

ref: 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.

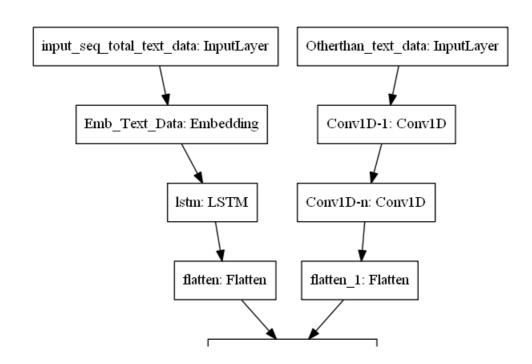
 Use given predefined glove word vectors, don't train any word vectors. After this use LSTM and get the LSTM output and Flatten that output.
- Input_school_state --- Give 'school_state' column as input to embedding layer and Train the Keras Embedding layer.
- **Project_grade_category** --- Give 'project_grade_category' 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_numerical_digits._price ---concatenate remaining columns and add a Dense layer after that.

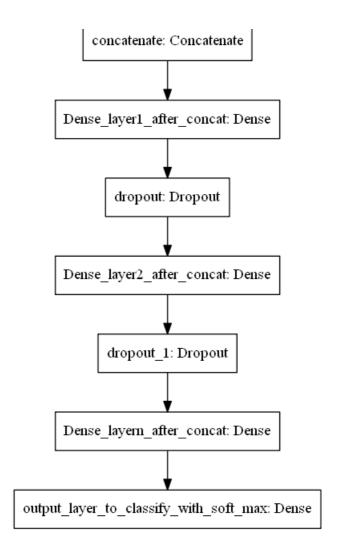
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Model-2

1.LSTM after removing the Low and High idf value words. (In model-1 Trained on total data b ut in Model-2 trained on data after removing some words based on IDF values)

Model-3





In [1]:

lst_ran.append(final_pos.iloc[-i])

```
##importing all required modules
import numpy as np
import pandas as pd
import sklearn
import seaborn as sns
import matplotlib.pyplot as plt
from time import time
import tensorflow as tf
from tensorflow.keras.layers import Dense,Conv1D,
MaxPooling1D, concatenate, BatchNormalization, Activation, Dropout, Input, Flatten, Embedding, LSTM
from tensorflow.keras.models import Model
from tensorflow.keras.callbacks import TensorBoard
from tensorflow.keras.preprocessing.text import one hot
{\bf from\ tensorflow.keras.preprocessing.sequence\ \underline{import}\ {\tt pad\_sequences}
In [2]:
final = pd.read_csv("lstm_preprocessed_data.csv")
final.shape
Out[2]:
(109248, 9)
In [3]:
"""final_neg = final[final["project_is_approved"] == 0]
final_pos = final[final["project_is_approved"] == 1]
lst_ran = []
for i in range (10000):
    lst ran.append(final neg.iloc[i])
```

```
project data = pd.DataFrame(lst ran)
project_data.shape"""
project_data = final
In [4]:
project data.head(2)
Out[4]:
   school_state teacher_prefix project_grade_category teacher_number_of_previously_posted_projects project_is_approved clean_cate
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In [5]:
from sklearn.model_selection import train_test_split
project_data_train, project_data_test, y_train, y_test = train_test_split(project_data, project_dat
a["project_is_approved"], test_size=0.2, stratify=project_data["project_is_approved"],
random state = 0)
print("Number of data points in train data", project_data_train.shape)
print('-'*50)
print("Number of data points in test data", project_data_test.shape)
print('-'*50)
Number of data points in train data (87398, 9)
Number of data points in test data (21850, 9)
```

One Hot Encoding

Categorical Features

```
In [6]:
```

```
val = dict(project data train["clean categories"].value counts())
val d = dict(enumerate(val.keys()))
category_val = {v: k for k, v in val_d.items()}
cate_train = project_data_train["clean_categories"].values
category_train = []
for i in cate_train:
    category_train.append(category_val[i])
cate_test = project_data_test["clean_categories"].values
category_test = []
for i in cate test:
   if i in category_val:
       category_test.append(category_val[i])
    else:
       category_test.append(len(val_d.keys()))
category_train = np.array(category_train)
category_test = np.array(category_test)
```

ın [/]:

```
val = dict(project data train["clean subcategories"].value counts())
val d = dict(enumerate(val.keys()))
subcategory val = {v: k for k, v in val d.items()}
subcate train = project data train["clean subcategories"].values
subcategory train = []
for i in subcate train:
    subcategory train.append(subcategory val[i])
subcate test = project data test["clean subcategories"].values
subcategory test = []
for i in subcate test:
   if i in subcategory_val:
       subcategory test.append(subcategory val[i])
    else:
       subcategory test.append(len(val d.keys()))
subcategory train = np.array(subcategory train)
subcategory_test = np.array(subcategory test)
```

In [8]:

```
val = dict(project data train["school state"].value_counts())
val d = dict(enumerate(val.keys()))
state val = {v: k for k, v in val d.items()}
state tr = project data train["school state"].values
state_enc_train = []
for i in state tr:
    state enc train.append(state val[i])
state te = project data test["school state"].values
state_enc_test = []
for i in state_te:
   if i in state val:
       state_enc_test.append(state_val[i])
    else:
        state enc test.append(len(val d.keys()))
state_enc_train = np.array(state_enc_train)
state_enc_test = np.array(state_enc_test)
```

In [9]:

```
val = dict(project data train["project grade category"].value counts())
val d = dict(enumerate(val.keys()))
grade val = {v: k for k, v in val d.items()}
grade train = project data train["project grade category"].values
grade_enc_train = []
for i in grade train:
    grade enc train.append(grade val[i])
grade test = project data test["project grade category"].values
grade_enc_test = []
for i in grade_test:
   if i in grade val:
       grade_enc_test.append(grade_val[i])
    else:
       grade_enc_test.append(len(val_d.keys()))
grade enc train = np.array(grade enc train)
grade_enc_test = np.array(grade_enc_test)
```

In [10]:

```
val = dict(project_data_train["teacher_prefix"].value_counts())
val_d = dict(enumerate(val.keys()))
prefix_val = {v: k for k, v in val_d.items()}

prefix_train = project_data_train["teacher_prefix"].values
prefix_enc_train = []
for i in prefix_train:
```

```
prefix_enc_train.append(prefix_val[i])

prefix_test = project_data_test["teacher_prefix"].values

prefix_enc_test = []

for i in prefix_test:
    if i in prefix_val:
        prefix_enc_test.append(prefix_val[i])
    else:
        prefix_enc_test.append(len(val_d.keys()))

prefix_enc_train = np.array(prefix_enc_train)

prefix_enc_test = np.array(prefix_enc_test)
```

