

Assignment : 14

1. Download the preprocessed DonorsChoose data from here [Dataset \(https://drive.google.com/file/d/1GU3LIJJ3zS1xLXXe-sdItSJHtI5txjV0/view?usp=sharing\)](https://drive.google.com/file/d/1GU3LIJJ3zS1xLXXe-sdItSJHtI5txjV0/view?usp=sharing).
2. Split the data into train, cv, and test
3. After step 2 you have to train 3 types of models as discussed below.
4. For all the model use 'auc' (https://scikit-learn.org/stable/modules/model_evaluation.html#roc-metrics) as a metric. check [this \(https://datascience.stackexchange.com/a/20192\)](https://datascience.stackexchange.com/a/20192) for using auc as a metric. you need to print the AUC value for each epoch. Note: you should NOT use the tf.metric.auc
5. You are free to choose any number of layers/hiddenn units but you have to use same type of architectures shown below.
6. You can use any one of the optimizers and choice of Learning rate and momentum, resources: [cs231n class notes \(http://cs231n.github.io/neural-networks-3/\)](http://cs231n.github.io/neural-networks-3/), [cs231n class video \(https://www.youtube.com/watch?v=hd_KFJ5ktUc\)](https://www.youtube.com/watch?v=hd_KFJ5ktUc).
7. You should Save the best model weights.
8. For all the model's use [TensorBoard \(https://www.youtube.com/watch?v=2U6Jl7oqRkM\)](https://www.youtube.com/watch?v=2U6Jl7oqRkM) and plot the Metric value and Loss with epoch. While submitting, take a screenshot of plots and include those images in .ipynb notebook and PDF.
9. Use Categorical Cross Entropy as Loss to minimize.
10. try to get AUC more than 0.8 for atleast one model

Model-1

Build and Train deep neural network as shown below



ref: <https://i.imgur.com/w395Yk9.png> (<https://i.imgur.com/w395Yk9.png>)

- **Input_seq_total_text_data** --- You have to give Total text data columns. After this use the Embedding layer to get word vectors. Use given predefined glove word vectors, don't train any word vectors. After this use LSTM and get the LSTM output and Flatten that output.
- **Input_school_state** --- Give 'school_state' column as input to embedding layer and Train the Keras Embedding layer.
- **Project_grade_category** --- Give 'project_grade_category' column as input to embedding layer and Train the Keras Embedding layer.
- **Input_clean_categories** --- Give 'input_clean_categories' column as input to embedding layer and Train the Keras Embedding layer.
- **Input_clean_subcategories** --- Give 'input_clean_subcategories' column as input to embedding layer and Train the Keras Embedding layer.
- **Input_clean_subcategories** --- Give 'input_teacher_prefix' column as input to embedding layer and Train the Keras Embedding layer.
- **Input_remaining_teacher_number_of_previously_posted_projects._resource_summary_contains_** --- concatenate remaining columns and add a Dense layer after that.

- For LSTM, you can choose your sequence padding methods on your own or you can train your LSTM without padding, there is no restriction on that.

Below is an example of embedding layer for a categorical columns. In below code all are dummy values, we gave only for reference.

In [0]:

```
# https://stats.stackexchange.com/questions/270546/how-does-keras-embedding-layer-work
input_layer = Input(shape=(n,))
embedding = Embedding(no_1, no_2, input_length=n)(input_layer)
flatten = Flatten()(embedding)
```

1. Go through this blog, if you have any doubt on using predefined Embedding values in Embedding layer - <https://machinelearningmastery.com/use-word-embedding-layers-deep-learning-keras/> (<https://machinelearningmastery.com/use-word-embedding-layers-deep-learning-keras/>)

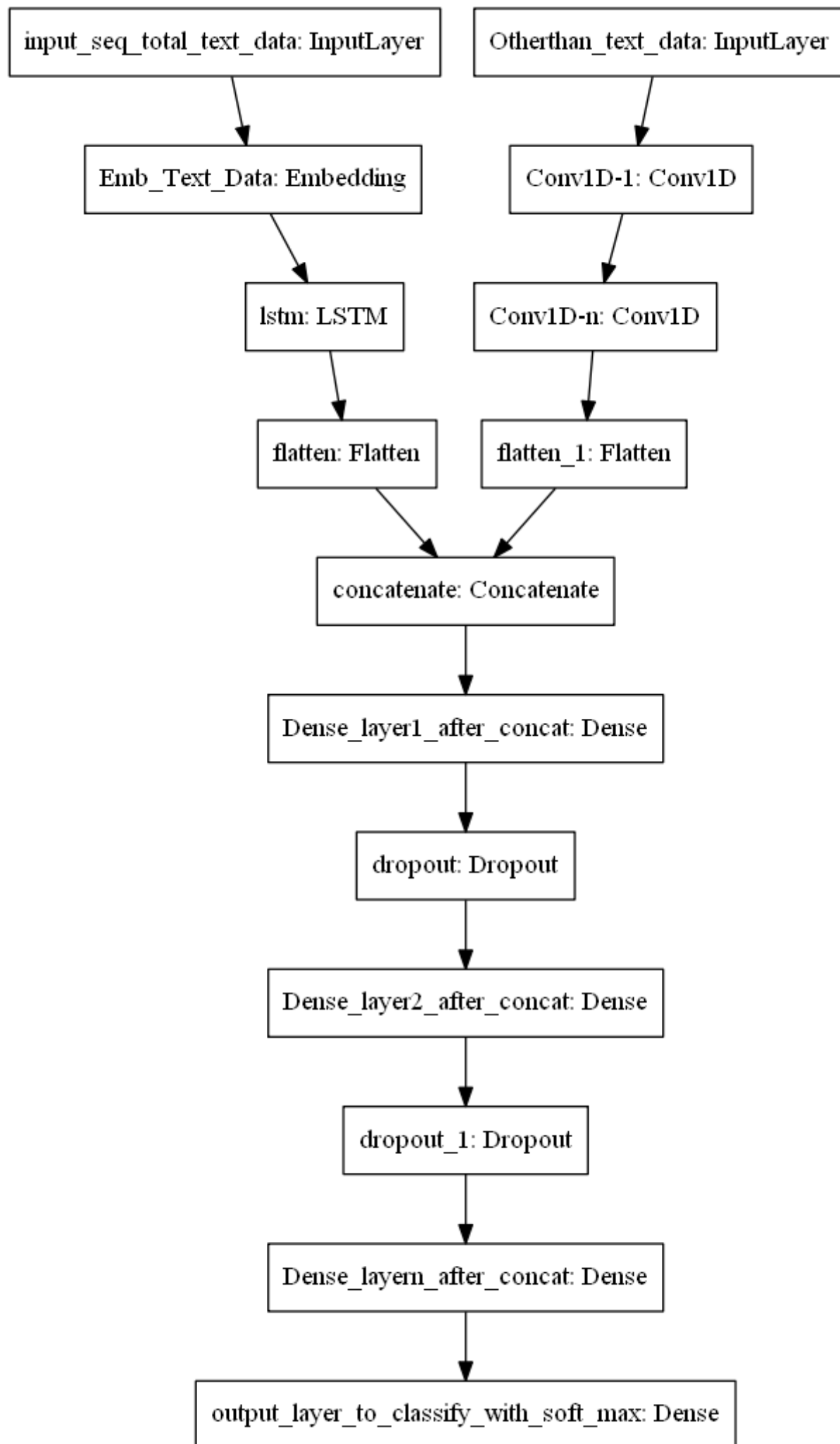
2. Please go through this link <https://keras.io/getting-started/functional-api-guide/> (<https://keras.io/getting-started/functional-api-guide/>) and check the 'Multi-input and multi-output models' then you will get to know how to give multiple inputs.

Model-2

Use the same model as above but for 'input_seq_total_text_data' give only some words in the sentence not all the words. Filter the words as below.

1. Train the TF-IDF on the Train data feature 'essay'
2. Get the idf value for each word we have in the train data.
3. Remove the low idf value and high idf value words from our data. Do some analysis on the Idf values and based on those values choose the low and high threshold value. Because very frequent words and very very rare words don't give much information. (you can plot a box plots and take only the idf scores within IQR range and corresponding words)
4. Train the LSTM after removing the Low and High idf value words. (In model-1 Train on total data but in Model-2 train on data after removing some words based on IDF values)

Model-3



ref: <https://i.imgur.com/fkQ8nGo.png> (<https://i.imgur.com/fkQ8nGo.png>)

- **input_seq_total_text_data:**

- . Use text column('essay'), and use the Embedding layer to get word vectors.
- . Use given predefined glove word vectors, don't train any word vectors.
- . Use LSTM that is given above, get the LSTM output and Flatten that output.
- . You are free to preprocess the input text as you needed.

- **Other_than_text_data:**

- . Convert all your Categorical values to onehot coded and then concatenate all these onehot vectors
- . Neumerical values and use [CNN1D \(https://keras.io/getting-started/sequential-model-guide/#sequence-classification-with-1d-convolutions\)](https://keras.io/getting-started/sequential-model-guide/#sequence-classification-with-1d-convolutions) as shown in above figure.
- . You are free to choose all CNN parameters like kernel sizes, stride.

</pre>

In [1]:

```
import numpy as np
import pandas as pd
from keras.preprocessing.sequence import pad_sequences
from keras.models import Sequential
from keras.layers import Dense, Input, Dropout
from keras.layers import Flatten
from keras.layers import concatenate
from keras.layers.embeddings import Embedding
from keras.models import Model
from keras.utils import to_categorical
from sklearn.model_selection import train_test_split
from keras.preprocessing.text import Tokenizer
import matplotlib.pyplot as plt
import pickle
from keras.layers import LSTM
from keras.preprocessing.text import text_to_word_sequence
import tensorflow as tf
from keras.callbacks import ModelCheckpoint, TensorBoard, ReduceLROnPlateau, EarlyStopping
from keras.layers.normalization import BatchNormalization
from sklearn.feature_extraction.text import TfidfVectorizer
import seaborn as sns
from keras.regularizers import l2
from sklearn.metrics import roc_auc_score
from keras.models import load_model
from IPython.display import Image
from scipy.sparse import hstack
from keras.layers import Conv1D
from sklearn.feature_extraction.text import CountVectorizer
from prettytable import PrettyTable
```

Using TensorFlow backend.

