

```
In [7]: import warnings
from sklearn.exceptions import DataConversionWarning
warnings.filterwarnings(action='ignore', category=DataConversionWarning)

import sqlite3
import datetime as dt
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
from collections import Counter
from itertools import islice

from sklearn.model_selection import train_test_split
from keras.models import Sequential
from keras.preprocessing import sequence
from keras.initializers import he_normal
from keras.layers import BatchNormalization, Dense, Dropout, Flatten, LSTM
from keras.layers.embeddings import Embedding
from keras.regularizers import L1L2
```

Using TensorFlow backend.

```
In [8]: connection = sqlite3.connect('final.sqlite')
# Load data into pandas dataframe.
reviews_df = pd.read_sql_query(""" SELECT * FROM Reviews """,connection)

# Drop index column
reviews_df = reviews_df.drop(columns=['index'])
reviews_df['Time'] = reviews_df[['Time']].applymap(lambda x: dt.datetime.fromtimestamp(x))

reviews_df=reviews_df.sample(50000)

# Sort the data on the basis of time.
reviews_df = reviews_df.sort_values(by=['Time'])
cleaned_text = reviews_df['CleanedText'].values

print("Dataset Shape : \n",cleaned_text.shape)
```

Dataset Shape :  
(50000,)

```
In [9]: all_words=[]
        for sentence in cleaned_text:
            words = sentence.split()
            all_words += words

        print("Shape of the data : ",cleaned_text.shape)
        print("Number of sentences present in complete dataset : ",len(all_words))

        counts = Counter(all_words)
        print("Number of unique words present in whole corpus: ",len(counts.most_common()))
        vocab_size = len(counts.most_common()) + 1
        top_words_count = 5000
        sorted_words = counts.most_common(top_words_count)

        word_index_lookup = dict()
        i = 1
        for word,frequency in sorted_words:
            word_index_lookup[word] = i
            i += 1

        print()
        print("Top 25 words with their frequencies:")
        print(counts.most_common(25))
        print()
        print("Top 25 words with their index:")
        print(list(islice(word_index_lookup.items(), 25)))
```

```
Shape of the data : (50000,)
Number of sentences present in complete dataset : 1793783
Number of unique words present in whole corpus: 27002
```

Top 25 words with their frequencies:

```
[('like', 21033), ('tast', 21020), ('tea', 17974), ('good', 16777), ('flavor', 16771), ('great', 15565), ('product', 14812), ('one', 14672), ('love', 14609), ('use', 14514), ('tri', 12949), ('make', 11617), ('get', 10395), ('coffe', 8748), ('food', 8308), ('eat', 8283), ('would', 8206), ('buy', 7973), ('time', 7922), ('best', 7750), ('realli', 7547), ('also', 7425), ('find', 7365), ('dont', 7168), ('amazon', 7096)]
```

Top 25 words with their index:

```
[('like', 1), ('tast', 2), ('tea', 3), ('good', 4), ('flavor', 5), ('great', 6), ('product', 7), ('one', 8), ('love', 9), ('use', 10), ('tri', 11), ('make', 12), ('get', 13), ('coffe', 14), ('food', 15), ('eat', 16), ('would', 17), ('buy', 18), ('time', 19), ('best', 20), ('realli', 21), ('also', 22), ('find', 23), ('don t', 24), ('amazon', 25)]
```

```
In [10]: def apply_text_index(row):
          holder = []
          for word in row['CleanedText'].split():
              if word in word_index_lookup:
                  holder.append(word_index_lookup[word])
              else:
                  holder.append(0)
          return holder

reviews_df['CleanedText_Index'] = reviews_df.apply(lambda row: apply_text_index(row),axis=1)
reviews_df.head(5)
```

Out[10]:

	<b>Id</b>	<b>ProductId</b>	<b>UserId</b>	<b>ProfileName</b>	<b>HelpfulnessNumerator</b>	<b>HelpfulnessDenominator</b>	<b>S</b>
<b>0</b>	150524	0006641040	ACITT7DI6IDDL	shari zychinski	0	0	pos
<b>30</b>	150501	0006641040	AJ46FKXOVC7NR	Nicholas A Mesiano	2	2	pos
<b>424</b>	451856	B00004CXX9	AIUWLEQ1ADEG5	Elizabeth Medina	0	0	pos
<b>330</b>	374359	B00004CI84	A344SMIA5JECGM	Vincent P. Ross	1	2	pos
<b>423</b>	451855	B00004CXX9	AJH6LUC1UT1ON	The Phantom of the Opera	0	0	pos

```
In [11]: reviews_df['Score'] = reviews_df['Score'].map(lambda x : 1 if x == 'positive' else 0)
reviews_df.head(5)
```

```
Out[11]:
```