Text Data

```
In [11]:
```

```
from tensorflow.keras.preprocessing.text import Tokenizer
from tensorflow.keras.preprocessing.sequence import pad_sequences

essay_train = project_data_train["essay"].values
essay_test = project_data_test["essay"].values

t = Tokenizer()
t.fit_on_texts(essay_train)
vocab_size = len(t.word_index) + 1
```

In [12]:

```
# integer encode the documents
#Train
enc_docs_train = t.texts_to_sequences(essay_train)
print(enc_docs_train[1])
```

[175, 20, 23, 2131, 53, 80, 14, 491, 312, 10, 1600, 457, 3729, 116, 10, 25, 1510, 292, 1, 816, 3, 30, 182, 840, 344, 4, 1, 6163, 1263, 547, 142, 1897, 4, 1, 1853, 1176, 880, 10, 14, 2480, 391, 160, 284, 4, 1, 40, 67, 26, 6, 1480, 1603, 463, 1824, 78, 6, 1441, 14, 1441, 44, 88, 102, 9, 3465, 14, 46, 32, 8818, 346, 8986, 38, 32, 101, 46, 827, 32, 6, 167, 98, 33016, 54, 2, 34, 153, 2720, 32 54, 3692, 231, 487, 376, 32, 2, 34, 1, 407, 538, 5053, 2705, 57, 118, 14, 15, 2852, 33, 1519, 14, 115, 5, 33, 2852, 193, 4, 1, 28, 15, 2346, 2519, 664, 1894, 3649, 530, 142, 13]

In [13]:

```
#Test
enc_docs_test = t.texts_to_sequences(essay_test)
print(enc_docs_test[1])
```

[14, 492, 20, 70, 26, 120, 864, 146, 4, 1, 19, 6164, 968, 624, 65, 1052, 118, 4578, 6, 7897, 2433, 161, 194, 20, 14, 19, 792, 73, 854, 109, 686, 14, 19, 17, 337, 23, 767, 1236, 74, 102, 241, 32, 69, 4, 23, 26, 373, 6, 4280, 14, 347, 50, 132, 138, 1, 169, 67, 273, 414, 680, 206, 3, 115, 216, 352, 623, 39, 1, 295, 48, 429, 6173, 110, 65, 41, 26, 638, 1047, 944, 1, 67, 134, 1526, 422, 125, 26, 38, 118, 791, 101, 160, 4318, 68, 309, 242, 207, 472, 1519, 4467, 4368, 1212, 3172, 19404, 50, 330, 32, 1941, 288, 1, 25, 23, 46, 32, 131, 2491, 303, 1941, 288, 68, 309, 203, 255, 1222, 1184, 32, 28, 49, 1313, 186, 2254, 3771, 202, 2712, 45, 15, 186, 2375, 159, 27, 100, 15, 10, 17, 102, 2, 89, 40, 3618, 1950, 23, 46, 32, 131, 13]

In [14]:

```
# truncate and/or pad input sequences
max_review_length = 400
pad_docs_train = pad_sequences(enc_docs_train, maxlen=max_review_length)
pad_docs_test = pad_sequences(enc_docs_test, maxlen=max_review_length)
print(pad_docs_train.shape)
print(pad_docs_train[10])

(87398, 400)
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62 2833 248 1257 123 587 4419 409 2
                                  94 1740 49 261 157
76 1892 249 112
               6 525 3264 319
                              90 801 39 503 249 112
282 11 1 40 249 112 282
                           131
```

In [15]:

```
"""# Reading glove vectors in python: https://stackoverflow.com/a/38230349/4084039
def loadGloveModel(gloveFile):
    print ("Loading Glove Model")
    f = open(gloveFile,'r', encoding="utf8")
    model = {}
    for line in f:
        splitLine = line.split()
        word = splitLine[0]
        embedding = np.array([float(val) for val in splitLine[1:]])
        model[word] = embedding
    print ("Done.",len(model)," words loaded!")
    return model
glovemodel = loadGloveModel('glove.42B.300d.txt')"""
```

Loading Glove Model
Done. 1917495 words loaded!

In [16]:

```
# create a weight matrix for words in training docs
embedding_matrix = np.zeros((vocab_size, 300))
for word, i in t.word_index.items():
    embedding_vector = glovemodel.get(word)
    if embedding_vector is not None:
        embedding_matrix[i] = embedding_vector
```

Numerical Features

In [17]:

```
from sklearn.preprocessing import StandardScaler

price_scalar = StandardScaler()
price_scalar.fit(project_data_train['price'].values.reshape(-1,1)) # finding the mean and standard deviation of this data
print(f"Mean : {price_scalar.mean_[0]}, Standard deviation : {np.sqrt(price_scalar.var_[0])}")

# Now standardize the data with above maen and variance.
price_standardized = price_scalar.transform(project_data_train['price'].values.reshape(-1, 1))
```

```
Mean: 298.3417191468912, Standard deviation: 365.8936474660223
In [18]:
from sklearn.preprocessing import StandardScaler
price scalar = StandardScaler()
price_scalar.fit(project_data_train['price'].values.reshape(-1,1)) # finding the mean and standard
deviation of this data
print(f"Mean : {price_scalar.mean_[0]}, Standard deviation : {np.sqrt(price scalar.var [0])}")
# Now standardize the data with above maen and variance.
price standardized test = price scalar.transform(project data test['price'].values.reshape(-1, 1))
Mean : 298.3417191468912, Standard deviation : 365.8936474660223
In [19]:
from sklearn.preprocessing import StandardScaler
import warnings
previouspro scalar = StandardScaler()
previouspro scalar.fit(project data train['teacher number of previously posted projects'].values.r
eshape(-1,1)) # finding the mean and standard deviation of this data
print(f"Mean : {previouspro_scalar.mean_[0]}, Standard deviation :
{np.sqrt(previouspro scalar.var [0])}")
previouspro standardized train =
previouspro_scalar.transform(project_data_train['teacher_number_of_previously_posted_projects'].va
lues.reshape(-1, 1))
warnings.filterwarnings("ignore")
Mean: 11.131524748850088, Standard deviation: 27.733763561081116
In [20]:
from sklearn.preprocessing import StandardScaler
previouspro scalar = StandardScaler()
previouspro_scalar.fit(project_data_train['teacher_number_of_previously_posted_projects'].values.r
eshape(-1,1)) # finding the mean and standard deviation of this data
print(f"Mean : {previouspro scalar.mean [0]}, Standard deviation :
{np.sqrt(previouspro scalar.var [0])}")
previouspro standardized test =
previouspro scalar.transform(project data test['teacher number of previously posted projects'].val
ues.reshape(-1, 1)
warnings.filterwarnings("ignore")
Mean: 11.131524748850088, Standard deviation: 27.733763561081116
In [21]:
remain train = np.stack((price standardized, previouspro standardized train), axis=-1)
remain train = remain train.reshape(remain train.shape[0],
remain train.shape[1]*remain train.shape[2])
remain test = np.stack((price standardized test, previouspro standardized test), axis=-1)
remain test = remain test.reshape(remain test.shape[0], remain test.shape[1]*remain test.shape[2])
```