The default version of TensorFlow in Colab will soon switch to TensorFlow 2.x.
We recommend you [upgrade \(https://www.tensorflow.org/guide/migrate\)](https://www.tensorflow.org/guide/migrate) now or ensure your notebook will continue to use TensorFlow 1.x via the `%tensorflow_version 1.x` magic:
[more info \(https://colab.research.google.com/notebooks/tensorflow_version.ipynb\)](https://colab.research.google.com/notebooks/tensorflow_version.ipynb).

In [2]:

```
import pandas as pd
data=pd.read_csv('preprocessed_data.csv')
data.shape
```

Out[2]:

(109248, 9)

In [3]:

```
data.columns
```

Out[3]:

```
Index(['school_state', 'teacher_prefix', 'project_grade_category',  
      'teacher_number_of_previously_posted_projects', 'project_is_approved',  
      'clean_categories', 'clean_subcategories', 'essay', 'price'],  
      dtype='object')
```

In [4]:

```
y=data['project_is_approved'].values  
y.shape
```

Out[4]:

```
(109248,)
```

In [0]:

```
data.drop(['project_is_approved'],axis=1,inplace=True)
```

In [6]:

```
data.shape
```

Out[6]:

```
(109248, 8)
```

SPLITTING THE DATA

In [7]:

```
from sklearn.model_selection import train_test_split  
X_train,X_test,y_train,y_test=train_test_split(data,y,test_size=0.30,stratify=y)  
X_train,X_cv,y_train,y_cv=train_test_split(X_train,y_train,test_size=0.30,stratify=y_train)  
print(X_train.shape,y_train.shape)  
print(X_cv.shape,y_cv.shape)  
print(X_test.shape,y_test.shape)
```

```
(53531, 8) (53531,)  
(22942, 8) (22942,)  
(32775, 8) (32775,)
```

In [0]:

```
y_train = to_categorical(y_train, num_classes=2)  
y_cv =to_categorical(y_cv,num_classes=2)  
y_test = to_categorical(y_test, num_classes=2)
```

TOKENIZING ESSAY WORDS

In [9]:

```
token = Tokenizer()  
token.fit_on_texts(X_train['essay'])  
vocab_size = len(token.word_index) + 1  
print('Total unique words in the x_train', vocab_size)  
encoded_train = token.texts_to_sequences(X_train['essay'])  
encoded_cv = token.texts_to_sequences(X_cv['essay'])  
encoded_test = token.texts_to_sequences(X_test['essay'])
```

Total unique words in the x_train 42722

In [10]:

```
len(encoded_train)
```

Out[10]:

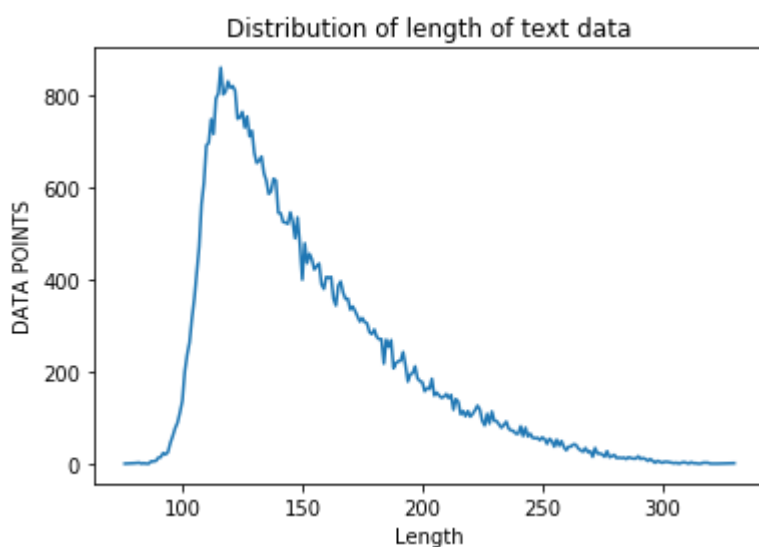
53531

In [0]:

```
length = []  
for sentence in encoded_train:  
    length.append(len(sentence))  
  
s = list(set(length))  
count = []  
for i in s:  
    count.append(length.count(i))
```

In [12]:

```
plt.plot(s, count)  
plt.xlabel('Length')  
plt.ylabel('DATA POINTS')  
plt.title('Distribution of length of text data')  
plt.show()
```



PADDING

In [13]:

```

max_length = 300
padded_train = pad_sequences(encoded_train, maxlen=max_length, padding='post')
padded_cv     = pad_sequences(encoded_cv, maxlen=max_length, padding='post')
padded_test  = pad_sequences(encoded_test, maxlen=max_length, padding='post')
print("length of padded_train data", len(padded_train))
print("length of padded_cv data", len(padded_cv))
print("length of padded_test data", len(padded_test))

```

length of padded_train data 53531

length of padded_cv data 22942

length of padded_test data 32775

In [0]:

In [0]:

```

with open('glove_vectors', 'rb') as f:
    model = pickle.load(f)
    glove_words = set(model.keys())
embedding_matrix_train = np.zeros((vocab_size, 300))
for word, i in token.word_index.items():
    if word in glove_words:
        embedding_vector = model[word]
        embedding_matrix_train[i] = embedding_vector

```

ENCODING CATEGORICAL VARIABLES

In [0]:

```

def label_encoder(column):
    unique = list(set(column))
    total = list(column)
    size = len(unique)
    count = []
    for category in unique:
        count.append([total.count(category), category])
    count.sort()
    rank = {}
    for i in range(1, len(count)+1):
        rank.update({count[i-1][1] : i})
    return (rank, unique, size)

```

In [17]:

```
category_rank,unique,size = label_encoder(X_train['clean_categories'])
print(category_rank)
categories_size = size
encoded_cat_train = []
encoded_cat_cv = []
encoded_cat_test = []
for category in X_train['clean_categories']:
    encoded_cat_train.append(category_rank[category])
for category in X_cv['clean_categories']:
    if category in unique:
        encoded_cat_cv.append(category_rank[category])
    else:
        encoded_cat_cv.append(0)

for category in X_test['clean_categories']:
    if category in unique:
        encoded_cat_test.append(category_rank[category])
    else:
        encoded_cat_test.append(0)

encoded_cat_train = np.asarray(encoded_cat_train)
encoded_cat_cv = np.asarray(encoded_cat_cv)
encoded_cat_test = np.asarray(encoded_cat_test)

print(encoded_cat_train.shape)
print(encoded_cat_cv.shape)
print(encoded_cat_test.shape)
```

```
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nguage math_science': 48, 'math_science': 49, 'literacy_language': 50}
(53531,)
(22942,)
(32775,)
```

In [18]:

```
subcategory_rank, unique, size = label_encoder(X_train['clean_subcategories'])
print(subcategory_rank)
subcategories_size = size
encoded_subcat_train = []
encoded_subcat_cv = []
encoded_subcat_test = []
for category in X_train['clean_subcategories']:
    encoded_subcat_train.append(subcategory_rank[category])
for category in X_cv['clean_subcategories']:
    if category in unique:
        encoded_subcat_cv.append(subcategory_rank[category])
    else:
        encoded_subcat_cv.append(0)

for category in X_test['clean_subcategories']:
    if category in unique:
        encoded_subcat_test.append(subcategory_rank[category])
    else:
        encoded_subcat_test.append(0)

encoded_subcat_train = np.asarray(encoded_subcat_train)
encoded_subcat_cv = np.asarray(encoded_subcat_cv)
encoded_subcat_test = np.asarray(encoded_subcat_test)

print(encoded_subcat_train.shape)
print(encoded_subcat_cv.shape)
print(encoded_subcat_test.shape)
```

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```
'esl literacy': 372, 'literacy specialneeds': 373, 'appliedsciences': 374,
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re_writing': 380, 'literature_writing mathematics': 381, 'literacy mathema
tics': 382, 'literacy': 383}
(53531,)
(22942,)
(32775,)
```

In [19]:

```
state_rank,unique,size =label_encoder(X_train['school_state'])
print(state_rank)
state_size = size
encoded_state_train = []
encoded_state_cv=[]
encoded_state_test = []
for state in X_train['school_state']:
    encoded_state_train.append(state_rank[state])
for state in X_cv['school_state']:
    if state in unique:
        encoded_state_cv.append(state_rank[state])
    else:
        encoded_state_cv.append(0)
for state in X_test['school_state']:
    if state in unique:
        encoded_state_test.append(state_rank[state])
    else:
        encoded_state_test.append(0)

encoded_state_train = np.asarray(encoded_state_train)
encoded_state_cv=np.asarray(encoded_state_cv)
encoded_state_test = np.asarray(encoded_state_test)

print(encoded_state_train.shape)
print(encoded_state_cv.shape)
print(encoded_state_test.shape)
```

```
{'vt': 1, 'wy': 2, 'nd': 3, 'ri': 4, 'mt': 5, 'sd': 6, 'ne': 7, 'nh': 8,
'ak': 9, 'de': 10, 'wv': 11, 'me': 12, 'hi': 13, 'dc': 14, 'nm': 15, 'ks':
16, 'id': 17, 'ia': 18, 'ar': 19, 'co': 20, 'mn': 21, 'or': 22, 'ms': 23,
'ky': 24, 'nv': 25, 'md': 26, 'ct': 27, 'ut': 28, 'tn': 29, 'al': 30, 'w
i': 31, 'va': 32, 'az': 33, 'ma': 34, 'ok': 35, 'nj': 36, 'la': 37, 'wa':
38, 'oh': 39, 'mo': 40, 'in': 41, 'pa': 42, 'mi': 43, 'sc': 44, 'ga': 45,
'il': 46, 'nc': 47, 'fl': 48, 'ny': 49, 'tx': 50, 'ca': 51}
(53531,)
(22942,)
(32775,)
```


In [20]:

```
teacher_prefix_rank, unique, size = label_encoder(X_train['teacher_prefix'])
print(teacher_prefix_rank)
teacher_prefix_size = size
encoded_prefix_train = []
encoded_prefix_cv=[]
encoded_prefix_test = []
for prefix in X_train['teacher_prefix']:
    encoded_prefix_train.append(teacher_prefix_rank[prefix])
for prefix in X_cv['teacher_prefix']:
    if prefix in unique:
        encoded_prefix_cv.append(teacher_prefix_rank[prefix])
    else:
        encoded_prefix_cv.append(0)

for prefix in X_test['teacher_prefix']:
    if prefix in unique:
        encoded_prefix_test.append(teacher_prefix_rank[prefix])
    else:
        encoded_prefix_test.append(0)

encoded_prefix_train = np.asarray(encoded_prefix_train)
encoded_prefix_cv=np.asarray(encoded_prefix_cv)
encoded_prefix_test = np.asarray(encoded_prefix_test)

print(encoded_prefix_train.shape)
print(encoded_prefix_cv.shape)
print(encoded_prefix_test.shape)
```

```
{'dr': 1, 'teacher': 2, 'mr': 3, 'ms': 4, 'mrs': 5}
(53531,)
(22942,)
(32775,)
```

In [21]:

```
project_grade_rank,unique,size =label_encoder(X_train['project_grade_category'])
print(project_grade_rank)
project_grade_categories_size = size
encoded_grade_train = []
encoded_grade_cv=[]
encoded_grade_test = []
for grade in X_train['project_grade_category']:
    encoded_grade_train.append(project_grade_rank[grade])
for grade in X_cv['project_grade_category']:
    if grade in unique:
        encoded_grade_cv.append(project_grade_rank[grade])
    else:
        encoded_grade_cv.append(0)

for grade in X_test['project_grade_category']:
    if grade in unique:
        encoded_grade_test.append(project_grade_rank[grade])
    else:
        encoded_grade_test.append(0)

encoded_grade_train = np.asarray(encoded_grade_train)
encoded_grade_cv=np.asarray(encoded_grade_cv)
encoded_grade_test = np.asarray(encoded_grade_test)

print(encoded_grade_train.shape)
print(encoded_grade_cv.shape)
print(encoded_grade_test.shape)
```

```
{'grades_9_12': 1, 'grades_6_8': 2, 'grades_3_5': 3, 'grades_prek_2': 4}
(53531,)
(22942,)
(32775,)
```

In [22]:

```
encoded_grade_train[0:5]
```

Out[22]:

```
array([2, 4, 3, 4, 2])
```

STANDARDIZING NUMERICAL FEATURES

In [23]:

```
# price

from sklearn.preprocessing import StandardScaler
scaler = StandardScaler()

scaler.fit(X_train['price'].values.reshape(-1,1))

X_train_price_norm = scaler.transform(X_train['price'].values.reshape(-1,1))
X_cv_price_norm = scaler.transform(X_cv['price'].values.reshape(-1,1))
X_test_price_norm = scaler.transform(X_test['price'].values.reshape(-1,1))

print("After vectorizations")
print(X_train_price_norm.shape, y_train.shape)
print(X_cv_price_norm.shape, y_cv.shape)
print(X_test_price_norm.shape, y_test.shape)
```

After vectorizations
 (53531, 1) (53531, 2)
 (22942, 1) (22942, 2)
 (32775, 1) (32775, 2)

In [24]:

```
scaler = StandardScaler()

scaler.fit(X_train['teacher_number_of_previously_posted_projects'].values.reshape(-1,1))

X_train_projects_norm = scaler.transform(X_train['teacher_number_of_previously_posted_projects'].values.reshape(-1,1))
X_cv_projects_norm = scaler.transform(X_cv['teacher_number_of_previously_posted_projects'].values.reshape(-1,1))
X_test_projects_norm = scaler.transform(X_test['teacher_number_of_previously_posted_projects'].values.reshape(-1,1))

print("After vectorizations")
print(X_train_projects_norm.shape, y_train.shape)
print(X_cv_projects_norm.shape, y_cv.shape)
print(X_test_projects_norm.shape, y_test.shape)
```

After vectorizations
 (53531, 1) (53531, 2)
 (22942, 1) (22942, 2)
 (32775, 1) (32775, 2)

In [0]:

```
left_input_train = np.hstack((X_train_price_norm, X_train_projects_norm))
left_input_cv = np.hstack((X_cv_price_norm, X_cv_projects_norm))
left_input_test = np.hstack((X_test_price_norm, X_test_projects_norm))
```

MODEL 1

In [0]:

```

essay_text=Input(shape=(300,),name='essay_text')
x=Embedding(vocab_size,300,weights=[embedding_matrix_train],input_length=300)(essay_text)
lstm_1=LSTM(50,recurrent_dropout=0.5,return_sequences=True)(x)
flatten_1=Flatten()(lstm_1)

state=Input(shape=(1,),name="state")
x=Embedding(state_size+1,2,input_length=1)(state)
flatten_2=Flatten()(x)

project_grade_category=Input(shape=(1,),name='project_grade_category')
x=Embedding(project_grade_categories_size+1,2,input_length=1)(project_grade_category)
flatten_3=Flatten()(x)

clean_categories=Input(shape=(1,),name='clean_categories')
x=Embedding(categories_size+1,4,input_length=1)(clean_categories)
flatten_4=Flatten()(x)

subcategory=Input(shape=(1,),name='subcategory')
x=Embedding(subcategories_size+1,4,input_length=1)(subcategory)
flatten_5=Flatten()(x)

teacher_prefix=Input(shape=(1,),name='teacher_prefix')
x=Embedding(teacher_prefix_size+1,4,input_length=1)(teacher_prefix)
flatten_6=Flatten()(x)

left_input=Input(shape=(2,),name='left_input')
dense_1 = Dense(1, activation='relu',kernel_initializer="he_normal",kernel_regularizer=
l2(0.001))(left_input)

x=concatenate([flatten_1,flatten_2,flatten_3,flatten_4,flatten_5,flatten_6,dense_1])

x = Dense(64, activation='relu',kernel_initializer="he_normal",kernel_regularizer=l2(0.
001))(x)
x = Dropout(.5)(x)
x = Dense(128, activation='relu',kernel_initializer="he_normal",kernel_regularizer=l2(
0.001))(x)
x = Dropout(.5)(x)
x = BatchNormalization()(x)

x = Dense(64, activation='relu',kernel_initializer="he_normal",kernel_regularizer=l2(0.
001))(x)
final_output = Dense(2, activation='softmax')(x)

model = Model(inputs=[essay_text,state,project_grade_category,clean_categories,subcateg
ory,teacher_prefix,left_input], outputs=[final_output])
print(model.summary())

```

Model: "model_2"

Layer (type) connected to	Output Shape	Param #	Connect
essay_text (InputLayer)	(None, 300)	0	
embedding_7 (Embedding) ext[0][0]	(None, 300, 300)	12768900	essay_t
state (InputLayer)	(None, 1)	0	
project_grade_category (InputLa	(None, 1)	0	
clean_categories (InputLayer)	(None, 1)	0	
subcategory (InputLayer)	(None, 1)	0	
teacher_prefix (InputLayer)	(None, 1)	0	
lstm_2 (LSTM) ng_7[0][0]	(None, 300, 50)	70200	embeddi
embedding_8 (Embedding) [0][0]	(None, 1, 2)	104	state
embedding_9 (Embedding) _grade_category[0][0]	(None, 1, 2)	10	project
embedding_10 (Embedding) ategories[0][0]	(None, 1, 4)	204	clean_c
embedding_11 (Embedding) gory[0][0]	(None, 1, 4)	1540	subcate
embedding_12 (Embedding) _prefix[0][0]	(None, 1, 4)	24	teacher
left_input (InputLayer)	(None, 2)	0	
flatten_7 (Flatten) [0][0]	(None, 15000)	0	lstm_2
flatten_8 (Flatten)	(None, 2)	0	embeddi

ng_8[0][0]

flatten_9 (Flatten) ng_9[0][0]	(None, 2)	0	embeddi
-----------------------------------	-----------	---	---------

flatten_10 (Flatten) ng_10[0][0]	(None, 4)	0	embeddi
-------------------------------------	-----------	---	---------

flatten_11 (Flatten) ng_11[0][0]	(None, 4)	0	embeddi
-------------------------------------	-----------	---	---------

flatten_12 (Flatten) ng_12[0][0]	(None, 4)	0	embeddi
-------------------------------------	-----------	---	---------

dense_6 (Dense) put[0][0]	(None, 1)	3	left_in
------------------------------	-----------	---	---------

concatenate_2 (Concatenate) _7[0][0]	(None, 15017)	0	flatten
_8[0][0]			flatten
_9[0][0]			flatten
_10[0][0]			flatten
_11[0][0]			flatten
_12[0][0]			flatten
[0][0]			dense_6

dense_7 (Dense) nate_2[0][0]	(None, 64)	961152	concate
---------------------------------	------------	--------	---------

dropout_3 (Dropout) [0][0]	(None, 64)	0	dense_7
-------------------------------	------------	---	---------

dense_8 (Dense) _3[0][0]	(None, 128)	8320	dropout
-----------------------------	-------------	------	---------

dropout_4 (Dropout) [0][0]	(None, 128)	0	dense_8
-------------------------------	-------------	---	---------

batch_normalization_2 (BatchNor _4[0][0]	(None, 128)	512	dropout
---	-------------	-----	---------

dense_9 (Dense) ormalization_2[0][0]	(None, 64)	8256	batch_n
---	------------	------	---------