	Id	ProductId	UserId	ProfileName	HelpfulnessNumerator	HelpfulnessDenominator	Score
0	150524	0006641040	ACITT7DI6IDDL	shari zychinski	0	0	
30	150501	0006641040	AJ46FKXOVC7NR	Nicholas A Mesiano	2	2	
424	451856	B00004CXX9	AIUWLEQ1ADEG5	Elizabeth Medina	0	0	
330	374359	B00004CI84	A344SMIA5JECGM	Vincent P. Ross	1	2	
423	451855	B00004CXX9	AJH6LUC1UT1ON	The Phantom of the Opera	0	0	

```
In [12]: x_train, x_test, y_train, y_test = train_test_split(reviews_df['CleanedText_Index']
.values,
                                                            reviews_df['Score'],
                                                            test_size=0.3,
                                                            shuffle=False,
                                                            random_state=0)
```

```
In [12]: x_train, x_test, y_train, y_test = train_test_split(reviews_df['CleanedText_Index']
.values,
                                                            reviews_df['Score'],
                                                            test_size=0.3,
                                                            shuffle=False,
                                                            random_state=0)
```

```
In [13]: print("Total number words present in first review:\n",len(x_train[1]))
print()
print("List of word indexes present in first review:\n", x_train[1])
print()
```

```
Total number words present in first review:
30
```

```
List of word indexes present in first review:
[499, 158, 781, 1094, 0, 43, 323, 836, 1146, 586, 74, 45, 3385, 122, 10, 2759,
797, 3581, 1963, 2869, 4411, 343, 116, 856, 76, 523, 797, 615, 561, 251]
```



```
In [15]: batch_size = 192

# Number of time whole data is trained
epochs = 10

# Embedding vector size
embedding_vecor_length = 32

# Bias regularizer value - we will use elasticnet
reg = L1L2(0.01, 0.01)

# Plot train and cross validation loss
def plot_train_cv_loss(trained_model, epochs, colors=['b']):
    fig, ax = plt.subplots(1,1)
    ax.set_xlabel('epoch')
    ax.set_ylabel('Categorical Crossentropy Loss')
    x_axis_values = list(range(1,epochs+1))

    validation_loss = trained_model.history['val_loss']
    train_loss = trained_model.history['loss']

    ax.plot(x_axis_values, validation_loss, 'b', label="Validation Loss")
    ax.plot(x_axis_values, train_loss, 'r', label="Train Loss")
    plt.legend()
    plt.grid()
    fig.canvas.draw()
```

```
In [16]: model = Sequential()

# Add Embedding Layer
model.add(Embedding(vocab_size, embedding_vector_length, input_length=max_review_length))

# Add batch normalization
model.add(BatchNormalization())

# Add dropout
model.add(Dropout(0.20))

# Add LSTM Layer
model.add(LSTM(100))

# Add dropout
model.add(Dropout(0.20))

# Add Dense Layer
model.add(Dense(1, activation='sigmoid'))

# Summary of the model
print("Model Summary: \n")
model.summary()
print()
print()

# Compile the model
model.compile(loss='binary_crossentropy', optimizer='adam', metrics=['accuracy'])

# Run the model
trained_model = model.fit(x_train, np.array(y_train), batch_size = batch_size, epochs = epochs, verbose=1, validation_data=(x_test, y_test))
```

Model Summary:

Layer (type)	Output Shape	Param #
embedding_1 (Embedding)	(None, 500, 32)	864096
batch_normalization_1 (Batch Normalization)	(None, 500, 32)	128
dropout_1 (Dropout)	(None, 500, 32)	0
lstm_1 (LSTM)	(None, 100)	53200
dropout_2 (Dropout)	(None, 100)	0
dense_1 (Dense)	(None, 1)	101
Total params: 917,525		
Trainable params: 917,461		
Non-trainable params: 64		

Train on 35000 samples, validate on 15000 samples

Epoch 1/10

35000/35000 [=====] - 337s 10ms/step - loss: 0.2658 - acc: 0.9049 - val\_loss: 0.3315 - val\_acc: 0.8988

Epoch 2/10

35000/35000 [=====] - 315s 9ms/step - loss: 0.1607 - acc: 0.9401 - val\_loss: 0.2279 - val\_acc: 0.9227

