

# LSTM - AFR dataset

## 1. Objective:

To find the polarity of review

In [0]:

```
import os
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
import pandas as pd
import warnings
warnings.filterwarnings("ignore")
```

## 2. Data Import:

In [0]:

```
# References
# https://www.programcreek.com/python/example/99451/sklearn.externals.joblib.dump

from sklearn.externals import joblib
```

In [0]:

```
# load data from pickle file

x_train = joblib.load("x_train.pkl")
x_test = joblib.load("x_test.pkl")
```

In [0]:

```
# Load label from pickle file

y_train = joblib.load("y_train.pkl")
y_test = joblib.load("y_test.pkl")
```

In [9]:

```
len(x_train[0])
```

Out[9]:

69

In [10]:

```
type(x_train[0])
```

Out[10]:

list

In [11]:

```
y_train.value_counts()
```

Out[11]:

```
1    245654
0     45682
Name: Score, dtype: int64
```

### 3. Padding:

In [12]:

```
# https://keras.io/preprocessing/sequence/
# https://keras.io/examples/imdb_lstm/

from keras.preprocessing import sequence
```

Using TensorFlow backend.

In [0]:

```
# padding of train and test data

x_train=sequence.pad_sequences(x_train,maxlen=100)
x_test =sequence.pad_sequences(x_test,maxlen=100)
```

In [14]:

```
x_train[0]
```

Out[14]:

```
array([[ 0,  0,  0,  0,  0,  0,  0,  0,  0,
        0,  0,  0,  0,  0,  0,  0,  0,  0,
        0,  0,  0,  0,  0,  0,  0,  0,  0,
        0,  0,  0,  0, 11, 3643, 187, 102, 341,
       17, 1762, 1903, 1081, 419,  1,  55, 419, 1222,
      225, 1770, 2082,  792, 2096, 2447, 746, 369, 352,
      858,  23,  209,  26,  5,  24, 209,  90, 580,
      127, 849,  114, 557, 1978, 16, 186,  1, 196,
     1646, 259,  975, 452, 1367, 41,  59, 1460, 638,
        1, 603, 4355,  96,  38, 624,  60, 11423, 237,
     1367, 11973, 265, 345, 273,  4, 207, 114, 19,
     562], dtype=int32)
```

In [15]:

x\_test[0]

Out[15]:

```
array([[ 0,  0,  0,  0,  0,  0,  0,  0,  0,  0,
        0,  0,  0,  0,  0,  0,  0,  0,  0,  0,
        0,  0,  0,  0,  0,  0,  0,  0,  0,  0,
        0,  0,  0,  0,  0,  0,  0,  0,  0,  0,
        0,  0,  0,  0,  0,  0,  0,  0,  0,  0,
        0,  0,  0,  0,  0,  0,  0,  0,  0,  0,
        0,  0,  0,  0,  0,  0,  0,  0,  0,  0,
        0,  0,  0, 29, 46, 341,  1,  3,  2, 151, 22
16, 178, 66, 833, 33, 84,  5, 151, 2216, 46,  4, 13
28, 1263,  3, 411, 284, 33, 178, 42, 29, 78, 204,
45, 86], dtype=int32)
```

In [16]:

```
print(x_train.shape)
print(x_test.shape)
```

```
(291336, 100)
(72835, 100)
```

## 4. LSTM Model

### 4.1 Single LSTM layer:

#### 4.1.1 Sequence model:

In [0]:

```
# References
# https://keras.io/layers/

from keras.models import Sequential
from keras.layers import Dense, Embedding, LSTM
from keras.initializers import glorot_normal
```

In [0]:

```

# References
# https://keras.io/layers/embeddings/#embedding
# https://keras.io/layers/recurrent/#lstm

# vocabulary size = 64868 +1 = 64869 = input_dim

# model creation

model=Sequential()

# embedding layer

model.add(Embedding(input_dim= 64869,output_dim=32,input_length=100,embeddings_init

# LSTM layer

model.add(LSTM(100))

# Dense layer

model.add(Dense(1,activation="sigmoid"))

```

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/tensorflow/python/framework/op\_def\_library.py:263: colocate\_with (from tensorflow.python.framework.ops) is deprecated and will be removed in a future version.

Instructions for updating:  
Colocations handled automatically by placer.

In [0]:

```

# Model Summary

model.summary()

```

Layer (type)	Output Shape	Param #
embedding_1 (Embedding)	(None, 100, 32)	2075808
lstm_1 (LSTM)	(None, 100)	53200
dense_1 (Dense)	(None, 1)	101

=====  
 Total params: 2,129,109  
 Trainable params: 2,129,109  
 Non-trainable params: 0

In [0]:

```

# Model compilation

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

```

#### 4.1.2 Model Training:

In [0]:

```
# model training
```

```
History=model.fit(x_train,y_train,batch_size=1000,epochs=10,verbose=1,validation_sp
```

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/tensorflow/python/ops/math\_ops.py:3066: to\_int32 (from tensorflow.python.ops.math\_ops) is deprecated and will be removed in a future version.

Instructions for updating:

Use tf.cast instead.

Train on 203935 samples, validate on 87401 samples

Epoch 1/10

203935/203935 [=====] - 415s 2ms/step - loss: 0.2981 - acc: 0.8864 - val\_loss: 0.1984 - val\_acc: 0.9216

Epoch 2/10

203935/203935 [=====] - 418s 2ms/step - loss: 0.1827 - acc: 0.9289 - val\_loss: 0.1947 - val\_acc: 0.9258

Epoch 3/10

203935/203935 [=====] - 419s 2ms/step - loss: 0.1681 - acc: 0.9348 - val\_loss: 0.1946 - val\_acc: 0.9245

Epoch 4/10

203935/203935 [=====] - 409s 2ms/step - loss: 0.1589 - acc: 0.9393 - val\_loss: 0.1971 - val\_acc: 0.9246

Epoch 5/10

203935/203935 [=====] - 416s 2ms/step - loss: 0.1500 - acc: 0.9430 - val\_loss: 0.1983 - val\_acc: 0.9229

Epoch 6/10

203935/203935 [=====] - 417s 2ms/step - loss: 0.1401 - acc: 0.9468 - val\_loss: 0.2006 - val\_acc: 0.9230

Epoch 7/10

203935/203935 [=====] - 414s 2ms/step - loss: 0.1326 - acc: 0.9498 - val\_loss: 0.2041 - val\_acc: 0.9227

Epoch 8/10

203935/203935 [=====] - 412s 2ms/step - loss: 0.1243 - acc: 0.9535 - val\_loss: 0.2063 - val\_acc: 0.9199

Epoch 9/10

203935/203935 [=====] - 410s 2ms/step - loss: 0.1147 - acc: 0.9575 - val\_loss: 0.2069 - val\_acc: 0.9231

Epoch 10/10

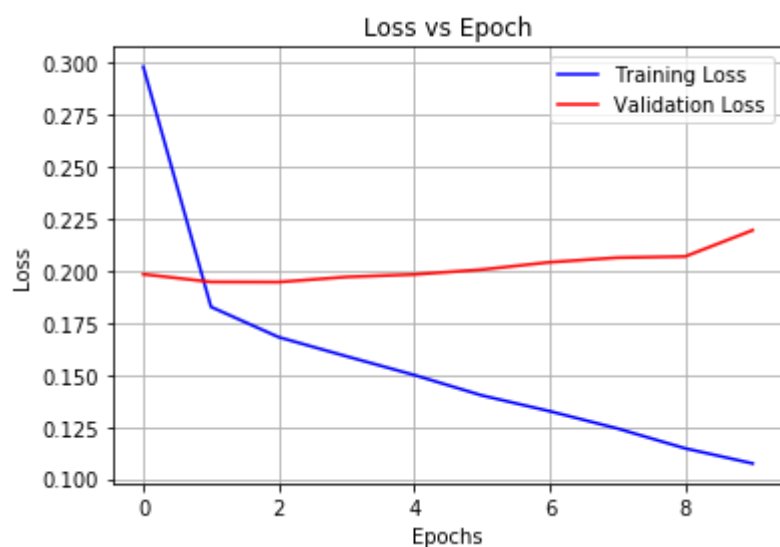
203935/203935 [=====] - 409s 2ms/step - loss: 0.1074 - acc: 0.9605 - val\_loss: 0.2196 - val\_acc: 0.9183

In [0]:

```
# References
# https://machinelearningmastery.com/display-deep-learning-model-training-history-i
# https://keras.io/models/sequential/

# Plotting Loss vs Epoch

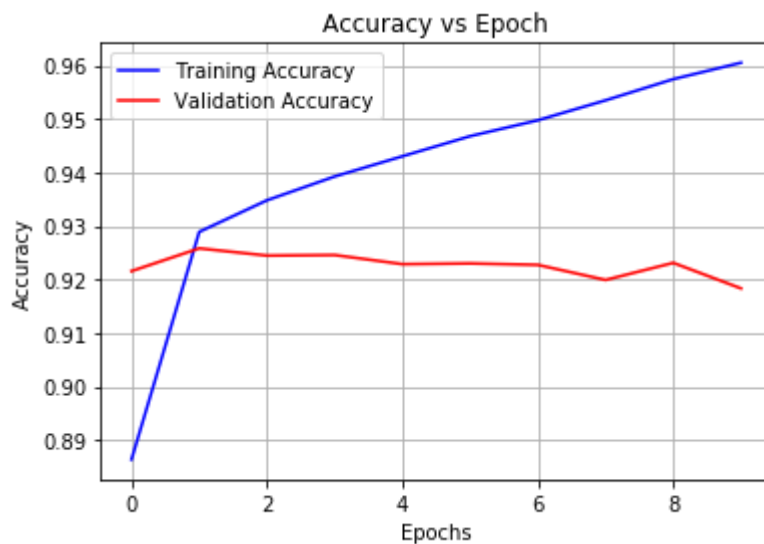
plt.close()
plt.plot(History.history['loss'],'b',label="Training Loss")
plt.plot(History.history['val_loss'],'r',label="Validation Loss")
plt.title("Loss vs Epoch")
plt.xlabel("Epochs")
plt.ylabel("Loss")
plt.legend()
plt.grid()
plt.show()
```



In [0]:

```
# Plotting Accuracy vs Epoch
```

```
plt.close()
plt.plot(History.history['acc'],'b',label="Training Accuracy")
plt.plot(History.history['val_acc'],'r',label="Validation Accuracy")
plt.title("Accuracy vs Epoch")
plt.xlabel("Epochs")
plt.ylabel("Accuracy")
plt.legend()
plt.grid()
plt.show()
```



In [0]:

```
# References
```

```
# https://stackoverflow.com/questions/43715047/keras-2-x-get-weights-of-layer
```

```
for layer in model.layers:
    print(layer.get_config())
```

```
{'name': 'embedding_1', 'trainable': True, 'batch_input_shape': (None, 100), 'dtype': 'float32', 'input_dim': 64869, 'output_dim': 32, 'embeddings_initializer': {'class_name': 'VarianceScaling', 'config': {'scale': 1.0, 'mode': 'fan_avg', 'distribution': 'normal', 'seed': None}}, 'embeddings_regularizer': None, 'activity_regularizer': None, 'embeddings_constraint': None, 'mask_zero': False, 'input_length': 100}
{'name': 'lstm_1', 'trainable': True, 'return_sequences': False, 'return_state': False, 'go_backwards': False, 'stateful': False, 'unroll': False, 'units': 100, 'activation': 'tanh', 'recurrent_activation': 'hard_sigmoid', 'use_bias': True, 'kernel_initializer': {'class_name': 'VarianceScaling', 'config': {'scale': 1.0, 'mode': 'fan_avg', 'distribution': 'uniform', 'seed': None}}, 'recurrent_initializer': {'class_name': 'Orthogonal', 'config': {'gain': 1.0, 'seed': None}}, 'bias_initializer': {'class_name': 'Zeros', 'config': {}}, 'unit_forget_bias': True, 'kernel_regularizer': None, 'recurrent_regularizer': None, 'bias_regularizer': None, 'activity_regularizer': None, 'kernel_constraint': None, 'recurrent_constraint': None, 'bias_constraint': None, 'dropout': 0.0, 'recurrent_dropout': 0.0, 'implementation': 1}
{'name': 'dense_1', 'trainable': True, 'units': 1, 'activation': 'sigmoid', 'use_bias': True, 'kernel_initializer': {'class_name': 'VarianceScaling', 'config': {'scale': 1.0, 'mode': 'fan_avg', 'distribution': 'uniform', 'seed': None}}, 'bias_initializer': {'class_name': 'Zeros', 'config': {}}, 'kernel_regularizer': None, 'bias_regularizer': None, 'activity_regularizer': None, 'kernel_constraint': None, 'bias_constraint': None}
```



In [0]:

```
# References
# https://stackoverflow.com/questions/43715047/keras-2-x-get-weights-of-layer

layer1_weights = model.layers[0].get_weights()[0]
layer2_weights = model.layers[1].get_weights()[0]
layer3_weights = model.layers[2].get_weights()[0]

print(" Embedding Layer Weight Matrix Shape")
print("="*125)
print(layer1_weights.shape)
print(" LSTM Weight Matrix Shape")
print("="*125)
print(layer2_weights.shape)
print(" Output Layer Weight Matrix Shape")
print("="*125)
print(layer3_weights.shape)
```

Embedding Layer Weight Matrix Shape

=====

=====

(64869, 32)

LSTM Weight Matrix Shape

=====

=====

(32, 400)

Output Layer Weight Matrix Shape

=====

=====

(100, 1)

◀ | ▶

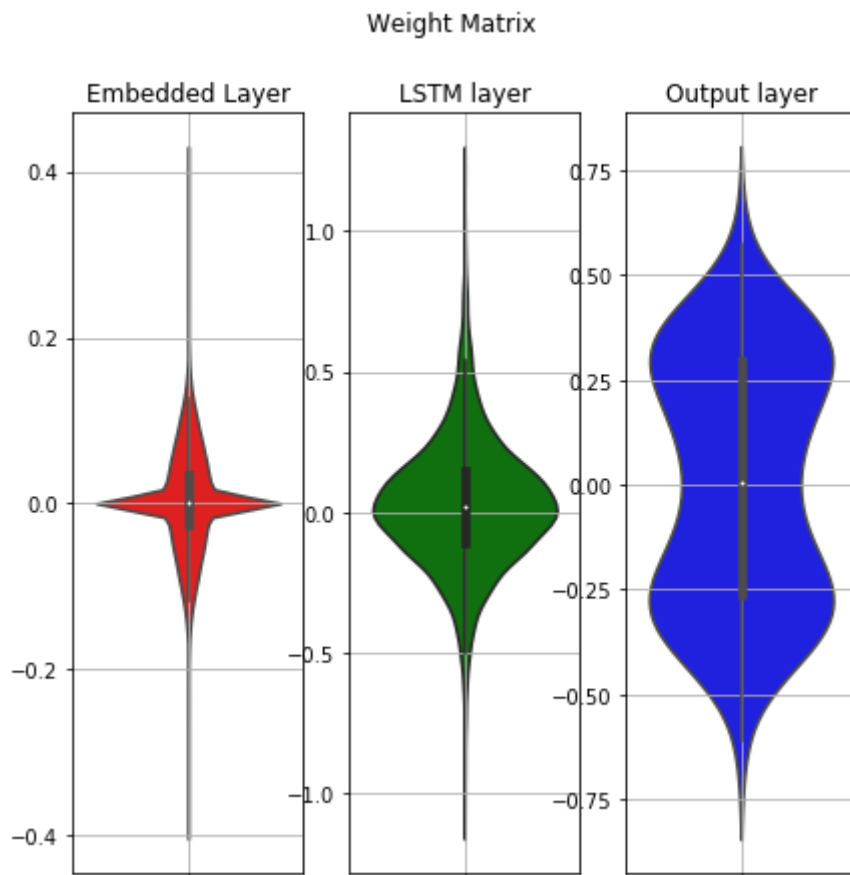
In [0]:

```
# References
# https://seaborn.pydata.org/generated/seaborn.violinplot.html
# https://matplotlib.org/api/_as_gen/matplotlib.pyplot.subplot.html

fig = plt.figure(1,figsize=(7,7))
fig.suptitle("Weight Matrix")
plt.subplot(1, 3, 1)
plt.title("Embedded Layer")
sns.violinplot(y=layer1_weights,color='r')
plt.grid()

plt.subplot(1,3,2)
plt.title("LSTM layer")
sns.violinplot(y=layer2_weights,color='g')
plt.grid()

plt.subplot(1,3,3)
plt.title("Output layer")
sns.violinplot(y=layer3_weights,color='b')
plt.grid()
```



#### 4.1.3 Model Evaluation:

In [0]:

```
# Refernces
# https://keras.io/models/model/#evaluate

evaluate_data=model.evaluate(x_test,y_test,verbose=0)
```

In [0]:

```

evaluate_loss = evaluate_data[0]
evaluate_acc = evaluate_data[1]

print(" Evaluate Loss")
print("="*100)
print(evaluate_loss)

print(" Evaluate Accuracy")
print("="*100)
print(evaluate_acc)

```

Evaluate Loss

0.2180799318348493

Evaluate Accuracy

0.9180613715941457

#### 4.1.4 Model Observation:

In [0]:

```
from prettytable import PrettyTable
```

In [0]:

```

a=PrettyTable()
b=PrettyTable()
b.field_names = ["Optimizer","Dropout rate","Batch Size","Epochs"]
b.add_row(["Adam","Nil",1000,10])
a.field_names = ["LSTM layer","units","Train loss","Val_loss","Test_loss","Train_acc","Val_acc","Test_acc"]
a.add_row([1,100,0.1074,0.2196 ,0.2180,0.9605,0.9183,0.9180])
print(b)
print(a)

```

```

+-----+-----+-----+-----+
| Optimizer | Dropout rate | Batch Size | Epochs |
+-----+-----+-----+-----+
| Adam      | Nil          | 1000       | 10      |
+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+
| LSTM layer | units | Train loss | Val_loss | Test_loss | Train_acc |
| Val_acc | Test_acc |
+-----+-----+-----+-----+-----+
| 1         | 100   | 0.1074    | 0.2196   | 0.218     | 0.9605    |
| 0.9183    | 0.918 |
+-----+-----+-----+-----+-----+
+-----+-----+

```

## 4.2 Two LSTM layer:

### 4.2.1 Sequence model:

In [0]:

```
from keras.layers import Dropout, BatchNormalization
```

In [0]:

```
# References
# https://keras.io/layers/embeddings/#embedding
# https://keras.io/layers/recurrent/#lstm
# https://stackoverflow.com/questions/42755820/how-to-use-return-sequences-option-a

# vocabulary size = 72395 + 1 = 72396 = input_dim

# model creation
model=Sequential()

# embedding layer
model.add(Embedding(input_dim= 72396,output_dim=32,input_length=100,embeddings_init

# Dropout layer
model.add(Dropout(0.4))

# LSTM layer
model.add(LSTM(50,dropout=0.4,recurrent_dropout=0.4,return_sequences=True))

# LSTM layer 2
model.add(LSTM(50,dropout=0.4,recurrent_dropout=0.4))

# Dropout layer 2
model.add(Dropout(0.4))

# Dense layer
model.add(Dense(1,activation="sigmoid"))
```

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensorflow\_backend.py:3445: 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`.

In [0]:

```
# Model Summary

model.summary()
```

Layer (type)	Output Shape	Param #
=====		
embedding_2 (Embedding)	(None, 100, 32)	2316672
dropout_1 (Dropout)	(None, 100, 32)	0
lstm_2 (LSTM)	(None, 100, 50)	16600
lstm_3 (LSTM)	(None, 50)	20200
dropout_2 (Dropout)	(None, 50)	0
dense_2 (Dense)	(None, 1)	51
=====		
Total params: 2,353,523		
Trainable params: 2,353,523		
Non-trainable params: 0		
=====		

In [0]:

```
# Model compilation

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

4.2.2 Model Training:

In [0]:

```
# model training
```

```
History=model.fit(x_train,y_train,batch_size=1000,epochs=10,verbose=1,validation_sp
```

Train on 203935 samples, validate on 87401 samples

Epoch 1/10

203935/203935 [=====] - 488s 2ms/step - loss: 0.3248 - acc: 0.8797 - val\_loss: 0.2151 - val\_acc: 0.9164

Epoch 2/10

203935/203935 [=====] - 484s 2ms/step - loss: 0.2148 - acc: 0.9173 - val\_loss: 0.2060 - val\_acc: 0.9203

Epoch 3/10

203935/203935 [=====] - 491s 2ms/step - loss: 0.1942 - acc: 0.9259 - val\_loss: 0.1960 - val\_acc: 0.9241

Epoch 4/10

203935/203935 [=====] - 489s 2ms/step - loss: 0.1845 - acc: 0.9292 - val\_loss: 0.1985 - val\_acc: 0.9243

Epoch 5/10

203935/203935 [=====] - 483s 2ms/step - loss: 0.1778 - acc: 0.9325 - val\_loss: 0.1944 - val\_acc: 0.9253

Epoch 6/10

203935/203935 [=====] - 481s 2ms/step - loss: 0.1721 - acc: 0.9345 - val\_loss: 0.1942 - val\_acc: 0.9256

Epoch 7/10

203935/203935 [=====] - 479s 2ms/step - loss: 0.1679 - acc: 0.9360 - val\_loss: 0.2005 - val\_acc: 0.9249

Epoch 8/10

203935/203935 [=====] - 485s 2ms/step - loss: 0.1637 - acc: 0.9382 - val\_loss: 0.1972 - val\_acc: 0.9249

Epoch 9/10

203935/203935 [=====] - 483s 2ms/step - loss: 0.1608 - acc: 0.9396 - val\_loss: 0.1986 - val\_acc: 0.9246

Epoch 10/10

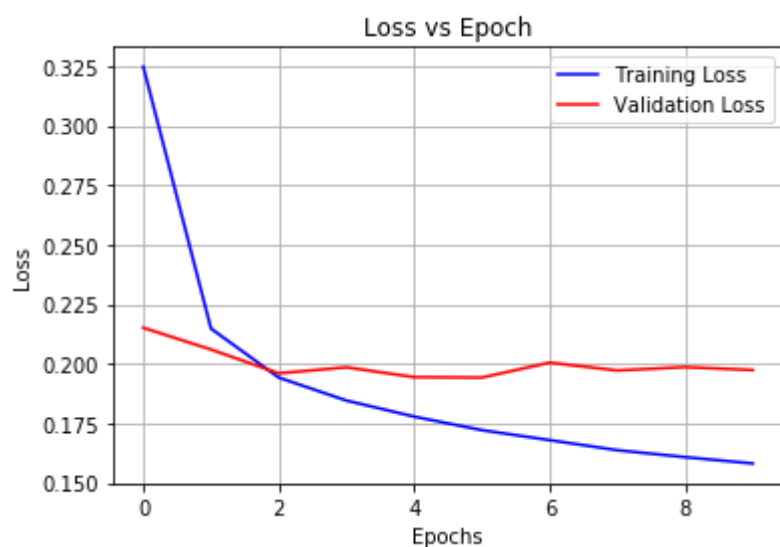
203935/203935 [=====] - 485s 2ms/step - loss: 0.1581 - acc: 0.9402 - val\_loss: 0.1973 - val\_acc: 0.9248

In [0]:

```
# References
# https://machinelearningmastery.com/display-deep-learning-model-training-history-i
# https://keras.io/models/sequential/

# Plotting Loss vs Epoch

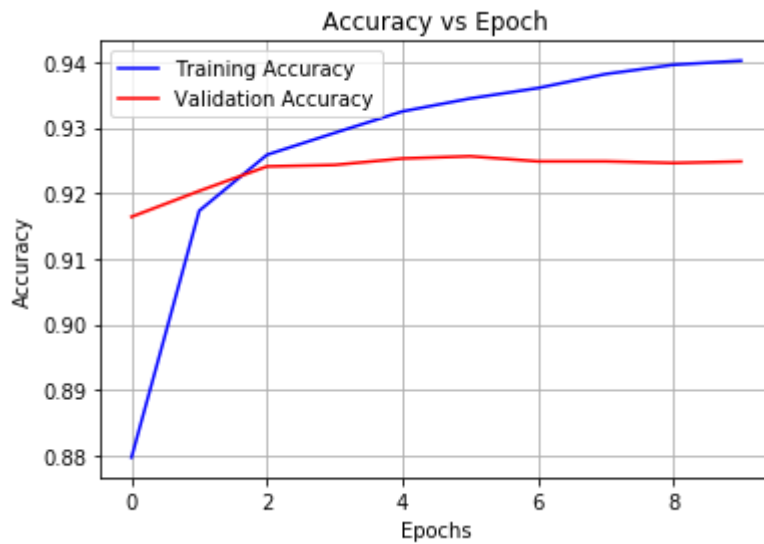
plt.close()
plt.plot(History.history['loss'], 'b', label="Training Loss")
plt.plot(History.history['val_loss'], 'r', label="Validation Loss")
plt.title("Loss vs Epoch")
plt.xlabel("Epochs")
plt.ylabel("Loss")
plt.legend()
plt.grid()
plt.show()
```



In [0]:

```
# Plotting Accuracy vs Epoch
```

```
plt.close()
plt.plot(History.history['acc'],'b',label="Training Accuracy")
plt.plot(History.history['val_acc'],'r',label="Validation Accuracy")
plt.title("Accuracy vs Epoch")
plt.xlabel("Epochs")
plt.ylabel("Accuracy")
plt.legend()
plt.grid()
plt.show()
```





In [0]:

```
# References
# https://stackoverflow.com/questions/43715047/keras-2-x-get-weights-of-layer
```

```
for layer in model.layers:
    print(layer.get_config())
```

```
{'name': 'embedding_2', 'trainable': True, 'batch_input_shape': (None, 100), 'dtype': 'float32', 'input_dim': 72396, 'output_dim': 32, 'embeddings_initializer': {'class_name': 'VarianceScaling', 'config': {'scale': 1.0, 'mode': 'fan_avg', 'distribution': 'normal', 'seed': None}}, 'embeddings_regularizer': None, 'activity_regularizer': None, 'embeddings_constraint': None, 'mask_zero': False, 'input_length': 100}
{'name': 'dropout_1', 'trainable': True, 'rate': 0.4, 'noise_shape': None, 'seed': None}
{'name': 'lstm_2', 'trainable': True, 'return_sequences': True, 'return_state': False, 'go_backwards': False, 'stateful': False, 'unroll': False, 'units': 50, 'activation': 'tanh', 'recurrent_activation': 'hard_sigmoid', 'use_bias': True, 'kernel_initializer': {'class_name': 'VarianceScaling', 'config': {'scale': 1.0, 'mode': 'fan_avg', 'distribution': 'uniform', 'seed': None}}, 'recurrent_initializer': {'class_name': 'Orthogonal', 'config': {'gain': 1.0, 'seed': None}}, 'bias_initializer': {'class_name': 'Zeros', 'config': {}}, 'unit_forget_bias': True, 'kernel_regularizer': None, 'recurrent_regularizer': None, 'bias_regularizer': None, 'activity_regularizer': None, 'kernel_constraint': None, 'recurrent_constraint': None, 'bias_constraint': None, 'dropout': 0.4, 'recurrent_dropout': 0.4, 'implementation': 1}
{'name': 'lstm_3', 'trainable': True, 'return_sequences': False, 'return_state': False, 'go_backwards': False, 'stateful': False, 'unroll': False, 'units': 50, 'activation': 'tanh', 'recurrent_activation': 'hard_sigmoid', 'use_bias': True, 'kernel_initializer': {'class_name': 'VarianceScaling', 'config': {'scale': 1.0, 'mode': 'fan_avg', 'distribution': 'uniform', 'seed': None}}, 'recurrent_initializer': {'class_name': 'Orthogonal', 'config': {'gain': 1.0, 'seed': None}}, 'bias_initializer': {'class_name': 'Zeros', 'config': {}}, 'unit_forget_bias': True, 'kernel_regularizer': None, 'recurrent_regularizer': None, 'bias_regularizer': None, 'activity_regularizer': None, 'kernel_constraint': None, 'recurrent_constraint': None, 'bias_constraint': None, 'dropout': 0.4, 'recurrent_dropout': 0.4, 'implementation': 1}
{'name': 'dropout_2', 'trainable': True, 'rate': 0.4, 'noise_shape': None, 'seed': None}
{'name': 'dense_2', 'trainable': True, 'units': 1, 'activation': 'sigmoid', 'use_bias': True, 'kernel_initializer': {'class_name': 'VarianceScaling', 'config': {'scale': 1.0, 'mode': 'fan_avg', 'distribution': 'uniform', 'seed': None}}, 'bias_initializer': {'class_name': 'Zeros', 'config': {}}, 'kernel_regularizer': None, 'bias_regularizer': None, 'activity_regularizer': None, 'kernel_constraint': None, 'bias_constraint': None}
```

In [0]:

```
# References
# https://stackoverflow.com/questions/43715047/keras-2-x-get-weights-of-layer

layer1_weights = model.layers[0].get_weights()[0]
layer2_weights = model.layers[2].get_weights()[0]
layer3_weights = model.layers[3].get_weights()[0]
layer4_weights = model.layers[5].get_weights()[0]

print(" Embedding Layer Weight Matrix Shape")
print("="*125)
print(layer1_weights.shape)
print(" LSTM 1  Weight Matrix Shape")
print("="*125)
print(layer2_weights.shape)
print(" LSTM 2 Weight Matrix Shape")
print("="*125)
print(layer3_weights.shape)
print(" Output Layer Weight Matrix Shape")
print("="*125)
print(layer4_weights.shape)
```

Embedding Layer Weight Matrix Shape

```
=====
(72396, 32)
```

LSTM 1 Weight Matrix Shape

```
=====
(32, 200)
```

LSTM 2 Weight Matrix Shape

```
=====
(50, 200)
```

Output Layer Weight Matrix Shape

```
=====
(50, 1)
```



In [0]:

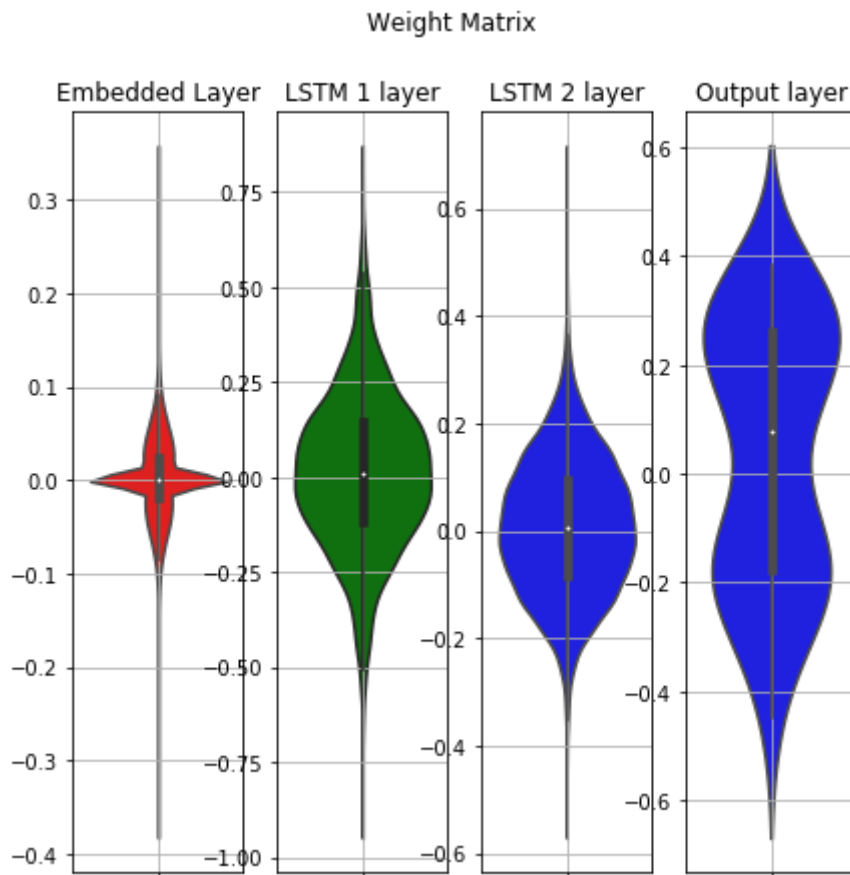
```
# References
# https://seaborn.pydata.org/generated/seaborn.violinplot.html
# https://matplotlib.org/api/_as_gen/matplotlib.pyplot.subplot.html

fig = plt.figure(1,figsize=(7,7))
fig.suptitle("Weight Matrix")
plt.subplot(1, 4, 1)
plt.title("Embedded Layer")
sns.violinplot(y=layer1_weights,color='r')
plt.grid()

plt.subplot(1,4,2)
plt.title("LSTM 1 layer")
sns.violinplot(y=layer2_weights,color='g')
plt.grid()

plt.subplot(1,4,3)
plt.title("LSTM 2 layer")
sns.violinplot(y=layer3_weights,color='b')
plt.grid()

plt.subplot(1,4,4)
plt.title("Output layer")
sns.violinplot(y=layer4_weights,color='b')
plt.grid()
```



#### 4.2.3 Model Evaluation:

In [0]:

```
# Refernces
# https://keras.io/models/model/#evaluate

evaluate_data=model.evaluate(x_test,y_test,verbose=0)
```

In [0]:

```
evaluate_loss = evaluate_data[0]
evaluate_acc = evaluate_data[1]

print(" Evaluate Loss")
print("="*100)
print(evaluate_loss)

print(" Evaluate Accuracy")
print("="*100)
print(evaluate_acc)
```

Evaluate Loss

0.1919281865505083

Evaluate Accuracy

0.9259010091310442

#### 4.2.4 Model Observation:

In [18]:

```
a=PrettyTable()
b=PrettyTable()
b.field_names = ["Optimizer","Dropout rate","Batch Size","Epochs"]
b.add_row(["Adam","0.4",1000,10])
a.field_names = ["LSTM layer","Units","Train loss","Val_loss","Test_loss","Train_acc","Val_acc","Test_acc"]
a.add_row([2,50,0.1581,0.1973 ,0.1919,0.9402,0.9248,0.9259])
print(b)
print(a)
```

```
+-----+-----+-----+-----+
| Optimizer | Dropout rate | Batch Size | Epochs |
+-----+-----+-----+-----+
| Adam      | 0.4          | 1000       | 10      |
+-----+-----+-----+-----+

+-----+-----+-----+-----+-----+-----+
| LSTM layer | Train loss | Val_loss | Test_loss | Train_acc | Val_acc |
+-----+-----+-----+-----+-----+-----+
| 2          | 0.1581     | 0.1973   | 0.1919    | 0.9402    | 0.9248  |
+-----+-----+-----+-----+-----+-----+
| 50         | 0.9259     |          |          |          |          |
+-----+-----+-----+-----+-----+-----+
| 50         |          |          |          |          |          |
+-----+-----+-----+-----+-----+-----+
```

## 4.3 Three LSTM layer:

### 4.3.1 Sequence model:

In [0]:

```
from keras.layers import Dropout, BatchNormalization
from keras import regularizers
```

In [0]:

```
# References
# https://keras.io/layers/embeddings/#embedding
# https://keras.io/layers/recurrent/#lstm
# https://stackoverflow.com/questions/42755820/how-to-use-return-sequences-option-a

# vocabulary size = 72395 +1 = 72396 = input_dim

# model creation
model=Sequential()

# embedding layer
model.add(Embedding(input_dim= 72396,output_dim=32,input_length=100,embeddings_init

# BN Layer
model.add(BatchNormalization())

# Dropout layer
model.add(Dropout(0.4))

# LSTM layer
model.add(LSTM(50,dropout=0.4,recurrent_dropout=0.4,return_sequences=True,kernel_re

# LSTM layer 2
model.add(LSTM(50,dropout=0.4,recurrent_dropout=0.4,return_sequences=True,kernel_re

# LSTM layer 3
model.add(LSTM(50,dropout=0.4,recurrent_dropout=0.4,kernel_regularizer=regularizers

# Dropout layer 2
model.add(Dropout(0.4))

# BN Layer 2
model.add(BatchNormalization())

# Dense layer
model.add(Dense(1,activation="sigmoid"))
```

In [25]:

```
# Model Summary
```

```
model.summary()
```

Layer (type)	Output Shape	Param #
=====		
embedding_2 (Embedding)	(None, 100, 32)	2316672
batch_normalization_3 (Batch Normalization)	(None, 100, 32)	128
dropout_3 (Dropout)	(None, 100, 32)	0
lstm_4 (LSTM)	(None, 100, 50)	16600
lstm_5 (LSTM)	(None, 100, 50)	20200
lstm_6 (LSTM)	(None, 50)	20200
dropout_4 (Dropout)	(None, 50)	0
batch_normalization_4 (Batch Normalization)	(None, 50)	200
dense_2 (Dense)	(None, 1)	51
=====		
Total params: 2,374,051		
Trainable params: 2,373,887		
Non-trainable params: 164		

In [0]:

```
# Model compilation
```

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

#### 4.3.2 Model Training:

In [27]:

```
# model training
```

```
History=model.fit(x_train,y_train,batch_size=1000,epochs=10,verbose=1,validation_sp
```

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/tensorflow/python/ops/math\_ops.py:3066: to\_int32 (from tensorflow.python.ops.math\_ops) is deprecated and will be removed in a future version.

Instructions for updating:

Use tf.cast instead.

Train on 203935 samples, validate on 87401 samples

Epoch 1/10

203935/203935 [=====] - 654s 3ms/step - loss: 0.9754 - acc: 0.8262 - val\_loss: 0.3066 - val\_acc: 0.9101

Epoch 2/10

203935/203935 [=====] - 647s 3ms/step - loss: 0.2609 - acc: 0.9082 - val\_loss: 0.2312 - val\_acc: 0.9097

Epoch 3/10

203935/203935 [=====] - 646s 3ms/step - loss: 0.2222 - acc: 0.9185 - val\_loss: 0.2213 - val\_acc: 0.9149

Epoch 4/10

203935/203935 [=====] - 647s 3ms/step - loss: 0.2098 - acc: 0.9229 - val\_loss: 0.2235 - val\_acc: 0.9145

Epoch 5/10

203935/203935 [=====] - 646s 3ms/step - loss: 0.2011 - acc: 0.9263 - val\_loss: 0.2234 - val\_acc: 0.9138

Epoch 6/10

203935/203935 [=====] - 647s 3ms/step - loss: 0.1954 - acc: 0.9284 - val\_loss: 0.2182 - val\_acc: 0.9186

Epoch 7/10

203935/203935 [=====] - 647s 3ms/step - loss: 0.1879 - acc: 0.9321 - val\_loss: 0.2207 - val\_acc: 0.9159

Epoch 8/10

203935/203935 [=====] - 645s 3ms/step - loss: 0.1843 - acc: 0.9338 - val\_loss: 0.2250 - val\_acc: 0.9165

Epoch 9/10

203935/203935 [=====] - 647s 3ms/step - loss: 0.1789 - acc: 0.9356 - val\_loss: 0.2158 - val\_acc: 0.9205

Epoch 10/10

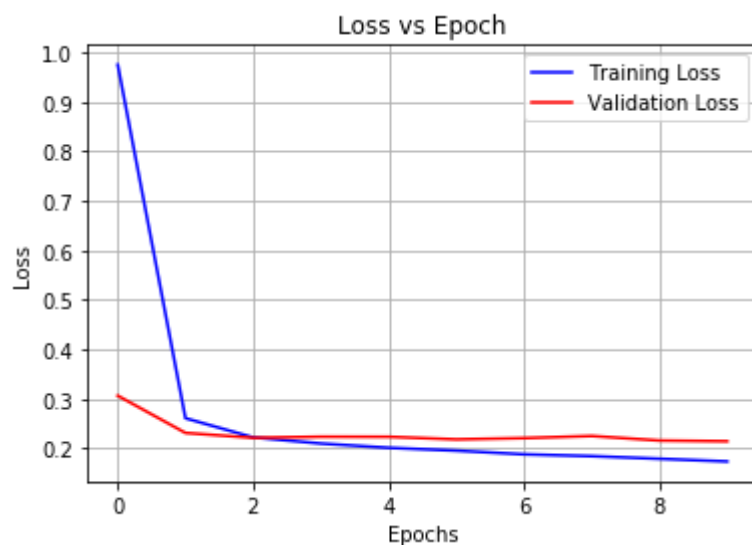
203935/203935 [=====] - 647s 3ms/step - loss: 0.1735 - acc: 0.9376 - val\_loss: 0.2143 - val\_acc: 0.9221

In [28]:

```
# References
# https://machinelearningmastery.com/display-deep-learning-model-training-history-i
# https://keras.io/models/sequential/

# Plotting Loss vs Epoch

plt.close()
plt.plot(History.history['loss'],'b',label="Training Loss")
plt.plot(History.history['val_loss'],'r',label="Validation Loss")
plt.title("Loss vs Epoch")
plt.xlabel("Epochs")
plt.ylabel("Loss")
plt.legend()
plt.grid()
plt.show()
```

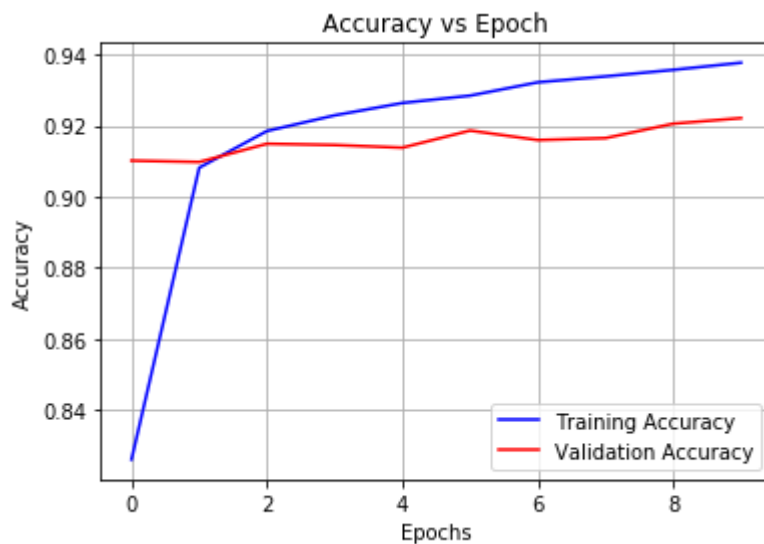




In [29]:

```
# Plotting Accuracy vs Epoch
```

```
plt.close()
plt.plot(History.history['acc'],'b',label="Training Accuracy")
plt.plot(History.history['val_acc'],'r',label="Validation Accuracy")
plt.title("Accuracy vs Epoch")
plt.xlabel("Epochs")
plt.ylabel("Accuracy")
plt.legend()
plt.grid()
plt.show()
```




In [30]:

```
# References
# https://stackoverflow.com/questions/43715047/keras-2-x-get-weights-of-layer
```

```
for layer in model.layers:
    print(layer.get_config())
```

```
{'name': 'embedding_2', 'trainable': True, 'batch_input_shape': (None, 100), 'dtype': 'float32', 'input_dim': 72396, 'output_dim': 32, 'embeddings_initializer': {'class_name': 'VarianceScaling', 'config': {'scale': 1.0, 'mode': 'fan_avg', 'distribution': 'normal', 'seed': None}}, 'embeddings_regularizer': None, 'activity_regularizer': None, 'embeddings_constraint': None, 'mask_zero': False, 'input_length': 100}
{'name': 'batch_normalization_3', 'trainable': True, 'axis': -1, 'momentum': 0.99, 'epsilon': 0.001, 'center': True, 'scale': True, 'beta_initializer': {'class_name': 'Zeros', 'config': {}}, 'gamma_initializer': {'class_name': 'Ones', 'config': {}}, 'moving_mean_initializer': {'class_name': 'Zeros', 'config': {}}, 'moving_variance_initializer': {'class_name': 'Ones', 'config': {}}, 'beta_regularizer': None, 'gamma_regularizer': None, 'beta_constraint': None, 'gamma_constraint': None}
{'name': 'dropout_3', 'trainable': True, 'rate': 0.4, 'noise_shape': None, 'seed': None}
{'name': 'lstm_4', 'trainable': True, 'return_sequences': True, 'return_state': False, 'go_backwards': False, 'stateful': False, 'unroll': False, 'units': 50, 'activation': 'tanh', 'recurrent_activation': 'hard_sigmoid', 'use_bias': True, 'kernel_initializer': {'class_name': 'VarianceScaling', 'config': {'scale': 1.0, 'mode': 'fan_avg', 'distribution': 'uniform', 'seed': None}}, 'recurrent_initializer': {'class_name': 'Orthogonal', 'config': {'gain': 1.0, 'seed': None}}, 'bias_initializer': {'class_name': 'Zeros', 'config': {}}, 'unit_forget_bias': True, 'kernel_regularizer': {'class_name': 'L1L2', 'config': {'l1': 0.0, 'l2': 0.009999999776482582}}, 'recurrent_regularizer': None, 'bias_regularizer': None, 'activity_regularizer': None, 'kernel_constraint': None, 'recurrent_constraint': None, 'bias_constraint': None, 'dropout': 0.4, 'recurrent_dropout': 0.4, 'implementation': 1}
{'name': 'lstm_5', 'trainable': True, 'return_sequences': True, 'return_state': False, 'go_backwards': False, 'stateful': False, 'unroll': False, 'units': 50, 'activation': 'tanh', 'recurrent_activation': 'hard_sigmoid', 'use_bias': True, 'kernel_initializer': {'class_name': 'VarianceScaling', 'config': {'scale': 1.0, 'mode': 'fan_avg', 'distribution': 'uniform', 'seed': None}}, 'recurrent_initializer': {'class_name': 'Orthogonal', 'config': {'gain': 1.0, 'seed': None}}, 'bias_initializer': {'class_name': 'Zeros', 'config': {}}, 'unit_forget_bias': True, 'kernel_regularizer': {'class_name': 'L1L2', 'config': {'l1': 0.0, 'l2': 0.009999999776482582}}, 'recurrent_regularizer': None, 'bias_regularizer': None, 'activity_regularizer': None, 'kernel_constraint': None, 'recurrent_constraint': None, 'bias_constraint': None, 'dropout': 0.4, 'recurrent_dropout': 0.4, 'implementation': 1}
{'name': 'lstm_6', 'trainable': True, 'return_sequences': False, 'return_state': False, 'go_backwards': False, 'stateful': False, 'unroll': False, 'units': 50, 'activation': 'tanh', 'recurrent_activation': 'hard_sigmoid', 'use_bias': True, 'kernel_initializer': {'class_name': 'VarianceScaling', 'config': {'scale': 1.0, 'mode': 'fan_avg', 'distribution': 'uniform', 'seed': None}}, 'recurrent_initializer': {'class_name': 'Orthogonal', 'config': {'gain': 1.0, 'seed': None}}, 'bias_initializer': {'class_name': 'Zeros', 'config': {}}, 'unit_forget_bias': True, 'kernel_regularizer': {'class_name': 'L1L2', 'config':
```

```
{'l1': 0.0, 'l2': 0.009999999776482582}}, 'recurrent_regularizer': No
ne, 'bias_regularizer': None, 'activity_regularizer': None, 'kernel_c
onstraint': None, 'recurrent_constraint': None, 'bias_constraint': No
ne, 'dropout': 0.4, 'recurrent_dropout': 0.4, 'implementation': 1}
{'name': 'dropout_4', 'trainable': True, 'rate': 0.4, 'noise_shape':
None, 'seed': None}
{'name': 'batch_normalization_4', 'trainable': True, 'axis': -1, 'mom
entum': 0.99, 'epsilon': 0.001, 'center': True, 'scale': True, 'beta_
initializer': {'class_name': 'Zeros', 'config': {}}, 'gamma_initializ
er': {'class_name': 'Ones', 'config': {}}, 'moving_mean_initializer':
{'class_name': 'Zeros', 'config': {}}, 'moving_variance_initializer':
{'class_name': 'Ones', 'config': {}}, 'beta_regularizer': None, 'gamm
a_regularizer': None, 'beta_constraint': None, 'gamma_constraint': No
ne}
{'name': 'dense_2', 'trainable': True, 'units': 1, 'activation': 'sig
moid', 'use_bias': True, 'kernel_initializer': {'class_name': 'Varian
ceScaling', 'config': {'scale': 1.0, 'mode': 'fan_avg', 'distributio
n': 'uniform', 'seed': None}}, 'bias_initializer': {'class_name': 'Ze
ros', 'config': {}}, 'kernel_regularizer': None, 'bias_regularizer':
None, 'activity_regularizer': None, 'kernel_constraint': None, 'bias_
constraint': None}
```



In [31]:

```
# References
# https://stackoverflow.com/questions/43715047/keras-2-x-get-weights-of-layer
```

```
layer1_weights = model.layers[0].get_weights()[0]
layer2_weights = model.layers[3].get_weights()[0]
layer3_weights = model.layers[4].get_weights()[0]
layer4_weights = model.layers[5].get_weights()[0]
layer5_weights = model.layers[8].get_weights()[0]
```

```
print(" Embedding Layer Weight Matrix Shape")
print("=="*125)
print(layer1_weights.shape)
print(" LSTM 1 Weight Matrix Shape")
print("=="*125)
print(layer2_weights.shape)
print(" LSTM 2 Weight Matrix Shape")
print("=="*125)
print(layer3_weights.shape)
print(" LSTM 3 Weight Matrix Shape")
print("=="*125)
print(layer4_weights.shape)
print(" Output Layer Weight Matrix Shape")
print("=="*125)
print(layer5_weights.shape)
```

```
Embedding Layer Weight Matrix Shape
```

```
=====
=====
(72396, 32)
```

```
LSTM 1 Weight Matrix Shape
```

```
=====
=====
(32, 200)
```

```
LSTM 2 Weight Matrix Shape
```

```
=====
=====
(50, 200)
```

```
LSTM 3 Weight Matrix Shape
```

```
=====
=====
(50, 200)
```

```
Output Layer Weight Matrix Shape
```

```
=====
=====
(50, 1)
```

◀  ▶

In [32]:

```
# References
# https://seaborn.pydata.org/generated/seaborn.violinplot.html
# https://matplotlib.org/api/_as_gen/matplotlib.pyplot.subplot.html

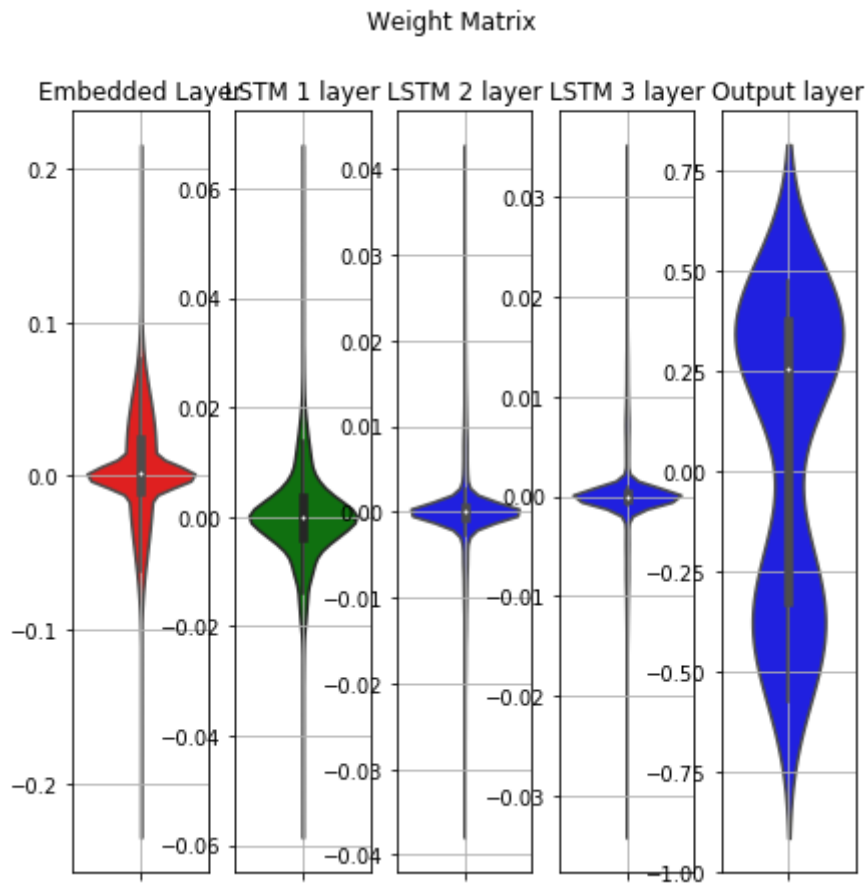
fig = plt.figure(1,figsize=(7,7))
fig.suptitle("Weight Matrix")
plt.subplot(1, 5, 1)
plt.title("Embedded Layer")
sns.violinplot(y=layer1_weights,color='r')
plt.grid()

plt.subplot(1,5,2)
plt.title("LSTM 1 layer")
sns.violinplot(y=layer2_weights,color='g')
plt.grid()

plt.subplot(1,5,3)
plt.title("LSTM 2 layer")
sns.violinplot(y=layer3_weights,color='b')
plt.grid()

plt.subplot(1,5,4)
plt.title("LSTM 3 layer")
sns.violinplot(y=layer4_weights,color='b')
plt.grid()

plt.subplot(1,5,5)
plt.title("Output layer")
sns.violinplot(y=layer5_weights,color='b')
plt.grid()
```



### 4.3.3 Model Evaluation:

In [0]:

```
# Refernces
# https://keras.io/models/model/#evaluate

evaluate_data=model.evaluate(x_test,y_test,verbose=0)
```

In [34]:

```
evaluate_loss = evaluate_data[0]
evaluate_acc  = evaluate_data[1]

print(" Evaluate Loss")
print("="*100)
print(evaluate_loss)

print(" Evaluate Accuracy")
print("="*100)
print(evaluate_acc)
```

```
Evaluate Loss
=====
=====
0.21034497041287203
Evaluate Accuracy
=====
=====
0.9225784307003447
```

### 4.3.4 Model Observation:

In [36]:

```
a=PrettyTable()
b=PrettyTable()
b.field_names = ["Optimizer","Dropout rate","Batch Size","Epochs"]
b.add_row(["Adam","0.4",1000,10])
a.field_names = ["LSTM layer","Units","Train loss","Val_loss","Test_loss","Train_ac"]
a.add_row([3,50,0.1735,0.2143,0.2103,0.9376,0.9221,0.9225])
print(b)
print(a)
```

+-----+-----+-----+-----+			
Optimizer	Dropout rate	Batch Size	Epochs
+-----+-----+-----+-----+			
Adam	0.4	1000	10
+-----+-----+-----+-----+			
+-----+-----+-----+-----+			
+-----+-----+			
LSTM layer	Units	Train loss	Val_loss
Val_acc	Test_acc	Test_loss	Train_acc
+-----+-----+-----+-----+			
+-----+-----+			
3	50	0.1735	0.2143
0.9221	0.9225	0.2103	0.9376
+-----+-----+-----+-----+			
+-----+-----+-----+-----+			

## 5. Conclusion:

In [37]:

```

a=PrettyTable()
b=PrettyTable()
b.field_names = ["LSTM layer","Optimizer","Dropout rate","Batch Size","Epochs","Kernel Regularizer","BN"]
b.add_row([1,"Adam","Nil",1000,10,"Nil","Nil"])
b.add_row([2,"Adam",0.4,1000,10,"Nil","Nil"])
b.add_row([3,"Adam",0.4,1000,10,0.01,"Yes"])
a.field_names = ["LSTM layer","Units","Train loss","Val_loss","Test_loss","Train_acc","Val_acc","Test_acc"]
a.add_row([1,100,0.1074,0.2196,0.2180,0.9605,0.9183,0.9180])
a.add_row([2,50,0.1581,0.1973,0.1919,0.9402,0.9248,0.9259])
a.add_row([3,50,0.1735,0.2143,0.2103,0.9376,0.9221,0.9225])
print(b)
print("Architecture loss and accuracy comparision")
print("="*100)
print(a)

```

```

+-----+-----+-----+-----+-----+-----+
+-----+-----+
| LSTM layer | Optimizer | Dropout rate | Batch Size | Epochs | Kernel Regularizer | BN |
+-----+-----+-----+-----+-----+-----+
+-----+-----+
|      1      |    Adam   |      Nil      |    1000    |    10    |
Nil          | Nil       |
|      2      |    Adam   |      0.4      |    1000    |    10    |
Nil          | Nil       |
|      3      |    Adam   |      0.4      |    1000    |    10    |
0.01         | Yes       |
+-----+-----+-----+-----+-----+-----+
+-----+-----+
Architecture loss and accuracy comparision
=====
=====
+-----+-----+-----+-----+-----+-----+
+-----+-----+
| LSTM layer | Units | Train loss | Val_loss | Test_loss | Train_acc |
| Val_acc | Test_acc |
+-----+-----+-----+-----+-----+-----+
+-----+-----+
|      1      |    100   |    0.1074   |    0.2196   |    0.218   |    0.9605   |
| 0.9183 | 0.918   |
|      2      |    50    |    0.1581   |    0.1973   |    0.1919   |    0.9402   |
| 0.9248 | 0.9259   |
|      3      |    50    |    0.1735   |    0.2143   |    0.2103   |    0.9376   |
| 0.9221 | 0.9225   |
+-----+-----+-----+-----+-----+-----+
+-----+-----+

```

### Data Preparation:

After conversion of IMDB dataset format, the pickle files were loaded by using joblib. Then the data was splitted as a train and test data.

### LSTM layers:



Using those splitted data's, padding and embedding layer, Different architecture of LSTM(1 layer, 2 layer, 3 layer) was implemented.

**Loss vs Epoch plot:**

Loss(Train and Validation) vs Epoch graph has been plotted successfully for each of the Architecture.

**Weight Plot:**

Weight (Weight matrix got After the Optimization) graph has been plotted successfully for each of the Architecture.