```

dense_10 (Dense)          (None, 2)          130          dense_9
[0][0]
=====
=====
Total params: 13,819,355
Trainable params: 13,819,099
Non-trainable params: 256

```

None

In [0]:

```

import tensorflow as tf
from sklearn.metrics import roc_auc_score

def auROC(y_true, y_pred):
    return tf.py_func(roc_auc_score, (y_true, y_pred), tf.double)

```

In [0]:

```

train_data = [padded_train,encoded_state_train,encoded_grade_train,encoded_cat_train,en
coded_subcat_train,encoded_prefix_train,left_input_train]
cv_data = [padded_cv,encoded_state_cv,encoded_grade_cv,encoded_cat_cv,encoded_subcat_cv
,encoded_prefix_cv,left_input_cv]
test_data = [padded_test,encoded_state_test,encoded_grade_test,encoded_cat_test,encoded
_subcat_test,encoded_prefix_test,left_input_test]

```

In [0]:

```

from keras.callbacks import TensorBoard
checkpoint_1 = ModelCheckpoint("model_1.1",
                              monitor="val_auroc",
                              mode="max",
                              save_best_only = True,
                              verbose=1)
earlystop = EarlyStopping(monitor = 'val_auroc',
                           mode="max",
                           min_delta = 0,
                           patience = 2,
                           verbose = 1,)
tensorboard=TensorBoard(log_dir='lstm_1',batch_size=512)
callbacks_1=[checkpoint_1,earlystop,tensorboard]

```

In [0]:

```
model.compile(optimizer='adam', loss='categorical_crossentropy', metrics=[auroc])
LSTM_1 = model.fit(train_data, y_train, batch_size=512, epochs=10, verbose=1, callbacks=
callbacks_1, validation_data=(cv_data, y_cv))
```

Train on 53531 samples, validate on 22942 samples

Epoch 1/10

53531/53531 [=====] - 456s 9ms/step - loss: 0.947

4 - auroc: 0.5201 - val_loss: 0.7758 - val_auroc: 0.4897

Epoch 00001: val_auroc improved from -inf to 0.48968, saving model to model_1.1

Epoch 2/10

53531/53531 [=====] - 455s 9ms/step - loss: 0.711

2 - auroc: 0.6025 - val_loss: 0.7190 - val_auroc: 0.7081

Epoch 00002: val_auroc improved from 0.48968 to 0.70813, saving model to model_1.1

Epoch 3/10

53531/53531 [=====] - 450s 8ms/step - loss: 0.590

6 - auroc: 0.7181 - val_loss: 0.6186 - val_auroc: 0.7448

Epoch 00003: val_auroc improved from 0.70813 to 0.74482, saving model to model_1.1

Epoch 4/10

53531/53531 [=====] - 449s 8ms/step - loss: 0.517

5 - auroc: 0.7632 - val_loss: 0.5965 - val_auroc: 0.7520

Epoch 00004: val_auroc improved from 0.74482 to 0.75200, saving model to model_1.1

Epoch 5/10

53531/53531 [=====] - 448s 8ms/step - loss: 0.462

6 - auroc: 0.8038 - val_loss: 0.5980 - val_auroc: 0.7385

Epoch 00005: val_auroc did not improve from 0.75200

Epoch 6/10

53531/53531 [=====] - 447s 8ms/step - loss: 0.411

1 - auroc: 0.8456 - val_loss: 0.5612 - val_auroc: 0.7269

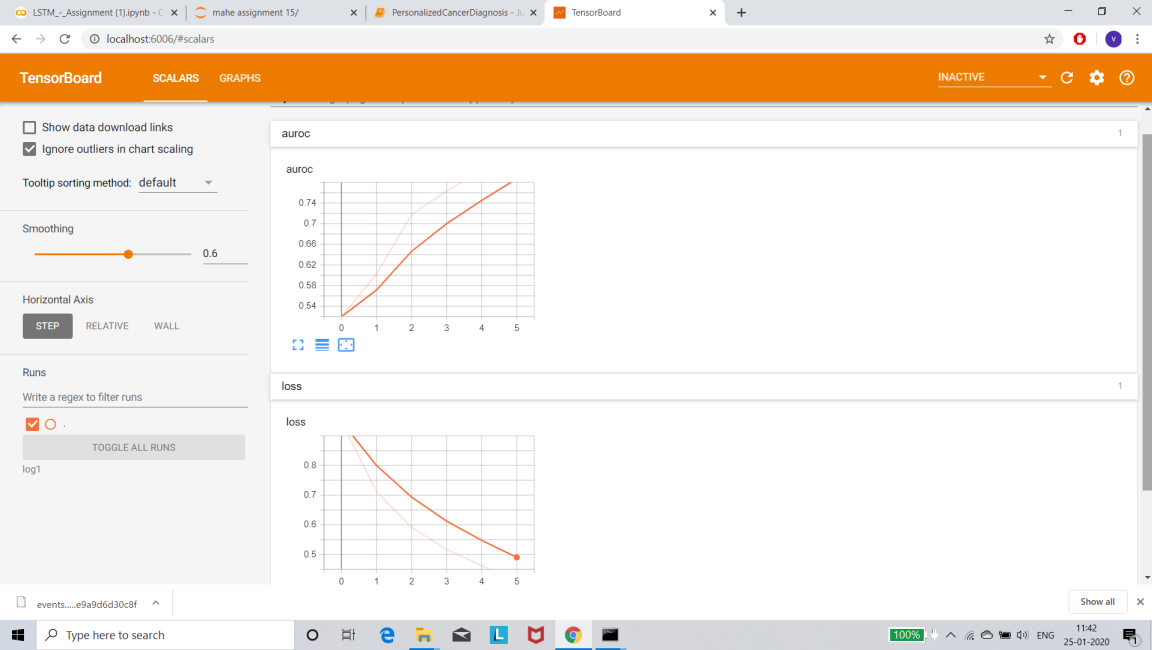
Epoch 00006: val_auroc did not improve from 0.75200

Epoch 00006: early stopping

In [0]:

```
Image(retina=True, filename='/content/Screenshot (26).png')
```

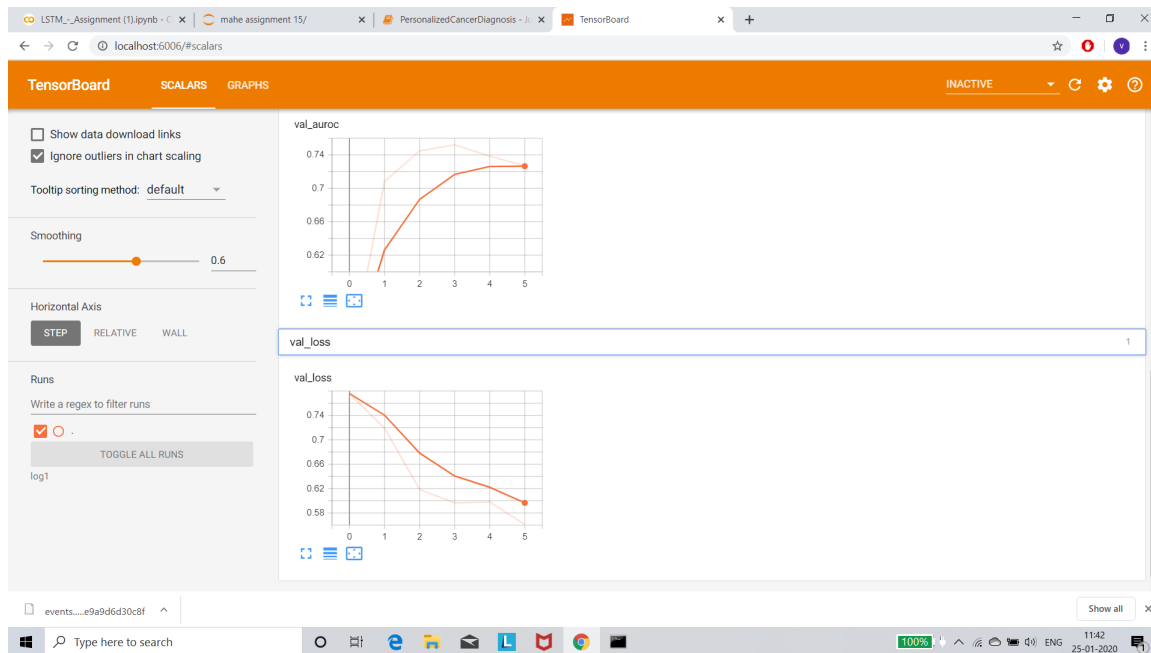
Out[0]:



In [0]:

```
Image(retina=True, filename='/content/Screenshot (27).png')
```

Out[0]:



MODEL 2

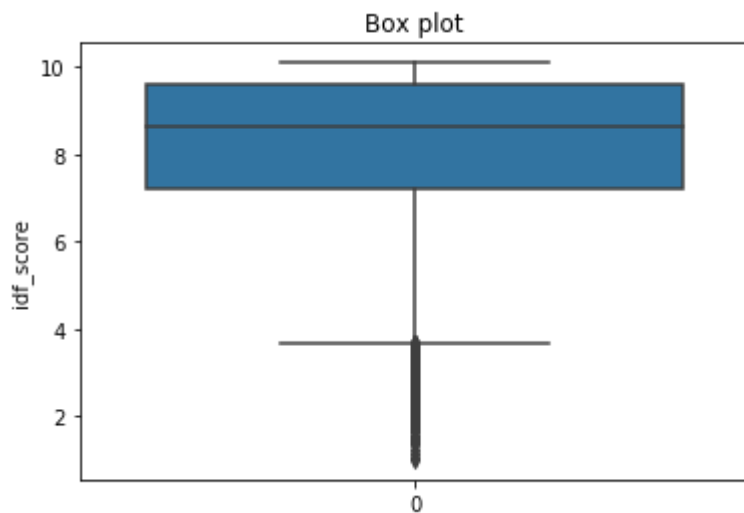
In [26]:

```
vectorizer = TfidfVectorizer(min_df=5)
X_train_essay_tfidf = vectorizer.fit_transform(X_train['essay'].values)
print("After vectorizations")
print(X_train_essay_tfidf.shape, y_train.shape)
```

After vectorizations
(53531, 16961) (53531, 2)

In [27]:

```
idf = vectorizer.idf_  
sns.boxplot(data=idf)  
plt.title('Box plot')  
plt.ylabel('idf_score')  
plt.show()
```



In [28]:

```
len(idf)
```

Out[28]:

16961

In [0]:

```
feature_names = np.asarray(vectorizer.get_feature_names())  
index = []  
for i in range(len(idf)):  
    if idf[i] >= 2 and idf[i] <=10:  
        index.append(i)  
imp_words = []  
for i in index:  
    imp_words.append(feature_names[i])
```

In [30]:

```
print('total words=',len(feature_names))  
print('important words=',len(imp_words))
```

total words= 16961
important words= 15665

In [31]:

```
# train_data
X_train_essay_imp = []
for essay in X_train['essay']:
    sentence = []
    for word in essay.split():
        if word in imp_words:
            sentence.append(word)
    X_train_essay_imp.append(' '.join(sentence))
print(len(X_train_essay_imp))
X_cv_essay_imp = []
for essay in X_cv['essay']:
    sentence = []
    for word in essay.split():
        if word in imp_words:
            sentence.append(word)
    X_cv_essay_imp.append(' '.join(sentence))
print(len(X_cv_essay_imp))
X_test_essay_imp = []
for essay in X_test['essay']:
    sentence = []
    for word in essay.split():
        if word in imp_words:
            sentence.append(word)
    X_test_essay_imp.append(' '.join(sentence))
print(len(X_test_essay_imp))
```

53531
22942
32775

In [32]:

```
print(len(X_train_essay_imp))
print(len(X_cv_essay_imp))
print(len(X_test_essay_imp))
```

53531
22942
32775

In [33]:

```
token_2 = Tokenizer()
token_2.fit_on_texts(X_train_essay_imp)
imp_vocab = len(token_2.word_index) + 1
print('Total unique words in the important words', imp_vocab)
encoded_train_imp = token_2.texts_to_sequences(X_train_essay_imp)
encoded_cv_imp = token_2.texts_to_sequences(X_cv_essay_imp)
encoded_test_imp = token_2.texts_to_sequences(X_test_essay_imp)
print(len(encoded_train_imp))
print(len(encoded_cv_imp))
print(len(encoded_test_imp))
```

Total unique words in the important words 15666
53531
22942
32775

In [34]:

```
max_length = 300
padded_train_imp = pad_sequences(encoded_train_imp, maxlen=max_length, padding='post')
padded_cv_imp = pad_sequences(encoded_cv_imp, maxlen=max_length, padding='post')
padded_test_imp = pad_sequences(encoded_test_imp, maxlen=max_length, padding='post')
print("length of padded_train_new data", len(padded_train_imp))
print("length of padded_cv_new data", len(padded_cv_imp))
print("length of padded_test_new data", len(padded_test_imp))
```

length of padded_train_new data 53531

length of padded_cv_new data 22942

length of padded_test_new data 32775

In [0]:

```
with open('glove_vectors', 'rb') as f:
    model = pickle.load(f)
    glove_words = set(model.keys())

# for train
embedding_matrix_train2 = np.zeros((imp_vocab, 300))
for word, i in token_2.word_index.items():
    if word in glove_words:
        embedding_vector = model[word]
        embedding_matrix_train2[i] = embedding_vector
```

In [37]:

```

essay_text=Input(shape=(300,),name='essay_text')
x=Embedding(imp_vocab,300,weights=[embedding_matrix_train2],input_length=300)(essay_text)
lstm_1=LSTM(50,recurrent_dropout=0.5,return_sequences=True)(x)
flatten_1=Flatten()(lstm_1)

state=Input(shape=(1,),name="state")
x=Embedding(state_size+1,2,input_length=1)(state)
flatten_2=Flatten()(x)

project_grade_category=Input(shape=(1,),name='project_grade_category')
x=Embedding(project_grade_categories_size+1,2,input_length=1)(project_grade_category)
flatten_3=Flatten()(x)

clean_categories=Input(shape=(1,),name='clean_categories')
x=Embedding(categories_size+1,4,input_length=1)(clean_categories)
flatten_4=Flatten()(x)

subcategory=Input(shape=(1,),name='subcategory')
x=Embedding(subcategories_size+1,4,input_length=1)(subcategory)
flatten_5=Flatten()(x)

teacher_prefix=Input(shape=(1,),name='teacher_prefix')
x=Embedding(teacher_prefix_size+1,4,input_length=1)(teacher_prefix)
flatten_6=Flatten()(x)

left_input=Input(shape=(2,),name='left_input')
dense_1 = Dense(1, activation='relu',kernel_initializer="he_normal",kernel_regularizer=
l2(0.001))(left_input)

x=concatenate([flatten_1,flatten_2,flatten_3,flatten_4,flatten_5,flatten_6,dense_1])

x = Dense(32, activation='relu',kernel_initializer="he_normal",kernel_regularizer=l2(0.
001))(x)
x = Dropout(.5)(x)
x = Dense(64, activation='relu',kernel_initializer="he_normal",kernel_regularizer=l2(0.
001))(x)
x = Dropout(.5)(x)
x = BatchNormalization()(x)

x = Dense(64, activation='relu',kernel_initializer="he_normal",kernel_regularizer=l2(0.
001))(x)
final_output = Dense(2, activation='softmax')(x)

model = Model(inputs=[essay_text,state,project_grade_category,clean_categories,subcateg
ory,teacher_prefix,left_input], outputs=[final_output])
print(model.summary())

```

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflow_backend.py:541: The name tf.placeholder is deprecated. Please use tf.compat.v1.placeholder instead.

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflow_backend.py:66: The name tf.get_default_graph is deprecated. Please use tf.compat.v1.get_default_graph instead.

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflow_backend.py:4432: The name tf.random_uniform is deprecated. Please use tf.random.uniform instead.

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflow_backend.py:190: The name tf.get_default_session is deprecated. Please use tf.compat.v1.get_default_session instead.

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflow_backend.py:197: The name tf.ConfigProto is deprecated. Please use tf.compat.v1.ConfigProto instead.

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflow_backend.py:203: The name tf.Session is deprecated. Please use tf.compat.v1.Session instead.

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflow_backend.py:207: The name tf.global_variables is deprecated. Please use tf.compat.v1.global_variables instead.

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflow_backend.py:216: The name tf.is_variable_initialized is deprecated. Please use tf.compat.v1.is_variable_initialized instead.

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflow_backend.py:223: The name tf.variables_initializer is deprecated. Please use tf.compat.v1.variables_initializer instead.

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflow_backend.py:148: The name tf.placeholder_with_default is deprecated. Please use tf.compat.v1.placeholder_with_default instead.

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflow_backend.py:3733: calling dropout (from tensorflow.python.ops.nn_ops) with keep_prob is deprecated and will be removed in a future version.

Instructions for updating:

Please use `rate` instead of `keep_prob`. Rate should be set to `rate = 1 - keep_prob`.

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflow_backend.py:4479: The name tf.truncated_normal is deprecated. Please use tf.random.truncated_normal instead.

Model: "model_1"

Layer (type)	Output Shape	Param #	Connected to
essay_text (InputLayer)	(None, 300)	0	

embedding_1 (Embedding) ext[0][0]	(None, 300, 300)	4699800	essay_t
state (InputLayer)	(None, 1)	0	
project_grade_category (InputLa	(None, 1)	0	
clean_categories (InputLayer)	(None, 1)	0	
subcategory (InputLayer)	(None, 1)	0	
teacher_prefix (InputLayer)	(None, 1)	0	
lstm_1 (LSTM) ng_1[0][0]	(None, 300, 50)	70200	embeddi
embedding_2 (Embedding) [0][0]	(None, 1, 2)	104	state
embedding_3 (Embedding) _grade_category[0][0]	(None, 1, 2)	10	project
embedding_4 (Embedding) ategories[0][0]	(None, 1, 4)	204	clean_c
embedding_5 (Embedding) gory[0][0]	(None, 1, 4)	1536	subcate
embedding_6 (Embedding) _prefix[0][0]	(None, 1, 4)	24	teacher
left_input (InputLayer)	(None, 2)	0	
flatten_1 (Flatten) [0][0]	(None, 15000)	0	lstm_1
flatten_2 (Flatten) ng_2[0][0]	(None, 2)	0	embeddi
flatten_3 (Flatten) ng_3[0][0]	(None, 2)	0	embeddi
flatten_4 (Flatten) ng_4[0][0]	(None, 4)	0	embeddi

flatten_5 (Flatten) ng_5[0][0]	(None, 4)	0	embeddi
flatten_6 (Flatten) ng_6[0][0]	(None, 4)	0	embeddi
dense_1 (Dense) put[0][0]	(None, 1)	3	left_in
concatenate_1 (Concatenate) _1[0][0]	(None, 15017)	0	flatten
_2[0][0]			flatten
_3[0][0]			flatten
_4[0][0]			flatten
_5[0][0]			flatten
_6[0][0]			flatten
[0][0]			dense_1
dense_2 (Dense) nate_1[0][0]	(None, 32)	480576	concate
dropout_1 (Dropout) [0][0]	(None, 32)	0	dense_2
dense_3 (Dense) _1[0][0]	(None, 64)	2112	dropout
dropout_2 (Dropout) [0][0]	(None, 64)	0	dense_3
batch_normalization_1 (BatchNor _2[0][0]	(None, 64)	256	dropout
dense_4 (Dense) ormalization_1[0][0]	(None, 64)	4160	batch_n
dense_5 (Dense) [0][0]	(None, 2)	130	dense_4
=====			
=====			
Total params: 5,259,115			
Trainable params: 5,258,987			
Non-trainable params: 128			

None

In [0]:

```
from keras.callbacks import TensorBoard
checkpoint_2 = ModelCheckpoint("model_2.1",
                              monitor="val_auroc",
                              mode="max",
                              save_best_only = True,
                              verbose=1)
earlystop_2 = EarlyStopping(monitor = 'val_auroc',
                             mode="max",
                             min_delta = 0,
                             patience = 2,
                             verbose = 1,)
tensorboard_1=TensorBoard(log_dir='lstm_2',batch_size=512)
callbacks_2=[checkpoint_2,earlystop_2,tensorboard_1]
```

In [0]:

```
train_data_1 = [padded_train_imp,encoded_state_train,encoded_grade_train,encoded_cat_train,encoded_subcat_train,encoded_prefix_train,left_input_train]
cv_data_1 = [padded_cv_imp,encoded_state_cv,encoded_grade_cv,encoded_cat_cv,encoded_subcat_cv,encoded_prefix_cv,left_input_cv]
test_data_1 = [padded_test_imp,encoded_state_test,encoded_grade_test,encoded_cat_test,encoded_subcat_test,encoded_prefix_test,left_input_test]
```

In [42]:

```
model.compile(optimizer='adam', loss='categorical_crossentropy', metrics=[auroc])  
LSTM_2 = model.fit(train_data_1, y_train, batch_size=512, epochs=10, verbose=1, callbacks=callbacks_2, validation_data=(cv_data_1, y_cv))
```

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/optimizers.py:793: The name tf.train.Optimizer is deprecated. Please use tf.compat.v1.train.Optimizer instead.

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflow_backend.py:3576: The name tf.log is deprecated. Please use tf.math.log instead.

WARNING:tensorflow:From <ipython-input-41-4a25250c5bd7>:5: py_func (from tensorflow.python.ops.script_ops) is deprecated and will be removed in a future version.

Instructions for updating:

tf.py_func is deprecated in TF V2. Instead, there are two options available in V2.

- tf.py_function takes a python function which manipulates tf eager tensors instead of numpy arrays. It's easy to convert a tf eager tensor to an ndarray (just call tensor.numpy()) but having access to eager tensors means `tf.py_function`s can use accelerators such as GPUs as well as being differentiable using a gradient tape.
- tf.numpy_function maintains the semantics of the deprecated tf.py_func (it is not differentiable, and manipulates numpy arrays). It drops the stateful argument making all functions stateful.

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/tensorflow_core/python/ops/math_grad.py:1424: where (from tensorflow.python.ops.array_ops) is deprecated and will be removed in a future version.

Instructions for updating:

Use tf.where in 2.0, which has the same broadcast rule as np.where

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflow_backend.py:1033: The name tf.assign_add is deprecated. Please use tf.compat.v1.assign_add instead.

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflow_backend.py:1020: The name tf.assign is deprecated. Please use tf.compat.v1.assign instead.

Train on 53531 samples, validate on 22942 samples

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/callbacks.py:1122: The name tf.summary.merge_all is deprecated. Please use tf.compat.v1.summary.merge_all instead.

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/callbacks.py:1125: The name tf.summary.FileWriter is deprecated. Please use tf.compat.v1.summary.FileWriter instead.

Epoch 1/10

53531/53531 [=====] - 386s 7ms/step - loss: 0.7865 - auroc: 0.5264 - val_loss: 0.6874 - val_auroc: 0.5568

Epoch 00001: val_auroc improved from -inf to 0.55678, saving model to model_2.1

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/callbacks.py:1265: The name tf.Summary is deprecated. Please use tf.compat.v1.Summary instead.

Epoch 2/10

53531/53531 [=====] - 385s 7ms/step - loss: 0.6209 - auroc: 0.6278 - val_loss: 0.6117 - val_auroc: 0.7106

Epoch 00002: val_auroc improved from 0.55678 to 0.71064, saving model to model_2.1

Epoch 3/10

53531/53531 [=====] - 390s 7ms/step - loss: 0.5425 - auroc: 0.7083 - val_loss: 0.6237 - val_auroc: 0.7147

Epoch 00003: val_auroc improved from 0.71064 to 0.71470, saving model to model_2.1

Epoch 4/10

53531/53531 [=====] - 391s 7ms/step - loss: 0.4845 - auroc: 0.7545 - val_loss: 0.5351 - val_auroc: 0.7470

Epoch 00004: val_auroc improved from 0.71470 to 0.74695, saving model to model_2.1

Epoch 5/10

53531/53531 [=====] - 390s 7ms/step - loss: 0.4419 - auroc: 0.7884 - val_loss: 0.5426 - val_auroc: 0.7382

Epoch 00005: val_auroc did not improve from 0.74695

Epoch 6/10

53531/53531 [=====] - 389s 7ms/step - loss: 0.4017 - auroc: 0.8239 - val_loss: 0.5385 - val_auroc: 0.7280

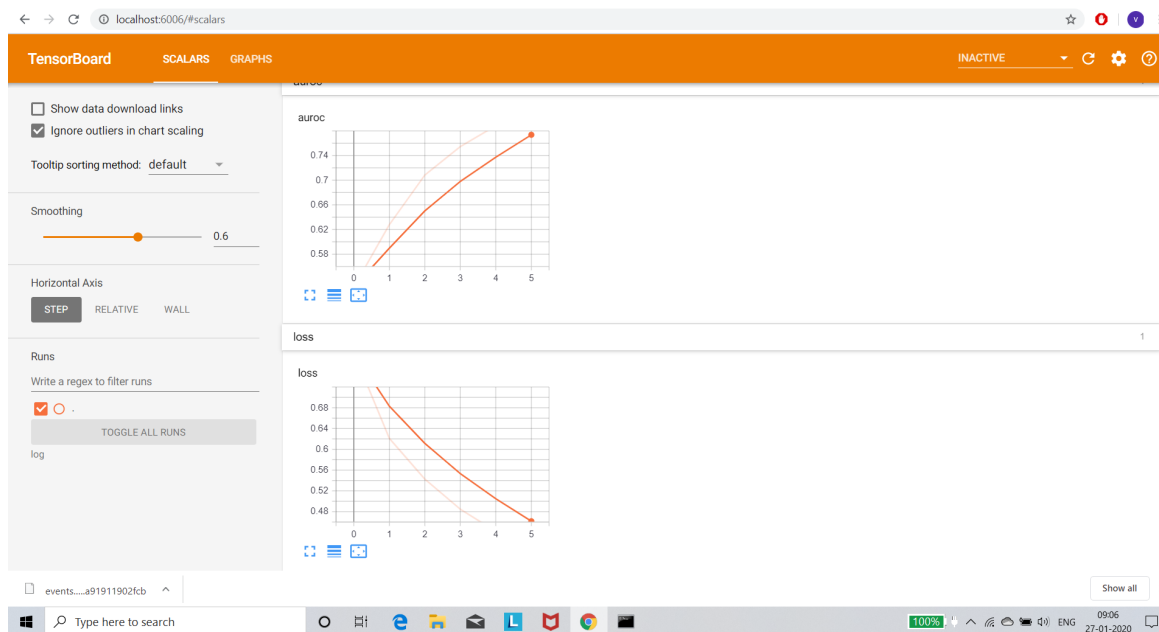
Epoch 00006: val_auroc did not improve from 0.74695

Epoch 00006: early stopping

In [44]:

```
Image(retina=True, filename='/content/Screenshot (32).png')
```

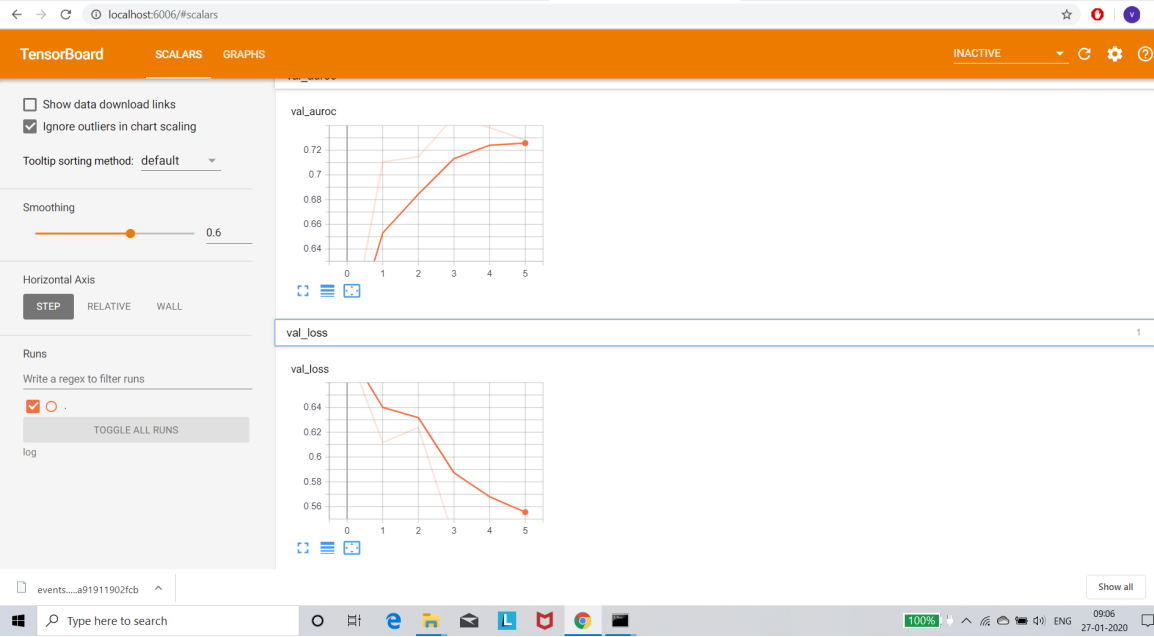
Out[44]:



In [45]:

