```
In [5]:
 from google.colab import drive
drive.mount('/content/gdrive')
Go to this URL in a browser: https://accounts.google.com/o/oauth2/auth?client id=947318989803-6bn6
qk8qdgf4n4g3pfee6491hc0brc4i.apps.googleusercontent.com&redirect uri=urn%3Aietf%3Awg%3Aoauth%3A2.0%
b\&scope=email \$20 https \$3A \$2F \$2F www.googleap is.com \$2F auth \$2F docs.test \$20 https \$3A \$2F \$2F www.googleap is.com \$2F auth \$2F docs.test \$20 https \$3A \$2F \$2F www.googleap is.com \$2F auth \$2F docs.test \$20 https \$3A \$2F \$2F www.googleap is.com \$2F auth \$2F docs.test \$20 https \$3A \$2F \$2F www.googleap is.com \$2F auth \$2F docs.test \$20 https \$3A \$2F \$2F www.googleap is.com \$2F auth \$2F docs.test \$20 https \$3A \$2F \$2F www.googleap is.com \$2F auth \$2F docs.test \$20 https \$3A \$2F \$2F www.googleap is.com \$2F auth \$2F docs.test \$20 https \$3A \$2F \$2F www.googleap is.com \$2F auth \$2F docs.test \$20 https \$3A \$2F \$2F www.googleap is.com \$2F auth \$2F docs.test \$20 https \$3A \$2F \$2F www.googleap is.com \$2F auth \$2F docs.test \$20 https \$3A \$2F \$2F www.googleap is.com \$2F auth \$2F docs.test \$20 https \$3A \$2F \$2F www.googleap is.com \$2F auth \$2F docs.test \$20 https \$3A \$2F \$2F www.googleap is.com \$2F auth \$2F docs.test \$2F auth \$2F 
2Fauth%2Fdrive%20https%3A%2F%2Fwww.googleapis.com%2Fauth%2Fdrive.photos.readonly%20https%3A%2F%2Fww
ogleapis.com%2Fauth%2Fpeopleapi.readonly&response type=code
Enter your authorization code:
Mounted at /content/gdrive
In [0]:
In [16]:
 !ls "/content/gdrive/My Drive/quora/data/train.csv"
1.Quora.ipynb
                                             nlp_feat_linear.csv
2.Quora Preprocessing.ipynb
                                                                                         nlp features train.csv
3.Q Mean W2V.ipynb
                                                                   sparse lr.pckl
4.ML models.ipynb
                                                                   sparse_lr_y.pckl
APPLIED AI QUORASUBMISSION.ipynb
{\tt df\_fe\_without\_preprocessing\_train.csv} \quad {\tt train.db}
distance
                                     train n.txt
Distance-0.1.3.tar.gz train_p.txt wordcloud-1.3.3-cp36-cp36m-win_amd64.whl
In [0]:
 path = "/content/gdrive/My Drive/quora/data/train.csv"
 save path = "/content/gdrive/My Drive/quora/data/"
In [0]:
import pandas as pd
In [0]:
data = pd.read csv(path)
In [127]:
data.head(2)
Out[127]:
                                                                                                              question1
       id qid1
                         qid2
                                                                                                                                                                                                                  question2 is_duplicate
 0 0
             1
                          2
                                     What is the step by step guide to invest in sh...
                                                                                                                                     What is the step by step guide to invest in sh...
                                                                                                                                                                                                                                          0
                                                                                                                                     What would happen if the Indian government
                                      What is the story of Kohinoor (Koh-i-Noor)
```

In [128]:	
data.tail(4)	