Epoch 3/10

35000/35000 [=====] - 357s 10ms/step - loss: 0.1282 - acc: 0.9534 - val\_loss: 0.2154 - val\_acc: 0.9224

Epoch 4/10

35000/35000 [=====] - 365s 10ms/step - loss: 0.1018 - acc: 0.9640 - val\_loss: 0.2437 - val\_acc: 0.9215

Epoch 5/10

35000/35000 [=====] - 396s 11ms/step - loss: 0.0813 - acc: 0.9717 - val\_loss: 0.2668 - val\_acc: 0.9197

Epoch 6/10

35000/35000 [=====] - 412s 12ms/step - loss: 0.0673 - acc: 0.9769 - val\_loss: 0.2840 - val\_acc: 0.9187

Epoch 7/10

35000/35000 [=====] - 633s 18ms/step - loss: 0.0572 - acc: 0.9803 - val\_loss: 0.3317 - val\_acc: 0.9186

Epoch 8/10

35000/35000 [=====] - 836s 24ms/step - loss: 0.0521 - acc: 0.9824 - val\_loss: 0.3433 - val\_acc: 0.9150

Epoch 9/10

35000/35000 [=====] - 939s 27ms/step - loss: 0.0399 - acc: 0.9863 - val\_loss: 0.3646 - val\_acc: 0.9120

Epoch 10/10

35000/35000 [=====] - 655s 19ms/step - loss: 0.0348 - acc: 0.9880 - val\_loss: 0.3746 - val\_acc: 0.9116

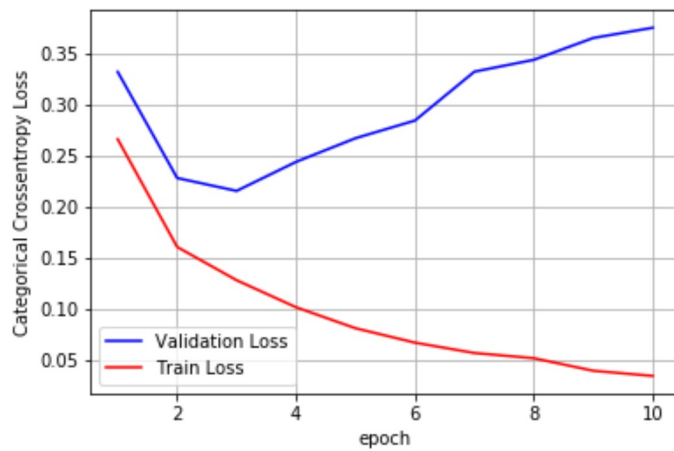
```
In [17]: score = model.evaluate(x_test, y_test, verbose=0)
print('Test accuracy: {0:.2f}%'.format(score[1]*100))
```

Test accuracy: 91.16%



```
In [18]: print()
print()

# Plot train and cross validation error
plot_train_cv_loss(trained_model, epochs)
```



## 2 LSTM Layer

```
In [19]: %%time
# Instantiate sequential model
model = Sequential()

# Add Embedding Layer
model.add(Embedding(vocab_size, embedding_vector_length, input_length=max_review_length))

# Add batch normalization
model.add(BatchNormalization())

# Add dropout
model.add(Dropout(0.20))

# Add LSTM Layer 1
model.add(LSTM(100, return_sequences=True))

# Add dropout
model.add(Dropout(0.20))

# Add LSTM Layer 2
model.add(LSTM(100))

# Add dropout
model.add(Dropout(0.20))

# Add Dense Layer
model.add(Dense(1, activation='sigmoid'))

# Summary of the model
print("Model Summary: \n")
model.summary()
print()
print()

# Compile the model
model.compile(loss='binary_crossentropy', optimizer='adam', metrics=['accuracy'])

# Run the model
trained_model = model.fit(x_train, np.array(y_train), batch_size=batch_size, epochs=epochs, verbose=1, validation_data=(x_test, y_test))
```

## Model Summary:

Layer (type)	Output Shape	Param #
embedding_2 (Embedding)	(None, 500, 32)	864096
batch_normalization_2 (Batch Normalization)	(None, 500, 32)	128
dropout_3 (Dropout)	(None, 500, 32)	0
lstm_2 (LSTM)	(None, 500, 100)	53200
dropout_4 (Dropout)	(None, 500, 100)	0
lstm_3 (LSTM)	(None, 100)	80400
dropout_5 (Dropout)	(None, 100)	0
dense_2 (Dense)	(None, 1)	101
Total params: 997,925		
Trainable params: 997,861		
Non-trainable params: 64		