Model 1

```
In [27]:
```

```
tf.keras.backend.clear_session()
Input_model_1 = Input(shape=(max_review_length,), name="Text_data")
```

```
layer11 = Embedding(vocab size, 300 ,weights=[embedding matrix],trainable=False ,name="layer11")
(Input model 1)
layer21 = LSTM(units = 100, activation='relu', kernel initializer='he normal', return sequences = Fals
e, name="layer21") (layer11)
layer31 = Flatten()(layer21)
Input_model_2 = Input(shape=(1,),name="School_state")
layer12 = Embedding(len(state_val) + 1, 32 ,name="layer12")(Input_model_2)
layer22 = Flatten()(layer12)
Input model 3 = Input(shape=(1,),name="Grade")
layer13 = Embedding(len(grade val) + 1, 32 ,name="layer13")(Input model 3)
layer23 = Flatten()(layer13)
Input model 4 = Input(shape=(1,),name="Categories")
layer14 = Embedding(len(category_val) + 1, 32 ,name="layer14")(Input model 4)
layer24 = Flatten()(layer14)
Input model 5 = Input(shape=(1,),name="sub Category")
layer15 = Embedding(len(subcategory val) + 1, 32 ,name="layer15")(Input model 5)
layer25 = Flatten()(layer15)
Input model 6 = Input(shape=(1,),name="Prefix")
layer16 = Embedding(len(prefix val) + 1, 32 ,name="layer16")(Input model 6)
layer26 = Flatten()(layer16)
Input_model_7 = Input(shape=(2,),name="Remaining_features")
layer27 = Dense(units=8,activation='relu',kernel initializer="he normal",name="layer27")(Input mode
1_7)
concat layer = concatenate(inputs=[layer31,layer22,layer23,layer24,layer25,layer26,layer27],name="
concat")
layer2 = Dense(units=512,activation='relu',kernel initializer='he normal',name="layer2")(concat lay
er)
norm 1 = BatchNormalization()(layer2)
layer3 = Dropout(0.25) (norm 1)
layer4 = Dense(units=256,activation='relu',kernel_initializer='he_normal',name="layer4") (layer3)
norm 2 = BatchNormalization()(layer4)
layer5 = Dense(units=128,activation='relu',kernel initializer='he normal',name="layer5") (norm 2)
norm 3 = BatchNormalization()(layer5)
layer6 = Dense(units=64,activation='relu',kernel initializer='he normal',name="layer6")(norm 3)
output = Dense(units=1,activation='sigmoid',kernel initializer="glorot uniform",name="output")(laye
model = Model(inputs=[Input model 1,Input model 2,Input model 3,Input model 4,Input model 5,Input m
odel 6, Input model 7], outputs=output)
4
                                                                                                 )
```

In [28]:

model.summary()

Model: "model"

| Layer (type) | Output Shape | Param # | Connected to | | |
|---------------------------|------------------------|----------|---|--|--|
| Text_data (InputLayer) | [(None, 400)] | 0 | :====================================== | | |
| layer11 (Embedding) | (None, 400, 300) | 15540900 | Text_data[0][0] | | |
| School_state (InputLayer) | [(None, 1)] | 0 | | | |
| Grade (InputLayer) | [(None, 1)] | | | | |
| Categories (InputLayer) | [(None, 1)] | 0 | | | |
| sub_Category (InputLayer) | [(None, 1)] | 0 | | | |
| Prefix (InputLayer) | [(None, 1)] | 0 | | | |
| layer21 (LSTM) | (None, 100) | 160400 | layer11[0][0] | | |
| layer12 (Embedding) | (None, 1, 32) | 1664 | School_state[0][0] | | |
| laver13 (Fmheddina) | Emheddinal (None 1 32) | | Grade[N][N] | | |

| rayerro (bumedaring) | / TAOTTE 1 | 1, 561 | T 0 0 | Grade[0][0] |
|---------------------------------|------------|--------|--------|---|
| layer14 (Embedding) | (None, | 1, 32) | 1664 | Categories[0][0] |
| layer15 (Embedding) | (None, | 1, 32) | 12608 | sub_Category[0][0] |
| layer16 (Embedding) | (None, | 1, 32) | 192 | Prefix[0][0] |
| Remaining_features (InputLayer) | [(None | , 2)] | 0 | |
| flatten (Flatten) | (None, | 100) | 0 | layer21[0][0] |
| flatten_1 (Flatten) | (None, | 32) | 0 | layer12[0][0] |
| flatten_2 (Flatten) | (None, | 32) | 0 | layer13[0][0] |
| flatten_3 (Flatten) | (None, | 32) | 0 | layer14[0][0] |
| flatten_4 (Flatten) | (None, | 32) | 0 | layer15[0][0] |
| flatten_5 (Flatten) | (None, | 32) | 0 | layer16[0][0] |
| layer27 (Dense) | (None, | 8) | 24 | Remaining_features[0][0] |
| concat (Concatenate) | (None, | 268) | 0 | flatten[0][0] flatten_1[0][0] flatten_2[0][0] flatten_3[0][0] flatten_4[0][0] flatten_5[0][0] layer27[0][0] |
| layer2 (Dense) | (None, | 512) | 137728 | concat[0][0] |
| batch_normalization (BatchNorma | (None, | 512) | 2048 | layer2[0][0] |
| dropout (Dropout) | (None, | 512) | 0 | batch_normalization[0][0] |
| layer4 (Dense) | (None, | 256) | 131328 | dropout[0][0] |
| batch_normalization_1 (BatchNor | (None, | 256) | 1024 | layer4[0][0] |
| layer5 (Dense) | (None, | 128) | 32896 | batch_normalization_1[0][0] |
| batch_normalization_2 (BatchNor | (None, | 128) | 512 | layer5[0][0] |
| layer6 (Dense) | (None, | 64) | 8256 | batch_normalization_2[0][0] |
| output (Dense) | (None, | 1) | 65 | layer6[0][0] |
| | | | | |

Total params: 16,031,469
Trainable params: 488,777

Non-trainable params: 15,542,692

In [29]:

```
from sklearn.metrics import roc_auc_score
def auc(y_true, y_pred):
    return tf.py_func(roc_auc_score, (y_true, y_pred), tf.double)
```