0

1 3

4

	id	qid1	qid2	question1 question2		is_duplicate
404286	404286	18840	155606	Do you believe there is life after death?	Is it true that there is life after death?	1
404287	404287	537928	537929	What is one coin?	What's this coin?	0
404288	404288	537930	537931	What is the approx annual cost of living while	I am having little hairfall problem but I want	0
404289	404289	537932	537933	What is like to have sex with cousin?	What is it like to have sex with your cousin?	

```
In [21]:
len(data)
Out[21]:
404290
In [9]:
from keras.preprocessing.text import Tokenizer
{\bf from\ keras.preprocessing.sequence\ import\ pad\_sequences}
Using TensorFlow backend.
In [0]:
tokenizer = Tokenizer(num words=None, filters='!"#$%&()*+,-./:;<=>?@[\]^ `{|}~ ', lower=True, spli
t=' ', char_level=False, oov_token='unk')
In [0]:
#slight aside
In [0]:
texts = data['question1'].astype('str').tolist()
In [0]:
import numpy as np
texts[0]
labels = np.asarray(data['is_duplicate'])
In [0]:
labels = labels.tolist()
In [18]:
labels[0:3]
Out[18]:
[0, 0, 0]
In [0]:
textsq2 = data['question2'].astype('str').tolist()
equalizedq1 = []
equalizedq2 = []
new labels = []
for que1, que2, labels in zip(texts, textsq2, labels):
```

```
if len(que1.split()) <= 30 and len(que2.split()) <= 30:</pre>
    equalizedq1.append(que1)
    equalizedq2.append(que2)
    new labels.append(labels)
In [23]:
len(equalizedq1)
Out[23]:
396450
In [0]:
tokenizer.fit_on_texts(texts)
In [33]:
len(new labels)
Out[33]:
396450
In [0]:
tokenizer.fit on texts(equalizedq1)
sequencesq1 = tokenizer.texts_to_sequences(equalizedq1)
word_indexq1 = tokenizer.word_index
tokenizer.fit_on_texts(equalizedq2)
sequencesq2 = tokenizer.texts to sequences(equalizedq2)
word indexq2 = tokenizer.word index
In [0]:
question2 = pad_sequences(sequencesq2, maxlen=30)
In [0]:
question1 = pad sequences(sequencesq1, maxlen = 30)
In [48]:
len(question2)
Out[48]:
396450
In [50]:
question2.shape
Out [50]:
(396450, 30)
In [0]:
#let's make a simple network for these
X validq1 = question1[:80000,:]
X_validq2 = question2[:80000,:]
question1 = question1[80000:,:]
```

```
questions - questions [00000.,.]
y_valid = new_labels[:80000]
labels = new labels[80000:]
In [52]:
len(y_valid),len(X_validq1),len(question1),len(labels)
Out[52]:
(80000, 80000, 316450, 316450)
In [0]:
n_1stm_1 = 100
drop_1stm = 0.3
drop\_embed = 0.3
vocab1 = len(word indexq1)
vocab2 = len(word_indexq2)
qu1 input = Input(shape = (30,))
#drop1 = SpatialDropout1D(drop_embed)(qu1_input)
#drop1 = SpatialDropout1D(drop embed) (qu1 input)
emb1 = Embedding(vocab1,64)(qu1_input)
qu2 input = Input(shape = (30,))
#drop2 = SpatialDropout1D(drop embed)(qu2 input)
emb2 = Embedding(vocab2,64)(qu2 input)
core = Bidirectional(LSTM(64, dropout=drop_lstm))
# out core = Bidirectional(LSTM(n lstm 1, dropout=drop lstm))
out1 = core(emb1)
# out further 1 = out core(out1)
out2 = core(emb2)
merged = concatenate([out1,out2])
predictions = Dense(1,activation = 'sigmoid') (merged)
model = Model([qu1_input, qu2_input], predictions)
In [56]:
len(question1),len(question2)
Out[56]:
(316450, 316450)
In [55]:
len(labels)
Out [55]:
316450
In [58]:
model.compile(loss='binary crossentropy', optimizer='adam', metrics=['accuracy'])
history = model.fit([question1,question2],labels,epochs = 12,batch_size = 512,validation_data = ([X
_validq1,X_validq2],y_valid))
```

```
Train on 316450 samples, validate on 80000 samples
Epoch 1/12
l loss: 0.4990 - val acc: 0.7634
Epoch 2/12
l loss: 0.5162 - val acc: 0.7644
Epoch 3/12
1_loss: 0.5488 - val_acc: 0.7613
Epoch 4/12
l loss: 0.5815 - val acc: 0.7617
Epoch 5/12
1_loss: 0.6318 - val_acc: 0.7585
Epoch 6/12
1_loss: 0.6909 - val_acc: 0.7589
Epoch 7/12
l loss: 0.7290 - val acc: 0.7574
Epoch 8/12
l loss: 0.7939 - val acc: 0.7591
Epoch 9/12
l loss: 0.8121 - val acc: 0.7557
Epoch 10/12
1_loss: 0.8895 - val_acc: 0.7551
Epoch 11/12
1 loss: 0.9595 - val acc: 0.7523
Epoch 12/12
l loss: 1.0038 - val acc: 0.7550
```