```
Image(retina=True, filename='/content/Screenshot (33).png')
```

Out[45]:



In [0]:

MODEL 3

ONE HOT ENCODING OF CATEGORICAL VARIABLES

In [0]:

```

vectorizer = CountVectorizer()
vectorizer.fit(X_train['clean_categories'].values)

X_train_categories_one_hot = vectorizer.transform(X_train['clean_categories'].values)
X_cv_categories_one_hot = vectorizer.transform(X_cv['clean_categories'].values)
X_test_categories_one_hot = vectorizer.transform(X_test['clean_categories'].values)
print(vectorizer.get_feature_names())
print("-"*100)
print("After vectorizations")
print(X_train_categories_one_hot.shape, y_train.shape)
print(X_cv_categories_one_hot.shape, y_cv.shape)
print(X_test_categories_one_hot.shape, y_test.shape)

```

```

['appliedlearning', 'care_hunger', 'health_sports', 'history_civics', 'literacy_language', 'math_science', 'music_arts', 'specialneeds', 'warmth']
-----
-----

```

```

After vectorizations
(53531, 9) (53531, 2)
(22942, 9) (22942, 2)
(32775, 9) (32775, 2)

```

In [0]:

```

vectorizer = CountVectorizer()
vectorizer.fit(X_train['clean_subcategories'].values)

X_train_subcategories_one_hot = vectorizer.transform(X_train['clean_subcategories'].values)
X_cv_subcategories_one_hot = vectorizer.transform(X_cv['clean_subcategories'].values)
X_test_subcategories_one_hot = vectorizer.transform(X_test['clean_subcategories'].values)
print(vectorizer.get_feature_names())
print("-"*100)
print("After vectorizations")
print(X_train_subcategories_one_hot.shape, y_train.shape)
print(X_cv_subcategories_one_hot.shape, y_cv.shape)
print(X_test_subcategories_one_hot.shape, y_test.shape)

```

```

['appliedsciences', 'care_hunger', 'charactereducation', 'civics_government', 'college_careerprep', 'communityservice', 'earlydevelopment', 'economics', 'environmentalscience', 'esl', 'extracurricular', 'financialliteracy', 'foreignlanguages', 'gym_fitness', 'health_lifescience', 'health_wellness', 'history_geography', 'literacy', 'literature_writing', 'mathematics', 'music', 'nutritioneducation', 'other', 'parentinvolvement', 'performingarts', 'socialsciences', 'specialneeds', 'teamsports', 'visualarts', 'warmth']
-----
-----

```

```

After vectorizations
(53531, 30) (53531, 2)
(22942, 30) (22942, 2)
(32775, 30) (32775, 2)

```

In [0]:

X_train.columns

Out[0]:

```
Index(['school_state', 'teacher_prefix', 'project_grade_category',
      'teacher_number_of_previously_posted_projects', 'clean_categories',
      'clean_subcategories', 'essay', 'price'],
      dtype='object')
```

In [0]:

```
vectorizer = CountVectorizer()
vectorizer.fit(X_train['teacher_prefix'].values)

X_train_prefix_one_hot = vectorizer.transform(X_train['teacher_prefix'].values)
X_cv_prefix_one_hot = vectorizer.transform(X_cv['teacher_prefix'].values)
X_test_prefix_one_hot = vectorizer.transform(X_test['teacher_prefix'].values)
print(vectorizer.get_feature_names())
print("-"*100)
print("After vectorizations")
print(X_train_prefix_one_hot.shape, y_train.shape)
print(X_cv_prefix_one_hot.shape, y_cv.shape)
print(X_test_prefix_one_hot.shape, y_test.shape)
```

```
['dr', 'mr', 'mrs', 'ms', 'teacher']
```

```
-----
After vectorizations
(53531, 5) (53531, 2)
(22942, 5) (22942, 2)
(32775, 5) (32775, 2)
```

In [0]:

```
vectorizer = CountVectorizer()
vectorizer.fit(X_train['school_state'].values)

X_train_state_one_hot = vectorizer.transform(X_train['school_state'].values)
X_cv_state_one_hot = vectorizer.transform(X_cv['school_state'].values)
X_test_state_one_hot = vectorizer.transform(X_test['school_state'].values)
print(vectorizer.get_feature_names())
print("-"*100)
print("After vectorizations")
print(X_train_state_one_hot.shape, y_train.shape)
print(X_cv_state_one_hot.shape, y_cv.shape)
print(X_test_state_one_hot.shape, y_test.shape)
```

```
['ak', 'al', 'ar', 'az', 'ca', 'co', 'ct', 'dc', 'de', 'fl', 'ga', 'hi',
 'ia', 'id', 'il', 'in', 'ks', 'ky', 'la', 'ma', 'md', 'me', 'mi', 'mn', 'm
o', 'ms', 'mt', 'nc', 'nd', 'ne', 'nh', 'nj', 'nm', 'nv', 'ny', 'oh', 'o
k', 'or', 'pa', 'ri', 'sc', 'sd', 'tn', 'tx', 'ut', 'va', 'vt', 'wa', 'w
i', 'wv', 'wy']
```

```
-----
After vectorizations
(53531, 51) (53531, 2)
(22942, 51) (22942, 2)
(32775, 51) (32775, 2)
```


In [0]:

```

vectorizer = CountVectorizer()
vectorizer.fit(X_train['project_grade_category'].values)

X_train_grade_one_hot = vectorizer.transform(X_train['project_grade_category'].values)
X_cv_grade_one_hot = vectorizer.transform(X_cv['project_grade_category'].values)
X_test_grade_one_hot = vectorizer.transform(X_test['project_grade_category'].values)
print(vectorizer.get_feature_names())
print("-"*100)
print("After vectorizations")
print(X_train_grade_one_hot.shape, y_train.shape)
print(X_cv_grade_one_hot.shape, y_cv.shape)
print(X_test_grade_one_hot.shape, y_test.shape)

```

```
['grades_3_5', 'grades_6_8', 'grades_9_12', 'grades_prek_2']
```

```
-----
```

```

After vectorizations
(53531, 4) (53531, 2)
(22942, 4) (22942, 2)
(32775, 4) (32775, 2)

```

CREATING INPUT DATA OTHER THAN TEXT

In [0]:

```

train_without_text=hstack((X_train_categories_one_hot,X_train_prefix_one_hot,X_train_state_one_hot,X_train_subcategories_one_hot,X_train_grade_one_hot,X_train_price_norm,X_train_projects_norm)).todense()
cv_without_text=hstack((X_cv_categories_one_hot,X_cv_prefix_one_hot,X_cv_state_one_hot,X_cv_subcategories_one_hot,X_cv_grade_one_hot,X_cv_price_norm,X_cv_projects_norm)).todense()
test_without_text=hstack((X_test_categories_one_hot,X_test_prefix_one_hot,X_test_state_one_hot,X_test_subcategories_one_hot,X_test_grade_one_hot,X_test_price_norm,X_test_projects_norm)).todense()
print(train_without_text.shape)
print(cv_without_text.shape)
print(test_without_text.shape)

```

```

(53531, 101)
(22942, 101)
(32775, 101)

```

In [0]:

```

train_final = np.expand_dims(train_without_text,2)
cv_final = np.expand_dims(cv_without_text,2)
test_final=np.expand_dims(test_without_text,2)
print(train_final.shape)
print(cv_final.shape)
print(test_final.shape)

```

```

(53531, 101, 1)
(22942, 101, 1)
(32775, 101, 1)

```

MODEL 3

In [0]:

```
essay_text=Input(shape=(300,),name='essay_text')
x=Embedding(vocab_size,300,weights=[embedding_matrix_train],input_length=300)(essay_text)
lstm_1=LSTM(50,recurrent_dropout=0.5,return_sequences=True)(x)
flatten_1=Flatten()(lstm_1)

other_features=Input(shape=(101,1),name='other_features')
x = Conv1D(filters=128, kernel_size = 2, padding='valid', kernel_initializer='he_normal')(other_features)
x = Conv1D(filters=128, kernel_size = 2, padding='valid', kernel_initializer='he_normal')(x)
x = Flatten()(x)

con=concatenate([flatten_1,x])

x=Dense(128, activation='relu',kernel_initializer="he_normal",kernel_regularizer=l2(0.001))(con)
x=Dropout(0.5)(x)
x=Dense(128,activation='relu',kernel_initializer="he_normal",kernel_regularizer=l2(0.001))(x)
x=Dropout(0.5)(x)
x=BatchNormalization()(x)
x=Dense(128, activation='relu',kernel_initializer="he_normal",kernel_regularizer=l2(0.001))(x)

final_output = Dense(2, activation='softmax')(x)

model_3 = Model(inputs=[essay_text,other_features], outputs=[final_output])
print(model_3.summary())
```

Model: "model_3"

Layer (type) to	Output Shape	Param #	Connected
essay_text (InputLayer)	(None, 300)	0	
other_features (InputLayer)	(None, 101, 1)	0	
embedding_13 (Embedding) t[0][0]	(None, 300, 300)	12768900	essay_text[0][0]
conv1d_1 (Conv1D) tures[0][0]	(None, 100, 128)	384	other_features[0][0]
lstm_3 (LSTM) _13[0][0]	(None, 300, 50)	70200	embedding_13[0][0]
conv1d_2 (Conv1D) [0][0]	(None, 99, 128)	32896	conv1d_1[0][0]
flatten_13 (Flatten) [0]	(None, 15000)	0	lstm_3[0][0]
flatten_14 (Flatten) [0][0]	(None, 12672)	0	conv1d_2[0][0]
concatenate_3 (Concatenate) 3[0][0]	(None, 27672)	0	flatten_13[0][0] flatten_14[0][0]
dense_11 (Dense) te_3[0][0]	(None, 128)	3542144	concatenate_3[0][0]
dropout_5 (Dropout) [0][0]	(None, 128)	0	dense_11[0][0]
dense_12 (Dense) [0][0]	(None, 128)	16512	dropout_5[0][0]
dropout_6 (Dropout) [0][0]	(None, 128)	0	dense_12[0][0]
batch_normalization_3 (BatchNor [0][0]	(None, 128)	512	dropout_6[0][0]

dense_13 (Dense) malization_3[0][0]	(None, 128)	16512	batch_norm
--	-------------	-------	------------

dense_14 (Dense) [0][0]	(None, 2)	258	dense_13
----------------------------	-----------	-----	----------