In [30]:

```
#compile
model.compile(optimizer='adam',loss='binary_crossentropy',metrics = ["accuracy", auc])
```

In [31]:

```
"""class_weights = {0 : 1,
1 :0.2}"""
```

Out[31]:

```
'class weights = \{0 : 1, \n
```

```
In [32]:
```

```
from time import time
logs = r"C:\Users\Dewang\Desktop\AAIC\ notes\0.0\ Assignments\AAIC\ classroom\ Assignments\logs\ex1"
tensorboard = TensorBoard(log dir = logs )
In [34]:
#train
model.fit([pad docs train, state enc train, grade enc train, category train, subcategory train, prefix e
nc_train,remain_train],y_train,
        batch size=256, epochs=10,
        validation data= ([pad docs test, state enc test, grade enc test, category test, subcategory
_test,prefix_enc test,remain test],y test),
       callbacks = [tensorboard])
4
                                                                           I
Train on 87398 samples, validate on 21850 samples
Epoch 1/10
87398/87398 [============== ] - 577s 7ms/sample - loss: 0.4248 - acc: 0.8480 - auc:
0.5906 - val loss: 0.4334 - val acc: 0.8486 - val auc: 0.4089
Epoch 2/10
87398/87398 [============== ] - 504s 6ms/sample - loss: 0.4169 - acc: 0.8485 - auc:
0.6133 - val loss: 0.4318 - val acc: 0.8486 - val auc: 0.4115
Epoch 3/10
87398/87398 [============== ] - 504s 6ms/sample - loss: 0.4134 - acc: 0.8486 - auc:
0.6260 - val_loss: 0.4389 - val_acc: 0.8486 - val auc: 0.4212
Epoch 4/10
0.6354 - val loss: 0.4345 - val acc: 0.8486 - val auc: 0.4400
Epoch 5/10
0.6408 - val loss: 0.4292 - val acc: 0.8486 - val auc: 0.5081
Epoch 6/10
87398/87398 [============== ] - 503s 6ms/sample - loss: 0.4076 - acc: 0.8488 - auc:
0.6480 - val_loss: 0.4304 - val_acc: 0.8486 - val auc: 0.5037
Epoch 7/10
0.6546 - val loss: 0.4263 - val acc: 0.8486 - val auc: 0.5504
Epoch 8/10
0.6623 - val loss: 0.4239 - val acc: 0.8486 - val auc: 0.5406
Epoch 9/10
87398/87398 [============== ] - 506s 6ms/sample - loss: 0.4010 - acc: 0.8496 - auc:
0.6694 - val_loss: 0.4246 - val_acc: 0.8486 - val_auc: 0.5396
Epoch 10/10
0.6783 - val_loss: 0.4350 - val_acc: 0.8486 - val_auc: 0.4985
Out[34]:
<tensorflow.python.keras.callbacks.History at 0x1c0cf7d4ef0>
In [57]:
model json = model.to json()
with open ("model1.json", "w") as json file:
   json file.write(model json)
model.save_weights("model1.h5")
print("Saved model to disk")
Saved model to disk
```

In [58]:

```
from tensorflow.keras.models import model_from_json
json_file = open('model1.json', 'r')
loaded_model_json = json_file.read()
json_file.close()

loaded_model = model_from_ison(loaded_model_ison)
```

```
# load weights into new model
loaded model.load weights("model1.h5")
print("Loaded model from disk")
W1011 12:07:16.377816 8056 deprecation.py:506] From C:\Users\Dewang\Anaconda3\lib\site-
packages\tensorflow\python\ops\init ops.py:97: calling Orthogonal. init (from
tensorflow.python.ops.init_ops) with dtype is deprecated and will be removed in a future version.
Instructions for updating:
Call initializer instance with the dtype argument instead of passing it to the constructor
\verb|W1011 12:07:16.409270 8056 deprecation.py:506| From C:\Users\Dewang\Anaconda3\lib\site-property | From C
packages\tensorflow\python\ops\init ops.py:97: calling Zeros.
                                                                                                                                                                          _init__ (from
tensorflow.python.ops.init_ops) with dtype is deprecated and will be removed in a future version.
Instructions for updating:
Call initializer instance with the dtype argument instead of passing it to the constructor
W1011 12:07:16.429961 8056 deprecation.py:506] From C:\Users\Dewang\Anaconda3\lib\site-
packages\tensorflow\python\ops\init_ops.py:97: calling Ones.__init__ (from
tensorflow.python.ops.init_ops) with dtype is deprecated and will be removed in a future version.
Instructions for updating:
Call initializer instance with the dtype argument instead of passing it to the constructor
W1011 12:07:16.437951 8056 deprecation.py:506] From C:\Users\Dewang\Anaconda3\lib\site-
packages\tensorflow\python\ops\init_ops.py:97: calling GlorotUniform.__init__ (from
tensorflow.python.ops.init_ops) with dtype is deprecated and will be removed in a future version.
Instructions for updating:
Call initializer instance with the dtype argument instead of passing it to the constructor
Loaded model from disk
In [59]:
loaded_model.compile(optimizer='adam',loss='binary_crossentropy',metrics = ["accuracy", auc])
In [63]:
score =
loaded\_model.evaluate([pad\_docs\_test, state\_enc\_test, grade\_enc\_test, category\_test, subcategory\_test, grade\_enc\_test, grade
prefix enc test,remain test],y test,batch size = 64, verbose=1)
print('Test score:', score[0])
print('Test accuracy:', score[1])
0.7261
Test score: 0.5534265434059998
Test accuracy: 0.7437071
In [ ]:
 %load ext tensorboard
%tensorboard --logdir {logs_base_dir}
In [66]:
%tensorboard --logdir=logs --host localhost --port 810
Reusing TensorBoard on port 810 (pid 10556), started 0:00:45 ago. (Use '!kill 10556' to kill it.)
```

Model 2

In [24]:

```
from sklearn.feature_extraction.text import TfidfVectorizer
essay_train = project_data_train["essay"]
vectorizer = TfidfVectorizer()
X = vectorizer.fit_transform(essay_train)
idf = vectorizer.idf_
idf_dict = dict(zip(vectorizer.get_feature_names(), idf))
idf_dict["i"] = 1
```

In [25]:

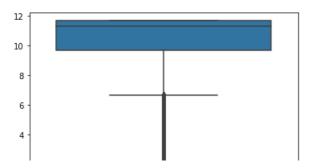
```
sorted_idf_dict = dict(sorted(idf_dict.items(), key=lambda kv: kv[1]))
sorted_idf = list(sorted_idf_dict.values())
```

In [26]:

```
sns.boxplot(y = sorted_idf)
```

Out[26]:

<matplotlib.axes._subplots.AxesSubplot at 0x20a69ed67b8>



```
In [27]:
```

```
"""Removing Values below 6 and above 10"""
remove_words = {}
for i,j in idf_dict.items():
    if(int(j)<6 or int(j) >10):
        remove_words[i] = idf_dict[i]
```

In [29]:

```
"""Removing Words from Data
from tqdm import tqdm

essay_train = project_data_train["essay"]
essay_test = project_data_test["essay"]

stopwords = list(remove_words.keys())

idf_essay_train = []

for query in tqdm(essay_train):

    querywords = query.split()
    resultwords = [word for word in querywords if word.lower() not in stopwords]
    result = ' '.join(resultwords)
    idf_essay_train.append(result)

"""
```

Out[29]:

'Removing Words from Data\nfrom tqdm import tqdm\n\nessay_train =
project_data_train["essay"]\nessay_test = project_data_test["essay"]\n\nstopwords =
list(remove_words.keys())\n\nidf_essay_train = []\n\nfor query in tqdm(essay_train):\n \n que
rywords = query.split()\n resultwords = [word for word in querywords if word.lower() not in
stopwords]\n result = \' \'.join(resultwords)\n idf_essay_train.append(result)\n \n'

In []:

```
"""idf_essay_test = []
for query in tqdm(essay_test):
    querywords = query.split()
    resultwords = [word for word in querywords if word.lower() not in stopwords]
    result = ' '.join(resultwords)
    idf_essay_test.append(result)"""
```

In [106]:

```
"""df2_train = pd.DataFrame(idf_essay_train,columns = ["train_essay"])
df2_test = pd.DataFrame(idf_essay_test,columns = ["test_essay"])
df2_train.to_csv(r'lstm_model2_train.csv')
df2_test.to_csv(r"lstm_model2_test.csv")"""

df2_train = pd.read_csv("lstm_model2_train.csv")
df2_test = pd.read_csv("lstm_model2_test.csv")

idf_essay_train= df2_train["train_essay"].astype(str)
idf_essay_test= df2_test["test_essay"].astype(str)
```

In [107]:

```
from tensorflow.keras.preprocessing.text import Tokenizer
from tensorflow.keras.preprocessing.sequence import pad_sequences

t = Tokenizer()
```

```
t.fit on texts(idf essay_train)
vocab size idf = len(t.word index) + 1
In [108]:
# integer encode the documents
idf docs_train = t.texts_to_sequences(idf_essay_train)
print(idf docs train[1])
[2360, 3424, 3381, 4526, 244]
In [109]:
idf_docs_test = t.texts_to_sequences(idf_essay_test)
print(idf docs test[1])
[370, 9405, 613, 613, 613, 370, 2857, 15363, 731, 987, 12924, 370, 836, 2937, 21536, 2453, 1561, 3
99, 278, 1773, 2163, 1147, 4067, 1346, 1147, 2761, 21536, 15181, 170]
In [110]:
# truncate and/or pad input sequences
max review length = 200
pad_idf_train = pad_sequences(idf_docs_train, maxlen=max_review_length)
pad idf test = pad sequences(idf docs test, maxlen=max review length)
print(pad idf train.shape)
print(pad_idf_train[10])
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 1525 542
          1
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In [111]:
# create a weight matrix for words in training docs
embedding matrix idf = np.zeros((vocab size idf, 300))
for word, i in t.word index.items():
   embedding vector = glovemodel.get(word)
   if embedding vector is not None:
       embedding matrix idf[i] = embedding vector
In [204]:
tf.keras.backend.clear session()
Input_model_1 = Input(shape=(max_review_length,) ,name="Text_data")
layer11 = Embedding (vocab size idf, 300, weights=[embedding matrix idf], trainable=False
, name="layer11") (Input_model_1)
layer21 = LSTM(units = 100,activation='relu',kernel initializer='he normal',return sequences = Fals
e, name="layer21") (layer11)
layer31 = Flatten()(layer21)
```

Input model 2 = Input(shape=(1,),name="School state")

layer12 = Embedding(state_size + 1, 2 ,name="layer12")(Input_model_2)

```
layer22 = Flatten()(layer12)
Input_model_3 = Input(shape=(3,),name="Grade")
layer13 = Embedding(grade size + 1, 2 ,name="layer13")(Input model 3)
layer23 = Flatten()(layer13)
Input model 4 = Input(shape=(max category length,),name="Categories")
layer14 = Embedding(cate_size + 1, 4 ,name="layer14")(Input_model_4)
layer24 = Flatten()(layer14)
Input_model_5 = Input(shape=(max_subcategory_length,),name="sub_Category")
layer15 = Embedding(subcate_size + 1, 4 ,name="layer15")(Input model 5)
layer25 = Flatten()(layer15)
Input model 6 = Input(shape=(1,),name="Prefix")
layer16 = Embedding(prefix size + 1, 2 ,name="layer16")(Input model 6)
layer26 = Flatten()(layer16)
Input model 7 = Input(shape=(2,),name="Remaining features")
layer27 = Dense (units=8,activation='relu', kernel initializer="he normal", name="layer27") (Input mode
concat layer = concatenate(inputs=[layer31,layer22,layer23,layer24,layer25,layer26,layer27],name="
concat")
layer2 = Dense(units=512,activation='relu',kernel initializer='he normal',name="layer2")(concat lay
er)
norm 1 = BatchNormalization()(layer2)
layer3 = Dropout(0.25) (norm 1)
layer4 = Dense(units=256,activation='relu',kernel initializer='he normal',name="layer4") (layer3)
norm 2 = BatchNormalization()(layer4)
layer5 = Dense(units=128,activation='relu',kernel initializer='he normal',name="layer5") (norm 2)
norm 3 = BatchNormalization()(layer5)
layer6 = Dense(units=64,activation='relu',kernel initializer='he normal',name="layer6")(norm 3)
output = Dense(units=1,activation='sigmoid',kernel_initializer="glorot_uniform",name="output")(laye
model = Model(inputs=[Input_model_1,Input_model_2,Input_model_3,Input_model_4,Input_model_5,Input_m
odel 6,Input model 7],outputs=output)
model.summary()
                                                                                                 •
4
```