In [0]:

```
sequences = tokenizer.texts_to_sequences(texts)
word_index = tokenizer.word_index
```

In [83]:

```
pd.DataFrame(np.array([len(seq) for seq in sequences]).reshape((len(sequences),1))).describe() # pad for 20
```

Out[83]:

	0
count	404290.000000
mean	11.033793
std	5.499495
min	0.000000
25%	7.000000
50%	10.000000
75%	13.000000
max	127.000000

In [0]:

```
question1 = pad_sequences(sequences, maxlen=20)
labels = np.asarray(data['is_duplicate'])
```

```
In [0]:
```

```
tokenizer2 = Tokenizer(num_words=None, filters='!"#$%&()*+,-./:;<=>?@[\]^_`{|}~', lower=True, spl
it=' ', char_level=False, oov_token='unk')
texts2 = data['question2'].astype('str').tolist()
tokenizer2.fit_on_texts(texts2)

sequences2 = tokenizer.texts_to_sequences(texts2)
pd.DataFrame(np.array([len(seq) for seq in sequences2]).reshape((len(sequences2),1))).describe() #
pad for 20
word_index2 = tokenizer2.word_index
```

In [86]:

```
\verb|pd.DataFrame(np.array([len(seq) | \textbf{for} | seq | \textbf{in} | sequences2]).reshape((len(sequences2), 1))).describe()| \\
```

Out[86]:

	0
count	404290.000000
mean	11.278891
std	6.377210
min	0.000000
25%	7.000000
50%	10.000000
75%	13.000000
max	237.000000

In [0]:

```
question2 = pad_sequences(sequences2, maxlen=20)
```

In [0]:

```
vocab2 = len(word_index2)
```

In [0]:

```
vocab1 = len(word_index)
```

In [0]:

```
from keras import Model

from keras import Input
from keras.datasets import imdb
from keras.preprocessing.sequence import pad_sequences
from keras.models import Sequential
from keras.layers import Dense, Dropout, Embedding, SpatialDropout1D, LSTM,concatenate
from keras.layers.wrappers import Bidirectional
from keras.callbacks import ModelCheckpoint
import os
from sklearn.metrics import roc_auc_score
import matplotlib.pyplot as plt
%matplotlib inline

def plt_dynamic(x, vy, ty, ax, colors=['b']):
    ax.plot(x, vv, 'b', label="Validation Loss")
```

```
ax.plot(x, ty, 'r', label="Train Loss")
plt.legend()
plt.grid()
fig.canvas.draw()
```

In [0]:

```
n lstm 1 = 100
drop 1stm = 0.3
drop = 0.3
qul_input = Input(shape = (20,))
#drop1 = SpatialDropout1D(drop_embed) (qu1_input)
#drop1 = SpatialDropout1D(drop_embed) (qu1_input)
emb1 = Embedding(vocab1,64)(qu1_input)
qu2_input = Input(shape = (20,))
#drop2 = SpatialDropout1D(drop embed) (qu2 input)
emb2 = Embedding(vocab2,64)(qu2_input)
core = Bidirectional(LSTM(64, dropout=drop lstm ))
# out core = Bidirectional(LSTM(n lstm 1, dropout=drop lstm))
out1 = core(emb1)
# out_further_1 = out_core(out1)
out2 = core(emb2)
merged = concatenate([out1,out2])
predictions = Dense(1,activation = 'sigmoid') (merged)
model = Model([qu1 input, qu2 input], predictions)
```

In [119]:

model.summary()