```
=====
Total params: 16,448,318
Trainable params: 16,448,062
Non-trainable params: 256
```

None



In [0]:

```
from keras.callbacks import TensorBoard
checkpoint_3 = ModelCheckpoint("model_3.1",
                              monitor="val_auroc",
                              mode="max",
                              save_best_only = True,
                              verbose=1)
earlystop = EarlyStopping(monitor = 'val_auroc',
                           mode="max",
                           min_delta = 0,
                           patience = 3,
                           verbose = 1,)
tensorboard_2=TensorBoard(log_dir='lstm_3',batch_size=512)
callbacks_2=[checkpoint_3,earlystop,tensorboard_2]
```

In [0]:

```
train_model_3=[padded_train,train_final]
cv_model_3=[padded_cv,cv_final]
test_model_3=[padded_test,test_final]
```

In [0]:

```
model_3.compile(optimizer='adam', loss='categorical_crossentropy', metrics=[auroc])
LSTM_3 = model_3.fit(train_model_3, y_train, batch_size=512, epochs=10, verbose=1, callbacks=callbacks_2, validation_data=(cv_model_3, y_cv))
```

Train on 53531 samples, validate on 22942 samples

Epoch 1/10

53531/53531 [=====] - 509s 10ms/step - loss: 1.1905 - auroc: 0.5222 - val_loss: 0.9167 - val_auroc: 0.5946

Epoch 00001: val_auroc improved from -inf to 0.59462, saving model to model_3.1

Epoch 2/10

53531/53531 [=====] - 504s 9ms/step - loss: 0.8501 - auroc: 0.5629 - val_loss: 0.7516 - val_auroc: 0.6469

Epoch 00002: val_auroc improved from 0.59462 to 0.64692, saving model to model_3.1

Epoch 3/10

53531/53531 [=====] - 496s 9ms/step - loss: 0.7080 - auroc: 0.6658 - val_loss: 0.6449 - val_auroc: 0.7220

Epoch 00003: val_auroc improved from 0.64692 to 0.72202, saving model to model_3.1

Epoch 4/10

53531/53531 [=====] - 495s 9ms/step - loss: 0.6096 - auroc: 0.7355 - val_loss: 0.6054 - val_auroc: 0.7544

Epoch 00004: val_auroc improved from 0.72202 to 0.75437, saving model to model_3.1

Epoch 5/10

53531/53531 [=====] - 495s 9ms/step - loss: 0.5375 - auroc: 0.7747 - val_loss: 0.5510 - val_auroc: 0.7550

Epoch 00005: val_auroc improved from 0.75437 to 0.75499, saving model to model_3.1

Epoch 6/10

53531/53531 [=====] - 497s 9ms/step - loss: 0.4876 - auroc: 0.8028 - val_loss: 0.5407 - val_auroc: 0.7480

Epoch 00006: val_auroc did not improve from 0.75499

Epoch 7/10

53531/53531 [=====] - 492s 9ms/step - loss: 0.4356 - auroc: 0.8382 - val_loss: 0.5047 - val_auroc: 0.7243

Epoch 00007: val_auroc did not improve from 0.75499

Epoch 8/10

53531/53531 [=====] - 492s 9ms/step - loss: 0.3830 - auroc: 0.8763 - val_loss: 0.5082 - val_auroc: 0.7154

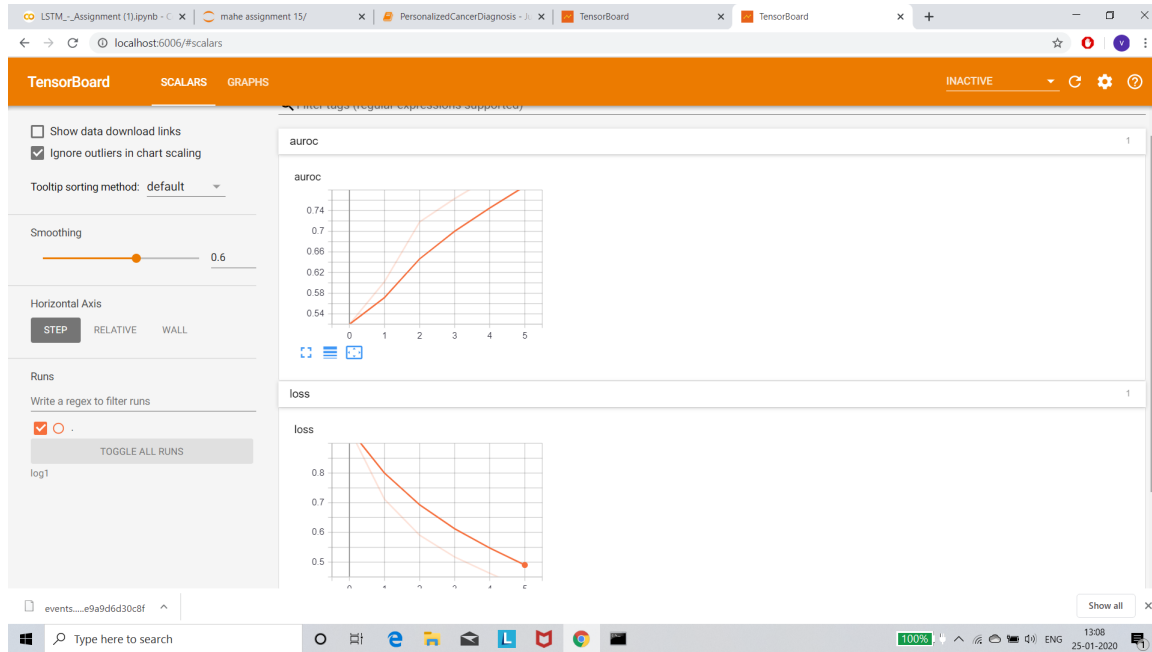
Epoch 00008: val_auroc did not improve from 0.75499

Epoch 00008: early stopping

In [0]:

```
Image(retina=True, filename='/content/Screenshot (28).png')
```

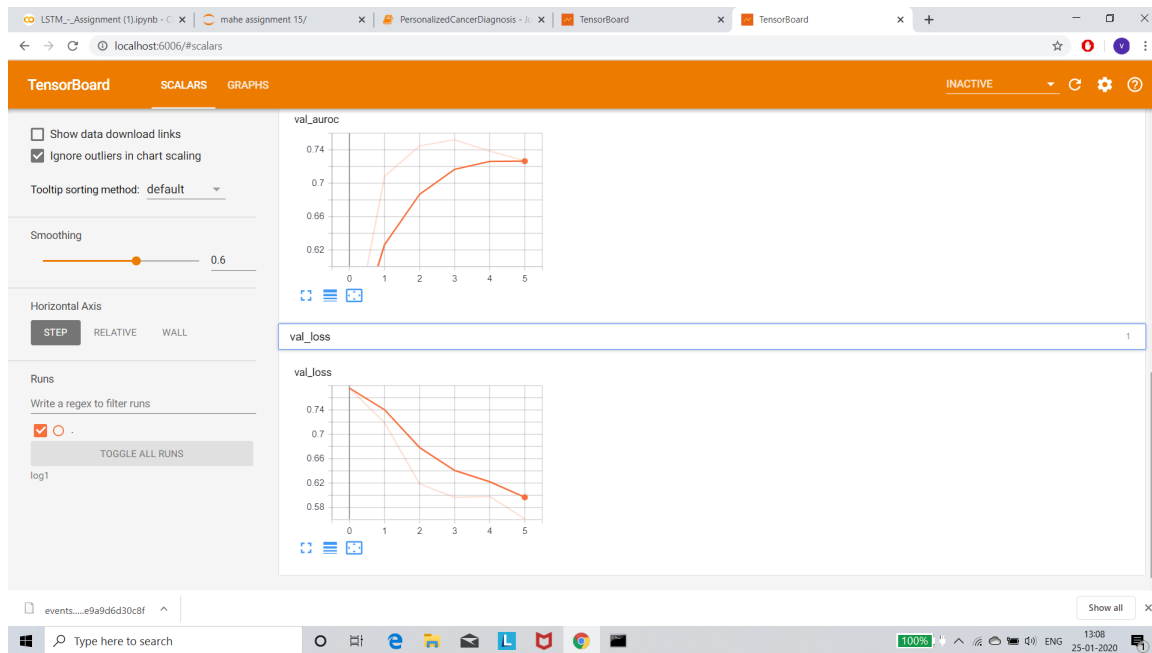
Out[0]:



In [0]:

```
Image(retina=True, filename='/content/Screenshot (29).png')
```

Out[0]:



CONCLUSION

In [43]:

```
from prettytable import PrettyTable

conclusion=PrettyTable()
conclusion.field_names = ["ARCHITECTURE", "EPOCHS", "TRAIN_auc", "VAL_auc",]
conclusion.add_row(["LSTM_1", 4, 0.7632,0.7520 ])
conclusion.add_row(["LSTM_2",5, 0.7545,0.7470])
conclusion.add_row(["LSTM_3", 5,0.7747,0.7550])

print(conclusion.get_string(start=0,end=7))
```

ARCHITECTURE	EPOCHS	TRAIN_auc	VAL_auc
LSTM_1	4	0.7632	0.752
LSTM_2	5	0.7545	0.747
LSTM_3	5	0.7747	0.755

In [0]: