mean initializer='zeros',
moving variance initializer='ones') (quantity)
    # Embedding layers
emb text layer = Embedding(text size, 300, weights=[embedding matrix], trainable=False)
emb category layer = Embedding(category size, 4)
emb_subcategory_layer = Embedding(subcategory_size, 4)
emb_state_layer = Embedding(state_size, 8)
emb prefix layer = Embedding(prefix size, 2)
emb grade layer = Embedding(grade_size, 2)
    # Giving Input to Embedding layers
emb text = emb text layer(total text)
emb_category =emb_category_layer(clean_categories)
emb subcategory =emb subcategory layer(clean subcategories)
emb state =emb state layer(school state)
emb prefix =emb prefix layer(teacher prefix)
emb_grade =emb_grade_layer(project_grade)
    # LSTM lavers
lstm text = LSTM(12, activation="relu", return sequences=True) (emb text)
   # Flatten layers
flatten text =Flatten()(lstm text)
flatten category =Flatten()(emb category)
flatten subcategory =keras.layers.Flatten()(emb subcategory)
```

```
flatten state =Flatten()(emb state)
flatten_prefix =Flatten()(emb_prefix)
flatten grade =Flatten()(emb grade)
    # concatenation of all numeric layers
numeric= concatenate([previously posted projects,
                                       price,
                                       digit_present,
                                       quantity])
    # Dense layer
dense numeric =Dense(4, activation='relu', kernel initializer=he normal(seed=5)) (numeric)
    # Merge all layers into one
x = concatenate([dense numeric,
                                  flatten text,
                                  flatten category,
                                  flatten subcategory,
                                  flatten_state,
                                  flatten prefix,
                                  flatten_grade])
{\tt dense\_x = Dense(8, activation='relu', kernel\_initializer=he\_normal(seed=3))(x)}
dense_x_bn= BatchNormalization(axis=1, momentum=0.99, epsilon=0.001, center=True, scale=True, beta
initializer='zeros', gamma_initializer='ones', moving_mean_initializer='zeros',
moving variance_initializer='ones') (dense_x)
\#drop \ x = Dropout(0.5) (dense \ x \ bn)
dense2 x =Dense(4, activation='relu', kernel initializer=he normal(seed=1))(dense x bn)
\#drop2 \ x = Dropout(0.5)(dense2 \ x)
dense3 x =Dense(2, activation='relu',kernel initializer=he normal(seed=2))(dense2 x)
    # Dense layers
    \#x = keras.layers.Dense(128, activation="relu")(x)
    # Output layers
output = Dense(1, activation="sigmoid", name='final output')(dense3 x)
model = Model(inputs=[previously posted projects,price,digit present,quantity,school state,teacher
prefix,project grade,total text,clean categories,clean subcategories], outputs=[output])
model.summary()
```

Layer (type)	Output	Shape	Param #	Connected to
total_text (InputLayer)	(None,	300)	0	
posted_projects (InputLayer)	(None,	1)	0	
price (InputLayer)	(None,	1)	0	
Is_digit_present (InputLayer)	(None,	1)	0	
quantity (InputLayer)	(None,	1)	0	
embedding_7 (Embedding)	(None,	300, 300)	18616800	total_text[0][0]
clean_categories (InputLayer)	(None,	4)	0	
clean_subcategories (InputLayer	(None,	4)	0	
school_state (InputLayer)	(None,	1)	0	
teacher_prefix (InputLayer)	(None,	1)	0	
project_grade_category (InputLa	(None,	3)	0	
concatenate_1 (Concatenate)	(None,	4)	0	<pre>posted_projects[0][0] price[0][0] Is_digit_present[0][0] quantity[0][0]</pre>

lstm_2 (LSTM)	(None,	300, 12)	15024	embedding_7[0][0]
embedding_8 (Embedding)	(None,	4, 4)	64	clean_categories[0][0]
embedding_9 (Embedding)	(None,	4, 4)	152	clean_subcategories[0][0]
embedding_10 (Embedding)	(None,	1, 8)	416	school_state[0][0]
embedding_11 (Embedding)	(None,	1, 2)	12	teacher_prefix[0][0]
embedding_12 (Embedding)	(None,	3, 2)	20	<pre>project_grade_category[0][0]</pre>
dense_1 (Dense)	(None,	4)	20	concatenate_1[0][0]
flatten_3 (Flatten)	(None,	3600)	0	lstm_2[0][0]
flatten_4 (Flatten)	(None,	16)	0	embedding_8[0][0]
flatten_5 (Flatten)	(None,	16)	0	embedding_9[0][0]
flatten_6 (Flatten)	(None,	8)	0	embedding_10[0][0]
flatten_7 (Flatten)	(None,	2)	0	embedding_11[0][0]
flatten_8 (Flatten)	(None,	6)	0	embedding_12[0][0]
concatenate_2 (Concatenate)	(None,	3652)	0	dense_1[0][0]
				flatten_3[0][0]
				flatten_4[0][0]
				flatten_5[0][0]
				flatten_6[0][0]
				flatten_7[0][0] flatten_8[0][0]
dense_2 (Dense)	(None,	8)	29224	concatenate_2[0][0]
batch_normalization_1 (BatchNor	(None,	8)	32	dense_2[0][0]
dense_3 (Dense)	(None,	4)	36	batch_normalization_1[0][0]
dense_4 (Dense)	(None,	2)	10	dense_3[0][0]
final_output (Dense)	(None,	1)	3	dense_4[0][0]

Non-trainable params: 18,616,816

Running Model 1 but with words removed based on their TF-IDF values

In [41]:

```
tb =TensorBoard(log dir="logs2/{}".format(time()))
metrics=[roc auc])
model.fit({"posted projects":X train['posted projects'], "price":X train['price'],
            "Is_digit_present":X_train['Is_digit_present'], "quantity":X_train['quantity'],
            "school_state":X_train['school_state'], "teacher_prefix":X_train['teacher_prefix'],
       "project_grade_category":X_train['project_grade_category'], "total_text":X_train['total_tex
t'],
       "clean_categories":X_train['clean_categories'], "clean_subcategories":X_train['clean_subcat
egories']}
            "final output":X train['output']},
              batch_size=2500,
              epochs=40,
```

```
validation data=({"posted projects":X test['posted projects'], "price":X test['price
'],
           "Is digit present":X test['Is digit present'], "quantity":X test['quantity'],
           "school state":X test['school state'], "teacher prefix":X test['teacher prefix'],
      "project grade category":X test['project grade category'], "total text":X test['total text'
],
      "clean categories":X test['clean categories'], "clean subcategories":X test['clean subcateg
ories']}
           "final_output":X_test['output']}),
            callbacks=[tb])
4
                                                                           Þ
WARNING:tensorflow:From C:\ProgramData\Anaconda3\lib\site-
packages\tensorflow\python\ops\metrics impl.py:526: to float (from tensorflow.python.ops.math ops)
is deprecated and will be removed in a future version.
Instructions for updating:
Use tf.cast instead.
{\tt WARNING:tensorflow:From C:\ProgramData\Anaconda3\lib\site-}
packages\tensorflow\python\ops\metrics impl.py:788: div (from tensorflow.python.ops.math ops) is d
eprecated and will be removed in a future version.
Instructions for updating:
Deprecated in favor of operator or tf.math.divide.
WARNING:tensorflow:From C:\ProgramData\Anaconda3\lib\site-
packages\tensorflow\python\ops\math_ops.py:3066: to_int32 (from tensorflow.python.ops.math_ops) is
deprecated and will be removed in a future version.
Instructions for updating:
Use tf.cast instead.
Train on 98323 samples, validate on 10925 samples
Epoch 1/40
al_loss: 0.6201 - val_roc_auc: 0.4675
Epoch 2/40
al loss: 0.5089 - val roc auc: 0.5061
Epoch 3/40
98323/98323 [============= ] - 33s 337us/step - loss: 0.5117 - roc auc: 0.5199 - v
al_loss: 0.4726 - val_roc_auc: 0.5307
Epoch 4/40
98323/98323 [============= ] - 33s 336us/step - loss: 0.4589 - roc auc: 0.5406 - v
al loss: 0.4307 - val_roc_auc: 0.5484
Epoch 5/40
98323/98323 [============= ] - 33s 335us/step - loss: 0.4236 - roc auc: 0.5563 - v
al loss: 0.4505 - val roc auc: 0.5630
Epoch 6/40
al loss: 0.4707 - val roc_auc: 0.5766
Epoch 7/40
al loss: 0.4460 - val roc_auc: 0.5899
Epoch 8/40
98323/98323 [============= ] - 33s 333us/step - loss: 0.3817 - roc auc: 0.5967 - v
al loss: 0.4476 - val roc auc: 0.6026
Epoch 9/40
98323/98323 [============== ] - 33s 338us/step - loss: 0.3781 - roc auc: 0.6085 - v
al loss: 0.4495 - val roc auc: 0.6137
Epoch 10/40
al loss: 0.4454 - val roc auc: 0.6238
Epoch 11/40
al loss: 0.4161 - val roc auc: 0.6334
Epoch 12/40
98323/98323 [============== ] - 34s 344us/step - loss: 0.3661 - roc auc: 0.6383 - v
al_loss: 0.4534 - val_roc_auc: 0.6421
Epoch 13/40
al loss: 0.4164 - val roc auc: 0.6499
Epoch 14/40
98323/98323 [============= ] - 33s 333us/step - loss: 0.3596 - roc auc: 0.6540 - v
al loss: 0.4248 - val roc auc: 0.6575
Epoch 15/40
98323/98323 [============= ] - 33s 336us/step - loss: 0.3567 - roc auc: 0.6612 - v
al loss: 0.4345 - val roc auc: 0.6643
Epoch 16/40
98323/98323 [============= ] - 33s 331us/step - loss: 0.3539 - roc auc: 0.6676 - v
al_loss: 0.4643 - val_roc_auc: 0.6705
```

```
EDOCH I//40
98323/98323 [============= ] - 33s 335us/step - loss: 0.3529 - roc auc: 0.6733 - v
al loss: 0.4027 - val_roc_auc: 0.6763
Epoch 18/40
98323/98323 [============== ] - 32s 329us/step - loss: 0.3489 - roc auc: 0.6795 - v
al_loss: 0.4042 - val_roc_auc: 0.6824
Epoch 19/40
al loss: 0.4231 - val_roc_auc: 0.6879
Epoch 20/40
al loss: 0.4086 - val roc auc: 0.6931
Epoch 21/40
98323/98323 [============== ] - 33s 333us/step - loss: 0.3409 - roc auc: 0.6957 - v
al loss: 0.4063 - val roc auc: 0.6981
Epoch 22/40
98323/98323 [============= ] - 33s 336us/step - loss: 0.3383 - roc auc: 0.7006 - v
al loss: 0.4158 - val roc auc: 0.7029
Epoch 23/40
98323/98323 [============= ] - 33s 336us/step - loss: 0.3356 - roc auc: 0.7054 - v
al loss: 0.4091 - val roc auc: 0.7075
Epoch 24/40
98323/98323 [============== ] - 33s 335us/step - loss: 0.3333 - roc auc: 0.7098 - v
al loss: 0.4304 - val roc auc: 0.7118
Epoch 25/40
98323/98323 [============= ] - 33s 337us/step - loss: 0.3309 - roc auc: 0.7140 - v
al loss: 0.4497 - val roc_auc: 0.7158
Epoch 26/40
98323/98323 [============== ] - 33s 333us/step - loss: 0.3286 - roc auc: 0.7178 - v
al loss: 0.4439 - val roc auc: 0.7196
Epoch 27/40
al_loss: 0.4760 - val_roc_auc: 0.7232
Epoch 28/40
98323/98323 [============== ] - 33s 333us/step - loss: 0.3231 - roc auc: 0.7251 - v
al loss: 0.5100 - val roc auc: 0.7266
Epoch 29/40
al_loss: 0.4288 - val_roc_auc: 0.7301
Epoch 30/40
al_loss: 0.4312 - val_roc_auc: 0.7336
Epoch 31/40
98323/98323 [============= ] - 34s 351us/step - loss: 0.3149 - roc auc: 0.7354 - v
al loss: 0.4542 - val roc auc: 0.7369
Epoch 32/40
98323/98323 [============= ] - 33s 334us/step - loss: 0.3138 - roc auc: 0.7386 - v
al_loss: 0.4366 - val_roc_auc: 0.7402
Epoch 33/40
98323/98323 [============== ] - 33s 334us/step - loss: 0.3122 - roc_auc: 0.7419 - v
al loss: 0.4799 - val roc auc: 0.7433
Epoch 34/40
98323/98323 [============== ] - 33s 334us/step - loss: 0.3110 - roc auc: 0.7448 - v
al loss: 0.5137 - val roc auc: 0.7460
Epoch 35/40
al loss: 0.5957 - val roc auc: 0.7483
Epoch 36/40
al loss: 0.5717 - val roc auc: 0.7505
Epoch 37/40
al_loss: 0.4656 - val_roc_auc: 0.7531
Epoch 38/40
98323/98323 [============== ] - 33s 336us/step - loss: 0.3026 - roc auc: 0.7545 - v
al_loss: 0.4533 - val_roc_auc: 0.7558
Epoch 39/40
al loss: 0.4777 - val_roc_auc: 0.7585
Epoch 40/40
98323/98323 [============== ] - 33s 337us/step - loss: 0.2981 - roc auc: 0.7599 - v
al loss: 0.4897 - val roc auc: 0.7610
```

Out[41]:

Saving my neural network model to JSON

```
In [42]:
```

```
# serialize model to JSON
model_json = model.to_json()
with open("model_reduce_words.json", "w") as json_file:
    json_file.write(model_json)
# serialize weights to HDF5
model.save_weights("model_reduce_words.h5")
```

One Hot encoding categorical data using keras preprocessing

```
In [38]:
```

```
train['teacher_prefix'] = train['teacher_prefix'].map(lambda a:keras.preprocessing.text.one_hot(a,1
00))
train['school_state'] = train['school_state'].map(lambda a:keras.preprocessing.text.one_hot(a,100))
train['project_grade_category'] = train['project_grade_category'].map(lambda a:keras.preprocessing.text.one_hot(a,100))
train['clean_categories'] = train['clean_categories'].map(lambda a:keras.preprocessing.text.one_hot(a,100))
train['clean_subcategories'] = train['clean_subcategories'].map(lambda a:keras.preprocessing.text.one_hot(a,100))
```

In [35]:

```
test['teacher_prefix'] = test['teacher_prefix'].map(lambda a:keras.preprocessing.text.one_hot(a,100))
test['school_state'] = test['school_state'].map(lambda a:keras.preprocessing.text.one_hot(a,100))
test['project_grade_category'] = test['project_grade_category'].map(lambda a:keras.preprocessing.text.one_hot(a,100))
test['clean_categories'] = test['clean_categories'].map(lambda a:keras.preprocessing.text.one_hot(a,100))
test['clean_subcategories'] = test['clean_subcategories'].map(lambda a:keras.preprocessing.text.one_hot(a,100))
```

Getting data in the form of Dictionary which then will be given as a input to Deep Learning Models

```
In [16]:
```

```
X_train={}

X_train={}

X_train["posted_projects"]= array(train["teacher_number_of_previously_posted_projects"]).reshape(1
en(train),1)

X_train["price"]= array(train["price"]).reshape(len(train),1)

X_train["quantity"]= array(train["quantity"]).reshape(len(train),1)

X_train["Is_digit_present"]= array(train["Is_digit_present"]).reshape(len(train),1)

X_train["teacher_prefix"]= pad_sequences(train["teacher_prefix"], maxlen=4)

X_train["school_state"]= pad_sequences(train["school_state"], maxlen=4)

X_train["project_grade_category"]= pad_sequences(train["project_grade_category"], maxlen=4)

X_train["total_text"]= pad_sequences(train["total_text"], maxlen=300)

X_train["clean_categories"]= pad_sequences(train["clean_categories"], maxlen=4)

X_train["clean_subcategories"]= pad_sequences(train["clean_subcategories"], maxlen=4)

X_train["output"]= array(train["project_is_approved"])
```

```
In [17]:
```

```
X_test={}

X_test["posted_projects"]= array(test["teacher_number_of_previously_posted_projects"]).reshape(len (test),1)

X_test["price"]= array(test["price"]).reshape(len(test),1)
```

```
X_test["quantity"]= array(test["quantity"]).reshape(len(test),1)
X_test["Is_digit_present"]= array(test["Is_digit_present"]).reshape(len(test),1)

X_test["teacher_prefix"]= pad_sequences(test["teacher_prefix"], maxlen=4)
X_test["school_state"]= pad_sequences(test["school_state"], maxlen=4)
X_test["project_grade_category"]= pad_sequences(test["project_grade_category"], maxlen=4)

X_test["total_text"]= pad_sequences(test["total_text"], maxlen=300)
X_test["clean_categories"]= pad_sequences(test["clean_categories"], maxlen=4)
X_test["clean_subcategories"]= pad_sequences(test["clean_subcategories"], maxlen=4)
X_test["output"]= array(test["project_is_approved"])
```

Model 2

In [18]:

```
from keras.layers.core import Reshape
from keras.layers import Conv1D
```

In [28]:

```
import keras
   # Input layers
previously posted projects = Input(shape=(1,), name="posted projects")
price = Input(shape=(1,), name="price")
digit_present = Input(shape=(1,), name="Is_digit_present")
quantity = Input(shape=(1,), name="quantity")
school_state = Input(shape=(4,), name="school_state")
teacher prefix = Input(shape=(4,), name="teacher prefix")
project grade= Input(shape=(4,), name="project grade category")
total text = Input(shape=(300,), name="total text")
clean_categories = Input(shape=(4,), name="clean categories")
clean subcategories = Input(shape=(4,), name="clean subcategories")
    # Embedding layers
emb text layer = Embedding(text size, 300, weights=[embedding matrix], trainable=False)
   # Giving Input to Embedding layers
emb_text = emb_text_layer(total_text)
   # LSTM layers
lstm_text = LSTM(14, activation="relu", return_sequences=True)(emb_text)
    # Flatten layers
flatten_text = Flatten()(lstm_text)
    # concatenation of all numeric and categorical layers
other= concatenate([
                                      school state,
                                      teacher_prefix,
                                      project_grade,
                                      clean categories,
                                      clean subcategories])
   # Dense layer
new = Reshape([4,-1])(other)
    # cnn layer
cnn 1= Conv1D(12,1, activation='relu') (new)
cnn 2= Conv1D(24, 3, activation='relu') (cnn 1)
    # Flatten layer
flatten cnn 2= Flatten()(cnn 2)
     # Batch normalization layer
#previously_posted_projects_bn = BatchNormalization(axis=1, momentum=0.99, epsilon=0.001, center=T
rue, scale=True, beta initializer='zeros', gamma initializer='ones',
moving mean initializer='zeros' moving variance initializer='ones') (previously nosted projects)
```

```
#price_bn= BatchNormalization(axis=1, momentum=0.99, epsilon=0.001, center=True, scale=True, beta_
initializer='zeros', gamma initializer='ones', moving mean initializer='zeros',
moving_variance_initializer='ones') (price)
#quantity_bn = BatchNormalization(axis=1, momentum=0.99, epsilon=0.001, center=True, scale=True, b
eta initializer='zeros', gamma initializer='ones', moving mean initializer='zeros',
moving variance initializer='ones') (quantity)
    # Merge all layers into one
x = concatenate([flatten text, flatten cnn 2,previously posted projects,
                                  price,
                                  quantity,
                                  digit present])
dense x = Dense(8, activation='relu', kernel initializer=he normal(seed=None))(x)
dense x bn= keras.layers.BatchNormalization(axis=1, momentum=0.99, epsilon=0.001, center=True, scal
e=True, beta_initializer='zeros', gamma_initializer='ones', moving_mean_initializer='zeros', moving
variance initializer='ones') (dense x)
\#drop \ x = Dropout(0.5) (dense \ x \ bn)
dense2 x = Dense(4, activation = 'relu', kernel initializer = he normal(seed = None)) (dense x bn)
\#drop2 \ x = Dropout(0.5) (dense2 \ x)
dense3 x = Dense(2, activation='relu', kernel initializer=he normal(seed=None)) (dense2 x)
    # Dense layers
    \#x = keras.layers.Dense(128, activation="relu")(x)
    # Output lavers
output = Dense(1, activation="sigmoid", name="final output")(dense3 x)
model = Model(inputs=[previously_posted_projects,price,digit_present,quantity,school_state,teacher_
prefix,project grade,total text,clean categories,clean subcategories], outputs=[output])
model.summary()
```

Layer (type)	Output	Shape	Param #	Connected to
school_state (InputLayer)	(None,	4)	0	
teacher_prefix (InputLayer)	(None,	4)	0	
project_grade_category (InputLa	(None,	4)	0	
clean_categories (InputLayer)	(None,	4)	0	
clean_subcategories (InputLayer	(None,	4)	0	
concatenate_13 (Concatenate)	(None,	20)	0	<pre>school_state[0][0] teacher_prefix[0][0] project_grade_category[0][0] clean_categories[0][0] clean_subcategories[0][0]</pre>
total_text (InputLayer)	(None,	300)	0	
reshape_7 (Reshape)	(None,	4, 5)	0	concatenate_13[0][0]
embedding_7 (Embedding)	(None,	300, 300)	18616800	total_text[0][0]
conv1d_13 (Conv1D)	(None,	4, 12)	72	reshape_7[0][0]
lstm_7 (LSTM)	(None,	300, 14)	17640	embedding_7[0][0]
convld_14 (ConvlD)	(None,	2, 24)	888	conv1d_13[0][0]
flatten_13 (Flatten)	(None,	4200)	0	lstm_7[0][0]
flatten_14 (Flatten)	(None,	48)	0	conv1d_14[0][0]
posted_projects (InputLayer)	(None,	1)	0	

price (InputLayer) (None, 1) quantity (InputLayer) (None, 1) 0 Is digit present (InputLayer) (None, 1) Ω concatenate_14 (Concatenate) flatten 13[0][0] (None, 4252) 0 flatten_14[0][0] posted projects[0][0] price[0][0] quantity[0][0] Is digit present[0][0] 34024 dense 17 (Dense) (None, 8) concatenate 14[0][0] dense_17[0][0] batch_normalization_6 (BatchNor (None, 8) 32 dense 18 (Dense) batch normalization 6[0][0] (None, 4) 36 dense 19 (Dense) 10 dense 18[0][0] (None, 2) final output (Dense) (None, 1) dense 19[0][0] Total params: 18,669,505 Trainable params: 52,689

- ----

Epoch 2/40

Epoch 3/40

Epoch 4/40

al loss: 0.3895 - val_roc_auc: 0.6835

al loss: 0.3843 - val roc auc: 0.7091

Non-trainable params: 18,616,816

```
In [29]:
tb =TensorBoard(log dir="logs3/{}".format(time()))
model.compile(optimizer= Adam(lr=1e-2),
                 loss={'final output': 'binary_crossentropy'},
                  metrics=[roc auc])
model.fit({"posted projects":X train['posted projects'], "price":X train['price'],
              "Is_digit_present":X_train['Is_digit_present'], "quantity":X_train['quantity'],
              "school_state":X_train['school_state'], "teacher_prefix":X_train['teacher_prefix'],
        "project_grade_category":X_train['project_grade_category'], "total_text":X_train['total_tex
t'],
        "clean_categories":X_train['clean_categories'], "clean_subcategories":X_train['clean_subcat
egories']}
              "final output": X train['output']},
                batch size=2500,
                epochs=40,
               validation data=({"posted projects":X test['posted projects'], "price":X test['price
'],
              "Is_digit_present":X_test['Is_digit_present'], "quantity":X_test['quantity'],
              "school_state":X_test['school_state'], "teacher_prefix":X_test['teacher_prefix'],
        "project_grade_category":X_test['project_grade_category'], "total_text":X_test['total_text'
],
        "clean categories":X test['clean categories'], "clean subcategories":X test['clean subcateg
ories']}
              "final_output":X_test['output']}),
                callbacks=[tb])
Train on 98323 samples, validate on 10925 samples
Epoch 1/40
98323/98323 [==================== ] - 36s 371us/step - loss: 0.4426 - roc auc: 0.5899 - v
al loss: 0.4213 - val_roc_auc: 0.6301
```

98323/98323 [=============] - 34s 341us/step - loss: 0.3742 - roc auc: 0.6606 - v

98323/98323 [=============] - 34s 341us/step - loss: 0.3647 - roc auc: 0.6986 - v

0 0570

A 71AF

```
al loss: 0.4054 - val roc auc: 0.7250
Epoch 5/40
98323/98323 [============== ] - 34s 342us/step - loss: 0.3504 - roc auc: 0.7317 - v
al loss: 0.3895 - val roc auc: 0.7371
Epoch 6/40
98323/98323 [============== ] - 34s 341us/step - loss: 0.3401 - roc auc: 0.7435 - v
al loss: 0.3904 - val roc_auc: 0.7484
Epoch 7/40
98323/98323 [============= ] - 34s 341us/step - loss: 0.3321 - roc auc: 0.7542 - v
al loss: 0.4048 - val roc auc: 0.7582
Epoch 8/40
98323/98323 [============= ] - 34s 343us/step - loss: 0.3228 - roc auc: 0.7634 - v
al loss: 0.4067 - val roc auc: 0.7672
Epoch 9/40
98323/98323 [============== ] - 33s 338us/step - loss: 0.3134 - roc auc: 0.7720 - v
al loss: 0.4245 - val roc auc: 0.7758
Epoch 10/40
98323/98323 [============== ] - 34s 343us/step - loss: 0.3072 - roc auc: 0.7803 - v
al_loss: 0.4499 - val_roc_auc: 0.7831
Epoch 11/40
98323/98323 [============== ] - 34s 341us/step - loss: 0.2974 - roc auc: 0.7874 - v
al_loss: 0.4687 - val_roc_auc: 0.7901
Epoch 12/40
98323/98323 [============= ] - 34s 341us/step - loss: 0.2916 - roc auc: 0.7939 - v
al loss: 0.4538 - val roc auc: 0.7965
Epoch 13/40
98323/98323 [============== ] - 34s 347us/step - loss: 0.2845 - roc auc: 0.8000 - v
al loss: 0.4703 - val roc auc: 0.8026
Epoch 14/40
98323/98323 [============= ] - 33s 340us/step - loss: 0.2759 - roc auc: 0.8060 - v
al loss: 0.4887 - val_roc_auc: 0.8084
Epoch 15/40
al_loss: 0.4735 - val_roc_auc: 0.8136
Epoch 16/40
al loss: 0.5081 - val_roc_auc: 0.8178
Epoch 17/40
al loss: 0.5350 - val roc auc: 0.8225
Epoch 18/40
al loss: 0.5052 - val roc auc: 0.8257
Epoch 19/40
98323/98323 [============== ] - 33s 339us/step - loss: 0.2589 - roc auc: 0.8278 - v
al loss: 0.5571 - val roc auc: 0.8293
Epoch 20/40
al loss: 0.5349 - val roc auc: 0.8331
Epoch 21/40
98323/98323 [============== ] - 34s 343us/step - loss: 0.2441 - roc auc: 0.8353 - v
al_loss: 0.5689 - val_roc_auc: 0.8368
Epoch 22/40
98323/98323 [============== ] - 33s 339us/step - loss: 0.2405 - roc auc: 0.8387 - v
al_loss: 0.5986 - val_roc_auc: 0.8401
Epoch 23/40
98323/98323 [============== ] - 33s 340us/step - loss: 0.2348 - roc auc: 0.8420 - v
al loss: 0.6061 - val roc auc: 0.8433
Epoch 24/40
98323/98323 [============= ] - 34s 341us/step - loss: 0.2321 - roc auc: 0.8450 - v
al loss: 0.5913 - val roc auc: 0.8463
Epoch 25/40
al loss: 0.6160 - val_roc_auc: 0.8493
Epoch 26/40
al_loss: 0.6249 - val_roc_auc: 0.8520
Epoch 27/40
al loss: 0.6534 - val_roc_auc: 0.8547
Epoch 28/40
98323/98323 [============== ] - 33s 340us/step - loss: 0.2179 - roc auc: 0.8562 - v
al loss: 0.6365 - val roc auc: 0.8572
Epoch 29/40
98323/98323 [=============] - 33s 340us/step - loss: 0.2166 - roc auc: 0.8586 - v
al loss: 0.6717 - val roc auc: 0.8595
```

```
Epoch 30/40
98323/98323 [============== ] - 34s 341us/step - loss: 0.2143 - roc auc: 0.8608 - v
al loss: 0.6637 - val roc auc: 0.8617
Epoch 31/40
al_loss: 0.6719 - val_roc_auc: 0.8638
Epoch 32/40
al loss: 0.6991 - val roc auc: 0.8657
Epoch 33/40
al loss: 0.7001 - val roc auc: 0.8677
Epoch 34/40
al loss: 0.6934 - val roc_auc: 0.8696
Epoch 35/40
al loss: 0.7564 - val roc auc: 0.8713
Epoch 36/40
al_loss: 0.7229 - val_roc_auc: 0.8730
Epoch 37/40
98323/98323 [============== ] - 33s 341us/step - loss: 0.2060 - roc auc: 0.8739 - v
al_loss: 0.6924 - val_roc_auc: 0.8745
Epoch 38/40
al_loss: 0.6980 - val_roc_auc: 0.8760
Epoch 39/40
98323/98323 [============== ] - 33s 339us/step - loss: 0.1936 - roc auc: 0.8769 - v
al_loss: 0.7701 - val_roc_auc: 0.8775
Epoch 40/40
al loss: 0.7549 - val roc auc: 0.8790
Out[29]:
<keras.callbacks.History at 0x2bff0e0e710>
```

Saving my neural network model to JSON

```
In [31]:
```

```
# serialize model to JSON
model_json = model.to_json()
with open("model_conv1.json", "w") as json_file:
    json_file.write(model_json)
# serialize weights to HDF5
model.save_weights("model_conv1.h5")
```

Result

In [15]:

```
#code copied from -http://zetcode.com/python/prettytable/
from prettytable import PrettyTable
x = PrettyTable()
x.field_names = ["Model no.", "train auc", "test auc"]
x.add_row(["Model 1", 0.7052 ,0.7055])
x.add_row(["Model 1 with reduced words", 0.7599, 0.7610])
x.add_row(["Model 2", 0.8784 ,0.8790])
print(x)
```