Layer (type)	Output	Shape	Param #	Connected to
input_18 (InputLayer)	(None,	20)	0	
input_19 (InputLayer)	(None,	20)	0	
embedding_15 (Embedding)	(None,	20, 64)	4732736	input_18[0][0]
embedding_16 (Embedding)	(None,	20, 64)	4361792	input_19[0][0]
bidirectional_12 (Bidirectional	(None,	128)	66048	embedding_15[0][0] embedding_16[0][0]
concatenate_2 (Concatenate)	(None,	256)	0	bidirectional_12[0][0] bidirectional_12[1][0]
dense_2 (Dense)	(None,	1)	257	concatenate_2[0][0]
dense_2 (Dense) Total params: 9,160,833 Trainable params: 9,160,833 Non-trainable params: 0	(None,	1)	257	concatenate_2[0][0]

In [0]:

```
from sklearn.model_selection import train_test_split
```

```
len(question1)
Out[129]:
404290
In [1301:
0.3*len(question1)
Out[130]:
121287.0
In [0]:
X validq1 = question1[:100000,:]
X validq2 = question2[:100000,:]
question1 = question1[100000:]
question2 = question2[100000:]
y valid = labels[:100000]
labels = labels[100000:]
In [132]:
labels[:10]
Out[132]:
array([0, 0, 0, 0, 0, 1, 0, 1, 0, 0, 0, 1, 1, 1, 0, 1, 1, 0, 1, 0, 1, 0,
       0, 0, 0, 0, 0, 0, 1, 0, 1, 1, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0,
       0, 0, 0, 0, 1, 1, 1, 1, 0, 1, 0, 0, 0, 0, 1, 0, 0, 0, 1, 0, 0, 1,
       1, 1, 0, 0, 0, 1, 1, 1, 1, 0, 0, 0, 0, 1, 0, 0, 0, 0, 1, 1, 1, 0,
       1, 0, 0, 0, 1, 1, 0, 1, 0, 0, 0, 0, 1, 0, 0, 0, 1, 0, 0, 1, 0, 0,
       0, 0, 0, 1, 0, 0, 0, 0, 0, 1, 0, 1, 0, 1, 0, 1, 0, 0, 0, 0,
       0, 0, 0, 1, 1, 0, 0, 0, 0, 0, 1, 1, 0, 0, 0, 0, 0, 0, 0, 1, 0,
       0, 0, 1, 0, 1, 1, 1, 0, 0, 1, 0, 1, 0, 0, 1, 0, 0, 0, 0, 1, 0, 1,
       1, 0, 1, 1, 1, 0, 1, 0, 0, 1, 0, 0, 1, 1, 1, 1, 0, 1, 1, 0, 0, 1,
       1, 1, 1, 0, 0, 1, 0, 0, 0, 0, 1, 1, 0, 0, 0, 0, 1, 1, 0, 0, 1,
       1, 1, 0, 0, 1, 0, 1, 0, 0, 1, 0, 0, 0, 0, 1, 1, 0, 1, 0, 0, 0,
       1, 1, 1, 0, 1, 0, 0, 1, 1, 1, 0, 1, 0, 1, 0, 0, 0, 0, 1, 1, 1, 0,
       0, 0, 0, 1, 0, 1, 1, 0, 0, 1, 1, 1, 0, 0, 0, 0, 0, 1, 0, 0, 1, 1,
       1, 1, 1, 0, 0, 1, 0, 1, 0, 1, 1, 0, 0, 1, 0, 0, 0, 0, 1, 0, 0,
       1, 1, 0, 0, 1, 0, 0, 0, 0, 1, 1, 0, 0, 1, 1, 1, 0, 0, 1, 0, 0, 1,
       0, 1, 0, 0, 0, 0, 0, 0, 1, 0, 1, 0, 0, 0, 0, 1, 1, 1, 1, 1, 0,
       0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 1, 0, 0, 1, 0, 0, 0, 1,
       0, 0, 0, 1, 0, 0, 1, 0, 0, 1, 0, 0, 0, 0, 0, 0, 1, 0, 0, 1, 1, 1,