Model: "model"

| model. | | | |
|---------------------------------|----------------------|---------|--------------------|
| Layer (type) | Output Shape | Param # | Connected to |
| Text_data (InputLayer) | [(None, 400)] | 0 | |
| layer11 (Embedding) | (None, 400, 300) | 7017000 | Text_data[0][0] |
| School_state (InputLayer) | [(None, 1)] | 0 | |
| Grade (InputLayer) | [(None, 3)] | 0 | |
| Categories (InputLayer) | [(None, 5)] | 0 | |
| sub_Category (InputLayer) | [(None, 6)] | 0 | |
| Prefix (InputLayer) | [(None, 1)] | 0 | |
| layer21 (LSTM) | (None, 100) | 160400 | layer11[0][0] |
| layer12 (Embedding) | (None, 1, 2) | 104 | School_state[0][0] |
| layer13 (Embedding) | (None, 3, 2) | 10 | Grade[0][0] |
| layer14 (Embedding) | (None, 5, 4) | 40 | Categories[0][0] |
| layer15 (Embedding) | (None, 6, 4) | 124 | sub_Category[0][0] |
| layer16 (Embedding) | (None, 1, 2) | 12 | Prefix[0][0] |
| Remaining_features (InputLayer) | [(None, 2)] | 0 | |
| flatten (Flatten) | (None, 100) | 0 | layer21[0][0] |
| flatten 1 /Flatten) | 1 /Flatton) /None 2) | | 120212[0][0] |

| rracceul (rracceu) | (11011 C , | ۷ ا | v | TalerTr[0][0] |
|---------------------------------|-------------------|------|--------|-----------------------------|
| flatten_2 (Flatten) | (None, | 6) | 0 | layer13[0][0] |
| flatten_3 (Flatten) | (None, | 20) | 0 | layer14[0][0] |
| flatten_4 (Flatten) | (None, | 24) | 0 | layer15[0][0] |
| flatten_5 (Flatten) | (None, | 2) | 0 | layer16[0][0] |
| Layer27 (Dense) | (None, | 8) | 24 | Remaining_features[0][0] |
| concat (Concatenate) | (None, | 162) | 0 | flatten[0][0] |
| | | | | flatten 1[0][0] |
| | | | | flatten 2[0][0] |
| | | | | flatten 3[0][0] |
| | | | | flatten 4[0][0] |
| | | | | flatten_5[0][0] |
| | | | | layer27[0][0] |
| layer2 (Dense) | (None, | 512) | 83456 | concat[0][0] |
| patch_normalization (BatchNorma | (None, | 512) | 2048 | layer2[0][0] |
| dropout (Dropout) | (None, | 512) | 0 | batch_normalization[0][0] |
| layer4 (Dense) | (None, | 256) | 131328 | dropout[0][0] |
| patch_normalization_1 (BatchNor | (None, | 256) | 1024 | layer4[0][0] |
| layer5 (Dense) | (None, | 128) | 32896 | batch_normalization_1[0][0] |
| oatch_normalization_2 (BatchNor | (None, | 128) | 512 | layer5[0][0] |
| layer6 (Dense) | (None, | 64) | 8256 | batch_normalization_2[0][0] |
| output (Dense) | (None, | 1) | 65 | layer6[0][0] |

Total params: 7,437,299
Trainable params: 418,507
Non-trainable params: 7,018,792

In [146]:

```
from sklearn.metrics import roc_auc_score
def auc(y_true, y_pred):
    return tf.py_func(roc_auc_score, (y_true, y_pred), tf.double)
```

In [147]:

```
#compile
model.compile(optimizer='adam',loss='binary_crossentropy',metrics = ["accuracy", auc])
```

In [148]:

```
from time import time
logs1 = r"C:\Users\Dewang\Desktop\AAIC notes\0.0 Assignments\AAIC classroom Assignments\logs\ex2"
tensorboard = TensorBoard(log_dir = logs1 )
```

In [149]:

```
class_weights = {0: 1,
1: 0.2}
```

In [150]:

```
cest,prefix_enc_test,remain_test],y_test),
      class weight=class weights,callbacks = [tensorboard])
4
Train on 87398 samples, validate on 21850 samples
Epoch 1/10
0.5190 - val loss: 0.6576 - val acc: 0.8486 - val auc: 0.5000
Epoch 2/10
0.5180 - val loss: 0.6976 - val acc: 0.1514 - val auc: 0.5002
Epoch 3/10
0.5192 - val loss: 0.6093 - val acc: 0.8486 - val auc: 0.4998
Epoch 4/10
0.5323 - val_loss: 0.6406 - val_acc: 0.8486 - val_auc: 0.5000
Epoch 5/10
0.5450 - val loss: 0.6357 - val acc: 0.8486 - val auc: 0.5000
Epoch 6/10
0.5499 - val loss: 0.6409 - val acc: 0.8486 - val auc: 0.5000
Epoch 7/10
0.5489 - val loss: 0.7018 - val acc: 0.1514 - val auc: 0.5000
Epoch 8/10
0.5512 - val_loss: 0.6591 - val_acc: 0.8486 - val_auc: 0.5003
Epoch 9/10
0.5562 - val loss: 0.6475 - val acc: 0.8486 - val auc: 0.5000
Epoch 10/10
0.5590 - val loss: 0.6431 - val acc: 0.8486 - val auc: 0.5000
Out[150]:
<tensorflow.python.keras.callbacks.History at 0x2434efff898>
In [56]:
model json = model.to_json()
with open ("model2.json", "w") as json file:
  json_file.write(model_json)
model.save weights("model2.h5")
print("Saved model to disk")
Saved model to disk
```

In [57]:

```
from tensorflow.keras.models import model from json
json file = open('model2.json', 'r')
loaded_model_json = json_file.read()
json file.close()
loaded_model = model_from_json(loaded_model_json)
# load weights into new model
loaded model.load weights("model2.h5")
print("Loaded model from disk")
```

Loaded model from disk

```
In [58]:
```

```
loaded model.compile(optimizer='adam',loss='binary crossentropy',metrics = ["accuracy", auc])
```

```
In [59]:
```

In [77]:

```
%tensorboard --logdir=logs1 --host localhost --port 8089

'kill' is not recognized as an internal or external command,
operable program or batch file.
```

Reusing TensorBoard on port 8089 (pid 11512), started 0:01:58 ago. (Use '!kill 11512' to kill it.)

Categorica Features

```
In [206]:
# we use count vectorizer to convert the values into one
from sklearn.feature extraction.text import CountVectorizer
vectorizer = CountVectorizer()
categories one hot = vectorizer.fit transform(project data train['clean categories'].values)
print(vectorizer.get feature names())
print("Shape of matrix after one hot encodig ", categories one hot.shape)
['appliedlearning', 'care hunger', 'health sports', 'history civics', 'literacy language',
'math science', 'music arts', 'specialneeds', 'warmth']
Shape of matrix after one hot encodig (87398, 9)
In [207]:
# we use count vectorizer to convert the values into one
from sklearn.feature extraction.text import CountVectorizer
vectorizer = CountVectorizer()
vectorizer.fit(project data train['clean categories'].values)
categories one hot test = vectorizer.transform(project data test['clean categories'].values)
print(vectorizer.get_feature_names())
print("Shape of matrix after one hot encodig ", categories one hot test.shape)
['appliedlearning', 'care_hunger', 'health_sports', 'history_civics', 'literacy_language',
'math_science', 'music_arts', 'specialneeds', 'warmth']
Shape of matrix after one hot encodig (21850, 9)
In [208]:
# we use count vectorizer to convert the values into one
vectorizer = CountVectorizer()
sub categories one hot = vectorizer.fit transform(project data train['clean subcategories'].values
print(vectorizer.get feature names())
print("Shape of matrix after one hot encodig ", sub categories one hot.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', 'm athematics', 'music', 'nutritioneducation', 'other', 'parentinvolvement', 'performingarts', 'socia
lsciences', 'specialneeds', 'teamsports', 'visualarts', 'warmth']
Shape of matrix after one hot encodig (87398, 30)
In [209]:
# we use count vectorizer to convert the values into one
vectorizer = CountVectorizer()
vectorizer.fit(project_data_train['clean_subcategories'].values)
sub categories one hot test = vectorizer.transform(project data test['clean subcategories'].values
print(vectorizer.get feature names())
print("Shape of matrix after one hot encodig ", sub categories one hot 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', 'm
athematics', 'music', 'nutritioneducation', 'other', 'parentinvolvement', 'performingarts', 'socia lsciences', 'specialneeds', 'teamsports', 'visualarts', 'warmth']
Shape of matrix after one hot encodig (21850, 30)
In [210]:
vectorizer = CountVectorizer()
```

school state one hot = vectorizer.fit transform(project data train['school state'].values)

```
print(vectorizer.get feature names())
print("Shape of matrix after one hot encodig ", school state one hot.shape)
['ak', 'al', 'ar', 'az', 'ca', 'co', 'ct', 'dc', 'de', 'fl', 'ga', 'hi', 'ia', 'id', 'il', 'in', 'k
s', 'ky', 'la', 'ma', 'md', 'me', 'mi', 'mn', 'mo', 'ms', 'mt', 'nc', 'nd', 'ne', 'nh', 'nj', 'nm',
'nv', 'ny', 'oh', 'ok', 'or', 'pa', 'ri', 'sc', 'sd', 'tn', 'tx', 'ut', 'va', 'vt', 'wa', 'wi', 'wv
', 'wy']
Shape of matrix after one hot encodig (87398, 51)
4
In [211]:
vectorizer = CountVectorizer()
vectorizer.fit(project_data_train['school_state'].values)
print(vectorizer.get feature names())
school state one hot test = vectorizer.transform(project data test['school state'].values)
print("Shape of matrix after one hot encodig ",school_state_one_hot_test.shape)
['ak', 'al', 'ar', 'az', 'ca', 'co', 'ct', 'dc', 'de', 'fl', 'ga', 'hi', 'ia', 'id', 'il', 'in', 'k
s', 'ky', 'la', 'ma', 'md', 'me', 'mi', 'mn', 'mo', 'ms', 'mt', 'nc', 'nd', 'ne', 'nh', 'nj', 'nm',
'nv', 'ny', 'oh', 'ok', 'or', 'pa', 'ri', 'sc', 'sd', 'tn', 'tx', 'ut', 'va', 'vt', 'wa', 'wi', 'wv
', 'wv']
Shape of matrix after one hot encodig (21850, 51)
In [212]:
vectorizer = CountVectorizer()
gra_cat_one_hot = vectorizer.fit_transform(project_data_train['project_grade_category'].values)
print(vectorizer.get feature names())
print("Shape of matrix after one hot encodig ",gra cat one hot.shape)
['grades_3_5', 'grades_6_8', 'grades_9_12', 'grades_prek_2']
Shape of matrix after one hot encodig (87398, 4)
In [213]:
vectorizer = CountVectorizer()
vectorizer.fit(project data train['project grade category'].values)
print(vectorizer.get_feature_names())
gra cat one hot test = vectorizer.transform(project data test['project grade category'].values)
print("Shape of matrix after one hot encodig ",gra cat one hot test.shape)
['grades_3_5', 'grades_6_8', 'grades_9_12', 'grades_prek_2']
Shape of matrix after one hot encodig (21850, 4)
In [214]:
vectorizer = CountVectorizer()
tea pre one hot = vectorizer.fit transform(project data train['teacher prefix'].values.astype('str'
) )
print(vectorizer.get feature names())
print ("Shape of matrix after one hot encodig ", tea pre one hot.shape)
['dr', 'mr', 'mrs', 'ms', 'teacher']
Shape of matrix after one hot encodig (87398, 5)
In [215]:
vectorizer = CountVectorizer()
vectorizer.fit(project data train['teacher prefix'].values.astype('str'))
tea pre one hot test =
vectorizer.transform(project data test['teacher prefix'].values.astype('str'))
print(vectorizer.get_feature_names())
print("Shape of matrix after one hot encodig ",tea_pre_one hot test.shape)
['dr', 'mr', 'mrs', 'ms', 'teacher']
Shape of matrix after one hot encodig (21850, 5)
```

Concating Numerical and Categorical Features

```
In [216]:
print(categories one hot.shape)
print(sub categories one hot.shape)
print(school state one hot.shape)
print(gra cat one hot.shape)
print(tea_pre_one_hot.shape)
print(remain_train.shape)
(87398, 9)
(87398, 30)
(87398, 51)
(87398, 4)
(87398, 5)
(87398, 2)
In [217]:
from scipy.sparse import hstack
concat train
hstack((categories_one_hot,sub_categories_one_hot,school_state_one_hot,gra_cat_one_hot,tea_pre_one_
hot, remain train))
concat train.shape
4
Out[217]:
(87398, 101)
In [218]:
from scipy.sparse import hstack
concat test =
hstack((categories_one_hot_test,sub_categories_one_hot_test,school_state_one_hot_test,gra_cat_one_h
ot test, tea pre one hot test, remain test))
concat test.shape
4
Out[218]:
(21850, 101)
In [219]:
concat array train = concat train.toarray()
concat_array_test = concat_test.toarray()
nrows, ncols = concat array train.shape
In [240]:
tf.keras.backend.clear session()
Input_model_1 = Input(shape=(max_review_length,) ,name="Text_data")
layer11 = Embedding (vocab size, 300 , weights=[embedding matrix], trainable=False , name="layer11")
(Input model 1)
layer21 = LSTM(units = 100, activation='relu', kernel initializer='he normal', return sequences = Fals
e, name="layer21") (layer11)
layer41 = Flatten()(layer21)
Input model 2 = Input(shape=(ncols,),name="Concat input")
layer12 = Embedding(ncols + 1, 16 ,name="layer12")(Input_model_2)
layer22 = Conv1D(64, kernel size = 10, activation='relu', name="layer22") (layer12)
pool 1 = MaxPooling1D(pool size=2)(layer22)
layer32 = Conv1D(32, kernel size = 10, activation='relu', name="layer32") (pool 1)
layer42 = Flatten()(layer32)
concat layer = concatenate(inputs=[layer41,layer42],name="concat")
layer2 = Dense(units=512,activation='relu',kernel_initializer='he_normal',name="layer2")(concat_lay
er)
norm 1 = BatchNormalization()(layer2)
```

```
layer3 = Dropout(0.25) (norm_1)
layer4 = Dense (units=256, activation='relu', kernel_initializer='he_normal', name="layer4") (layer3)
norm_2 = BatchNormalization() (layer4)
layer5 = Dense (units=128, activation='relu', kernel_initializer='he_normal', name="layer5") (norm_2)
norm_3 = BatchNormalization() (layer5)
layer6 = Dense (units=64, activation='relu', kernel_initializer='he_normal', name="layer6") (norm_3)
output = Dense (units=1, activation='sigmoid', kernel_initializer="glorot_uniform", name="output") (layer6)
model = Model (inputs=[Input_model_1, Input_model_2], outputs=output)
model.summary()
```

Model: "model"

| Layer (type) | Output | Shape | Param # | Connected to |
|---------------------------------|--------|----------------|----------|----------------------------------|
| Concat_input (InputLayer) | | , 101)] | 0 | |
| layer12 (Embedding) | (None, | 101, 16) | 1632 | Concat_input[0][0] |
| Text_data (InputLayer) | [(None | , 400)] | 0 | |
| layer22 (Conv1D) | (None, | 92, 64) | 10304 | layer12[0][0] |
| layer11 (Embedding) | (None, | 400, 300) | 15497100 | Text_data[0][0] |
| max_pooling1d (MaxPooling1D) | (None, | 46, 64) | 0 | layer22[0][0] |
| layer21 (LSTM) | (None, | 100) | 160400 | layer11[0][0] |
| layer32 (Conv1D) | (None, | 37, 32) | 20512 | max_pooling1d[0][0] |
| flatten (Flatten) | (None, | 100) | 0 | layer21[0][0] |
| flatten_1 (Flatten) | (None, | 1184) | 0 | layer32[0][0] |
| concat (Concatenate) | (None, | 1284) | 0 | flatten[0][0] flatten_1[0][0] |
| layer2 (Dense) | (None, | 512) | 657920 | concat[0][0] |
| batch_normalization (BatchNorma | (None, | 512) | 2048 | layer2[0][0] |
| dropout (Dropout) | (None, | 512) | 0 | batch_normalization[0][0] |
| layer4 (Dense) | (None, | 256) | 131328 | dropout[0][0] |
| batch_normalization_1 (BatchNor | (None, | 256) | 1024 | layer4[0][0] |
| layer5 (Dense) | (None, | 128) | 32896 | batch_normalization_1[0][0] |
| batch_normalization_2 (BatchNor | (None, | 128) | 512 | layer5[0][0] |
| layer6 (Dense) | (None, | 64) | 8256 | batch_normalization_2[0][0] |
| output (Dense) | (None, | 1) | 65 | layer6[0][0] |

Total params: 16,523,997 Trainable params: 1,025,105 Non-trainable params: 15,498,892

In [241]:

```
from sklearn.metrics import roc_auc_score
def auc(y_true, y_pred):
    return tf.py_func(roc_auc_score, (y_true, y_pred), tf.double)
```

In [242]:

```
#compile
model.compile(optimizer='adam',loss='binary_crossentropy',metrics = ["accuracy", auc])
```

```
In [243]:
from time import time
logs2 = r"C:\Users\Dewang\Desktop\AAIC notes\0.0 Assignments\AAIC classroom Assignments\logs\ex3"
tensorboard = TensorBoard(log dir = logs2 )
In [244]:
class weights = \{0: 1,
          1: 0.2}
In [245]:
#t.rain
model.fit([pad docs train, concat array train], y train,
       batch size=256,epochs=10,
       validation data= ([pad docs test,concat array test],y test),class weight = class weights
      ,callbacks = [tensorboard])
4
Train on 87398 samples, validate on 21850 samples
Epoch 1/10
0.5344 - val loss: 0.6402 - val acc: 0.8486 - val auc: 0.5170
Epoch 2/10
0.5768 - val loss: 0.6936 - val acc: 0.5209 - val auc: 0.5219
Epoch 3/10
0.6068 - val loss: 0.6650 - val acc: 0.8414 - val auc: 0.5246
Epoch 4/10
0.6239 - val loss: 0.6180 - val acc: 0.8486 - val auc: 0.5431
Epoch 5/10
0.6352 - val_loss: 0.6631 - val_acc: 0.8078 - val_auc: 0.5630
Epoch 6/10
0.6471 - val_loss: 0.6115 - val_acc: 0.8484 - val_auc: 0.5450
Epoch 7/10
87398/87398 [============== ] - 525s 6ms/sample - loss: 0.2094 - acc: 0.6563 - auc:
0.6580 - val loss: 0.5731 - val acc: 0.8486 - val auc: 0.5539
Epoch 8/10
87398/87398 [============= ] - 533s 6ms/sample - loss: 0.2080 - acc: 0.6567 - auc:
0.6660 - val loss: 0.6101 - val acc: 0.8333 - val auc: 0.5605
Epoch 9/10
0.6739 - val loss: 0.5590 - val acc: 0.8290 - val auc: 0.5790
Epoch 10/10
0.6846 - val loss: 0.5887 - val acc: 0.8335 - val auc: 0.5519
Out[245]:
<tensorflow.python.keras.callbacks.History at 0x2438487e080>
In [246]:
model json = model.to json()
with open ("model3.json", "w") as json file:
  json file.write(model json)
model.save weights("model3.h5")
print("Saved model to disk")
Saved model to disk
In [247]:
```

from tensorflow.keras.models import model_from_json

json_file = open('model3.json', 'r')
loaded model ison = ison file read()

```
toaded_modet_loom - loom_tite.tead()
 json_file.close()
 loaded_model = model_from_json(loaded_model_json)
 # load weights into new model
 loaded_model.load_weights("model3.h5")
print("Loaded model from disk")
Loaded model from disk
In [ ]:
 #compile
 loaded_model.compile(optimizer='adam',loss='binary_crossentropy',metrics = ["accuracy", auc])
In [ ]:
 score =
 loaded\_model.evaluate([pad\_idf\_test, state\_enc\_test, grade\_enc\_test, category\_test, subcategory\_test, particle and the content of the conte
 refix_enc_test,remain_test],y_test,
                                          callbacks = [tensorboard])
 print('Test score:', score[0])
 print('Test accuracy:', score[1])
```