Train on 35000 samples, validate on 15000 samples

Epoch 1/10

35000/35000 [=====] - 1476s 42ms/step - loss: 0.2645 - acc: 0.9071 - val\_loss: 0.2976 - val\_acc: 0.9019

Epoch 2/10

35000/35000 [=====] - 818s 23ms/step - loss: 0.1631 - acc: 0.9389 - val\_loss: 0.2325 - val\_acc: 0.9189

Epoch 3/10

35000/35000 [=====] - 818s 23ms/step - loss: 0.1304 - acc: 0.9522 - val\_loss: 0.2236 - val\_acc: 0.9256

Epoch 4/10

35000/35000 [=====] - 914s 26ms/step - loss: 0.1092 - acc: 0.9608 - val\_loss: 0.2322 - val\_acc: 0.9224

Epoch 5/10

35000/35000 [=====] - 824s 24ms/step - loss: 0.0868 - acc: 0.9707 - val\_loss: 0.2852 - val\_acc: 0.9205

Epoch 6/10

35000/35000 [=====] - 756s 22ms/step - loss: 0.0707 - acc: 0.9757 - val\_loss: 0.3011 - val\_acc: 0.9205

Epoch 7/10

35000/35000 [=====] - 781s 22ms/step - loss: 0.0582 - acc: 0.9805 - val\_loss: 0.3401 - val\_acc: 0.9164

Epoch 8/10

35000/35000 [=====] - 782s 22ms/step - loss: 0.0494 - acc: 0.9833 - val\_loss: 0.3270 - val\_acc: 0.9146

Epoch 9/10

35000/35000 [=====] - 755s 22ms/step - loss: 0.0412 - acc: 0.9856 - val\_loss: 0.4026 - val\_acc: 0.9179

Epoch 10/10

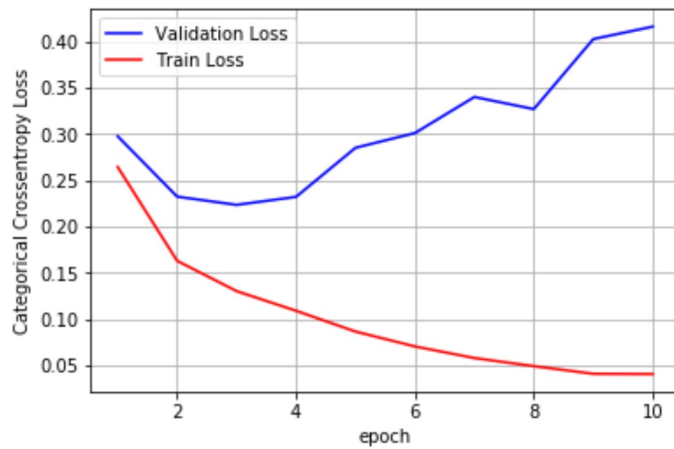
35000/35000 [=====] - 1213s 35ms/step - loss: 0.0409 - acc: 0.9860 - val\_loss: 0.4161 - val\_acc: 0.9126

Wall time: 2h 32min 27s

```
In [20]: score = model.evaluate(x_test, y_test, verbose=0)
print('Test accuracy: {0:.2f}%'.format(score[1]*100))
print()
print()

# Plot train and cross validation error
plot_train_cv_loss(trained_model, epochs)
```

Test accuracy: 91.26%



### 3 LSTM Layer

```
In [21]: model = Sequential()

# Add Embedding Layer
model.add(Embedding(vocab_size, embedding_vector_length, input_length=max_review_length))

# Add batch normalization
model.add(BatchNormalization())

# Add dropout
model.add(Dropout(0.20))

# Add LSTM Layer 1
model.add(LSTM(100, return_sequences=True, bias_regularizer=reg))

# Add dropout
model.add(Dropout(0.20))

# Add LSTM Layer 2
model.add(LSTM(80, return_sequences=True, bias_regularizer=reg))

# Add dropout
model.add(Dropout(0.20))

# Add LSTM Layer 3
model.add(LSTM(60, return_sequences=True, bias_regularizer=reg))

# Add dropout
model.add(Dropout(0.30))

# Add LSTM Layer 4
model.add(LSTM(40, return_sequences=True, bias_regularizer=reg))

# Add batch normalization
model.add(BatchNormalization())

# Add dropout
model.add(Dropout(0.40))

# Add LSTM Layer 5
model.add(LSTM(20))

# Add dropout
model.add(Dropout(0.50))

# Add Dense Layer
model.add(Dense(1, activation='sigmoid'))

# Summary of the model
print("Model Summary: \n")
model.summary()
print()
print()

# Compile the model
model.compile(loss='binary_crossentropy', optimizer='adam', metrics=['accuracy'])

# Run the model
trained_model = model.fit(x_train, np.array(y_train), batch_size = batch_size, epochs = epochs, verbose=1, validation_data=(x_test, y_test))
```

## Model Summary:

Layer (type)	Output Shape	Param #
embedding_3 (Embedding)	(None, 500, 32)	864096
batch_normalization_3 (Batch Normalization)	(None, 500, 32)	128
dropout_6 (Dropout)	(None, 500, 32)	0
lstm_4 (LSTM)	(None, 500, 100)	53200
dropout_7 (Dropout)	(None, 500, 100)	0
lstm_5 (LSTM)	(None, 500, 80)	57920
dropout_8 (Dropout)	(None, 500, 80)	0
lstm_6 (LSTM)	(None, 500, 60)	33840
dropout_9 (Dropout)	(None, 500, 60)	0
lstm_7 (LSTM)	(None, 500, 40)	16160
batch_normalization_4 (Batch Normalization)	(None, 500, 40)	160
dropout_10 (Dropout)	(None, 500, 40)	0
lstm_8 (LSTM)	(None, 20)	4880
dropout_11 (Dropout)	(None, 20)	0
dense_3 (Dense)	(None, 1)	21
Total params: 1,030,405		
Trainable params: 1,030,261		
Non-trainable params: 144		

Train on 35000 samples, validate on 15000 samples

Epoch 1/10

35000/35000 [=====] - 3090s 88ms/step - loss: 5.2278 - acc: 0.8877 - val\_loss: 4.4712 - val\_acc: 0.8883

Epoch 2/10

35000/35000 [=====] - 2180s 62ms/step - loss: 3.8200 - acc: 0.9122 - val\_loss: 3.2830 - val\_acc: 0.9165

Epoch 3/10

35000/35000 [=====] - 1937s 55ms/step - loss: 2.7252 - acc: 0.9354 - val\_loss: 2.3321 - val\_acc: 0.9216

Epoch 4/10

35000/35000 [=====] - 1936s 55ms/step - loss: 1.8562 - acc: 0.9475 - val\_loss: 1.5607 - val\_acc: 0.9215

Epoch 5/10

35000/35000 [=====] - 2827s 81ms/step - loss: 1.1682 - acc: 0.9573 - val\_loss: 1.0287 - val\_acc: 0.9203

Epoch 6/10

35000/35000 [=====] - 2600s 74ms/step - loss: 0.6386 - acc: 0.9641 - val\_loss: 0.5964 - val\_acc: 0.9198

Epoch 7/10

35000/35000 [=====] - 1940s 55ms/step - loss: 0.2390 - acc: 0.9685 - val\_loss: 0.2973 - val\_acc: 0.9138

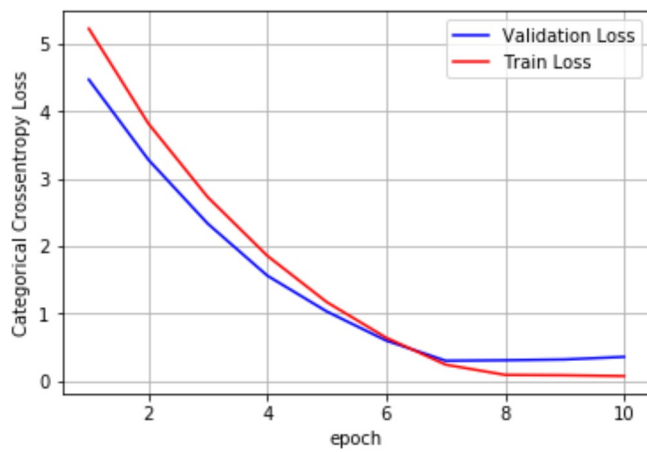
Epoch 8/10

```
In [22]: score = model.evaluate(x_test, y_test, verbose=0)
print('Test accuracy: {0:.2f}%'.format(score[1]*100))
```

Test accuracy: 91.33%

```
In [23]: print()
print()

# Plot train and cross validation error
plot_train_cv_loss(trained_model, epochs)
```



```
In [ ]:
```