         1, 0, 1, 1, 0, 1, 1, 1, 1, 0, 0, 0, 1, 1, 0, 1, 0, 0, 1, 0, 0,
       0.
       0, 0, 0, 1, 1, 0, 0, 0, 0, 0, 1, 0, 1, 1, 1, 0, 0, 0, 0, 0, 0, 1,
       0, 0, 1, 1, 0, 1, 1, 0, 0, 0, 1, 1, 0, 0, 0, 0, 0, 1, 1, 1, 1, 1,
       1, 0, 1, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 1, 0, 0, 1, 0, 0, 1,
       1, 0, 0, 0, 1, 1, 0, 1, 0, 1, 0, 0, 0, 0, 0, 0, 0, 1, 0, 1, 0, 1,
       1, 1, 1, 0, 0, 0, 0, 1, 0, 1, 0, 1, 1, 0, 1, 0, 0, 0, 0, 0, 0, 1,
       0, 0, 0, 0, 1, 1, 1, 1, 1, 0, 1, 0, 1, 0, 0, 0, 1, 0, 0, 1, 0,
       0, 1, 0, 1, 0, 1, 0, 1, 1, 0, 0, 1, 0, 0, 1, 0, 0, 0, 0, 1, 0, 1,
       0, 1, 0, 0, 0, 1, 0, 0, 1, 0, 0, 1, 1, 0, 1, 0, 0, 0, 1, 0, 1, 1,
       1, 1, 0, 0, 0, 0, 1, 0, 1, 0, 0, 0, 0, 1, 1, 1, 0, 0, 0, 0, 0, 0,
       0, 1, 1, 1, 1, 0, 0, 0, 1, 1, 0, 0, 0, 1, 0, 0, 0, 0, 1, 1,
         0, 0, 1, 0, 0, 1, 0, 0, 1, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0,
       1, 1, 0, 1, 1, 1, 0, 0, 0, 0, 0, 0, 0, 0, 1, 1, 0, 1, 0, 1, 0,
       0, 0, 0, 0, 0, 0, 1, 1, 0, 0, 0, 0, 0, 1, 0, 0, 1, 1, 0, 1, 0,
       1, 1, 0, 1, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 1, 1, 0, 0, 0,
       1, 1, 1, 1, 0, 0, 0, 1, 1, 0, 0, 0, 1, 0, 0, 1, 0, 0, 1, 0,
       0, 0, 1, 0, 1, 0, 0, 1, 0, 0, 0, 0, 1, 1, 0, 0, 0, 0, 1, 0, 0, 0,
                                                                  1,
       0, 0, 1, 1, 0, 0, 0, 0, 0, 1, 0,
                                       1, 0, 0, 1,
                                                   1, 0, 1, 0, 0,
       0, 0, 0, 0, 1, 0, 1, 0, 1, 0, 0, 1, 0, 1, 1, 0, 0, 0, 0, 0, 1, 0,
       0, 1, 0, 1, 0, 1, 0, 0, 0, 0, 1, 0, 1, 1, 0, 1, 0, 0, 0, 1, 1,
      1, 1, 0, 0, 1, 0, 0, 0, 0, 0, 0, 1, 1, 1, 0, 1, 0, 0, 1, 1, 0,
       1, 1, 0, 1, 0, 1, 0, 0, 0, 0, 1, 0, 0, 1, 0, 1, 0, 1, 1, 0, 0,
       1, 0, 0, 0, 0, 0, 0, 0, 0, 1, 1, 1, 1, 1, 0, 0, 0, 0, 1, 0, 0, 0,
       0, 1, 1, 1, 1, 1, 1, 0, 0, 0, 0, 0, 1, 1, 1, 1, 1, 1, 0, 0, 1, 1,
       0. 0. 0. 0. 1. 1. 0. 0. 0. 1. 0. 0. 0. 1. 1. 0. 0. 0. 1. 0. 1.
```

```
1, 0, 0, 1, 0, 1, 1, 1, 1, 0, 1, 0, 1, 1, 0, 0, 1, 0, 0, 1, 1, 1, 1, 1, 0, 0, 0, 1, 0, 0, 1, 1, 1, 1, 1, 0, 0, 0, 1, 0, 0, 0, 1, 0, 0, 1, 0, 0, 1, 0, 0, 1, 0, 0, 1, 0, 0, 1, 0, 0, 1, 0, 0, 0, 1, 0, 0, 1, 0, 0, 0, 1, 0])
```

In [0]:

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

In [135]:

l loss: 1.0456 - val acc: 0.7823

```
history = model.fit([question1,question2],labels,epochs = 40,batch size = 512,validation data = ([X
_validq1,X_validq2],y valid))
Train on 304290 samples, validate on 100000 samples
Epoch 1/40
1_loss: 0.3707 - val_acc: 0.8275
Epoch 2/40
1_loss: 0.3948 - val_acc: 0.8191
Epoch 3/40
1 loss: 0.4230 - val acc: 0.8109
Epoch 4/40
l loss: 0.4651 - val acc: 0.8022
Epoch 5/40
l loss: 0.4901 - val acc: 0.8025
Epoch 6/40
l loss: 0.5426 - val acc: 0.7913
Epoch 7/40
l loss: 0.5744 - val acc: 0.7946
Epoch 8/40
l_loss: 0.6148 - val_acc: 0.7939
Epoch 9/40
l loss: 0.6676 - val acc: 0.7908
Epoch 10/40
l_loss: 0.7175 - val_acc: 0.7918
Epoch 11/40
l loss: 0.7534 - val acc: 0.7892
Epoch 12/40
l_loss: 0.8215 - val_acc: 0.7903
Epoch 13/40
1_loss: 0.8431 - val_acc: 0.7856
Epoch 14/40
l loss: 0.8676 - val acc: 0.7916
Epoch 15/40
1 loss: 0.8800 - val acc: 0.7882
Epoch 16/40
l loss: 0.9048 - val acc: 0.7846
Epoch 17/40
1 loss: 0.9639 - val acc: 0.7829
Epoch 18/40
l loss: 0.9897 - val acc: 0.7865
Epoch 19/40
l loss: 1.0217 - val acc: 0.7870
Epoch 20/40
```

```
Epoch 21/40
1_loss: 1.0933 - val_acc: 0.7832
Epoch 22/40
l loss: 1.1222 - val acc: 0.7785
Epoch 23/40
l loss: 1.1008 - val acc: 0.7837
Epoch 24/40
l loss: 1.1354 - val acc: 0.7849
Epoch 25/40
1_loss: 1.1664 - val_acc: 0.7830
Epoch 26/40
l_loss: 1.2111 - val_acc: 0.7743
Epoch 27/40
l loss: 1.2179 - val acc: 0.7805
Epoch 28/40
1_loss: 1.2662 - val_acc: 0.7805
Epoch 29/40
l loss: 1.2722 - val acc: 0.7797
Epoch 30/40
l loss: 1.2531 - val acc: 0.7770
Epoch 31/40
l loss: 1.2847 - val acc: 0.7780
Epoch 32/40
l loss: 1.3264 - val acc: 0.7789
Epoch 33/40
l loss: 1.3271 - val acc: 0.7737
Epoch 34/40
l loss: 1.3599 - val acc: 0.7770
Epoch 35/40
l loss: 1.3407 - val acc: 0.7782
Epoch 36/40
l loss: 1.3818 - val acc: 0.7782
Epoch 37/40
l loss: 1.3910 - val acc: 0.7766
Epoch 38/40
l_loss: 1.4197 - val_acc: 0.7745
Epoch 39/40
1 loss: 1.3854 - val_acc: 0.7783
Epoch 40/40
l loss: 1.4237 - val acc: 0.7780
```

In [136]:

```
#trial 2
n_lstm_1 = 100
drop_lstm = 0.3
drop_embed = 0.3

qu1_input = Input(shape = (20,))
#drop1 = SpatialDropout1D(drop_embed)(qu1_input)
#drop1 = SpatialDropout1D(drop_embed)(qu1_input)
emb1 = Embedding(vocab1,128)(qu1_input)

qu2_input = Input(shape = (20,))

#drop2 = SpatialDropout1D(drop_embed)(qu2_input)
```

```
emb2 = Embedding(vocab2,128)(qu2 input)
core = Bidirectional(LSTM(128, dropout=drop lstm ))
# out core = Bidirectional(LSTM(n lstm 1, dropout=drop lstm))
out1 = core(emb1)
# out_further_1 = out core(out1)
out2 = core(emb2)
merged = concatenate([out1,out2])
predictions = Dense(1,activation = 'sigmoid') (merged)
model big = Model([qu1 input, qu2 input], predictions)
model big.compile(loss='binary crossentropy', optimizer='sgd', metrics=['accuracy'])
history big = model.fit([question1,question2],labels,epochs = 40,batch_size = 512,validation_data =
([X validq1,X validq2],y valid)) #model big
Train on 304290 samples, validate on 100000 samples
Epoch 1/40
l loss: 1.4221 - val acc: 0.7772
Epoch 2/40
l loss: 1.4492 - val acc: 0.7768
Epoch 3/40
l loss: 1.4428 - val acc: 0.7766
Epoch 4/40
l loss: 1.4855 - val acc: 0.7731
Epoch 5/40
l loss: 1.4668 - val acc: 0.7767
Epoch 6/40
l loss: 1.4932 - val acc: 0.7776
Epoch 7/40
l loss: 1.5330 - val acc: 0.7716
Epoch 8/40
l_loss: 1.5476 - val_acc: 0.7718
Epoch 9/40
l_loss: 1.5717 - val_acc: 0.7705
Epoch 10/40
l loss: 1.5436 - val acc: 0.7738
Epoch 11/40
l loss: 1.5686 - val acc: 0.7757
Epoch 12/40
1_loss: 1.6120 - val_acc: 0.7755
Epoch 13/40
l loss: 1.6069 - val acc: 0.7741
Epoch 14/40
l loss: 1.6004 - val acc: 0.7753
Epoch 15/40
l loss: 1.6457 - val acc: 0.7746
Epoch 16/40
l loss: 1.6198 - val acc: 0.7743
Epoch 17/40
204200/204200 [-----
                _____1 1710 E6100/0+00 1000. 0 0277 000. 0 00E/
```

```
l loss: 1.6284 - val acc: 0.7756
Epoch 18/40
l_loss: 1.6127 - val_acc: 0.7747
Epoch 19/40
l loss: 1.6317 - val acc: 0.7753
Epoch 20/40
l loss: 1.6188 - val acc: 0.7758
Epoch 21/40
l loss: 1.6282 - val acc: 0.7748
Epoch 22/40
l loss: 1.6511 - val acc: 0.7732
Epoch 23/40
1_loss: 1.6203 - val_acc: 0.7710
Epoch 24/40
l loss: 1.7050 - val acc: 0.7714
Epoch 25/40
l loss: 1.6916 - val acc: 0.7721
Epoch 26/40
l loss: 1.6946 - val acc: 0.7733
Epoch 27/40
304290/304290 [=============] - 170s 559us/step - loss: 0.0324 - acc: 0.9875 - va
l loss: 1.7037 - val acc: 0.7728
Epoch 28/40
l loss: 1.6984 - val acc: 0.7727
Epoch 29/40
l_loss: 1.7228 - val_acc: 0.7741
Epoch 30/40
l loss: 1.7440 - val acc: 0.7764
Epoch 31/40
l loss: 1.7383 - val acc: 0.7714
Epoch 32/40
l loss: 1.7133 - val acc: 0.7714
Epoch 33/40
l_loss: 1.7160 - val_acc: 0.7748
Epoch 34/40
l_loss: 1.7197 - val_acc: 0.7691
Epoch 35/40
l loss: 1.7455 - val acc: 0.7730
Epoch 36/40
l loss: 1.7378 - val acc: 0.7743
Epoch 37/40
l loss: 1.7633 - val acc: 0.7702
Epoch 38/40
l loss: 1.7821 - val acc: 0.7739
Epoch 39/40
l loss: 1.7631 - val acc: 0.7723
Epoch 40/40
l loss: 1.8035 - val acc: 0.7742
```