12/11/17, 12:25

```
In [1]: from __future__ import print_function

from sklearn.preprocessing import OneHotEncoder
from keras.layers.core import Dense, Activation, Dropout
from keras.preprocessing import sequence
from keras.models import Sequential
from keras.layers import Dense, Embedding
from keras.layers import LSTM, Merge
from keras.datasets import imdb
import pandas as pd
import numpy as np
import os
```

Using TensorFlow backend.

rnn

/Users/DuaaTashkandi/anaconda/lib/python3.6/importlib/_bootstrap.py: 205: RuntimeWarning: compiletime version 3.5 of module 'tensorflow.p ython.framework.fast_tensor_util' does not match runtime version 3.6 return f(*args, **kwds)

```
In [2]: #parameters
maxlen = 30
labels = 2
```

```
In [3]: input = pd.read_csv("data_new/merged.csv",header=None)
    input.columns = ['first', 'last','b_or_n']

# remove encode
    input['first'] = input['first'].str[2:-1]
    input['last'] = input['last'].str[2:-1]

input['firstlen'] = [len(str(i)) for i in input['first']]
    input['lastlen'] = [len(str(i)) for i in input['last']]
    input[ 'lastlen'] = [len(str(i)) for i in input['lastlen'] >= 2)]
```

```
In [4]: firsts = input['first']
    lasts = input['last']
    labels = input['b_or_n']

    vocab = set(' '.join([str(i) for i in firsts]))
    vocab = set(' '.join([str(i) for i in lasts]))

    vocab.add('END')
    len_vocab = len(vocab)
```

```
In [5]: char_index = dict((c, i) for i, c in enumerate(vocab))
```

```
In [6]: #train test split
         msk = np.random.rand(len(input1)) < 0.8</pre>
         train = input1[msk]
         test = input1[~msk]
 In [7]: def set flag(i):
             tmp = np.zeros(39);
             tmp[i] = 1
             return(tmp)
 In [8]: # Truncating and padding training data
         train X = []
         train Y = []
         train Z = []
         trunc train first = [str(i)[0:maxlen] for i in train['first']]
         trunc train last = [str(i)[0:maxlen] for i in train['last']]
         for i in trunc train first:
             tmp = [set flag(char index[j]) for j in str(i)]
              for k in range(0,maxlen - len(str(i))):
                  tmp.append(set flag(char index["END"]))
              train X.append(tmp)
         for i in trunc train last:
             tmp = [set flag(char index[j]) for j in str(i)]
              for k in range(0,maxlen - len(str(i))):
                  tmp.append(set flag(char index["END"]))
             train Y.append(tmp)
         for i in train['b or n']:
              if i == 1:
                 train Z.append([1,0])
             else:
                  train Z.append([0,1])
 In [9]: np.asarray(train_X).shape
 Out[9]: (48013, 30, 39)
In [10]: np.asarray(train Y).shape
Out[10]: (48013, 30, 39)
```

```
In [11]: np.asarray(train Z).shape
Out[11]: (48013, 2)
         #build the model: 2 stacked LSTM
In [12]:
         print('Building LSTM model')
         left branch = Sequential()
         left branch.add(LSTM(512, return sequences=True, input shape=(maxlen,1
         en vocab)))
         right branch = Sequential()
         right_branch.add(LSTM(512, return sequences=True, input shape=(maxlen,
         len vocab)))
         model = Sequential()
         model.add(Merge([left branch, right branch], mode='concat'))
         model.add(Dropout(0.2))
         model.add(LSTM(512, return sequences=False))
         model.add(Dropout(0.2))
         model.add(Dense(2))
         # Softmax activation function
         model.add(Activation('softmax'))
         # Cross-entropy loss, metric is accuracy
         model.compile(loss='categorical_crossentropy', optimizer='adam',metric
         s=['accuracy'])
```

Building LSTM model

/Users/DuaaTashkandi/anaconda/lib/python3.6/site-packages/ipykernel_launcher.py:10: UserWarning: The `Merge` layer is deprecated and wil low removed after 08/2017. Use instead layers from `keras.layers.me rge`, e.g. `add`, `concatenate`, etc.

Remove the CWD from sys.path while we load stuff.

```
In [13]: # Truncating and padding test data
         test X = []
         test_Y = []
         test Z = []
         trunc test first = [str(i)[0:maxlen] for i in test['first']]
         trunc test last = [str(i)[0:maxlen] for i in test['last']]
         for i in trunc test first:
             tmp = [set flag(char index[j]) for j in str(i)]
             for k in range(0,maxlen - len(str(i))):
                tmp.append(set flag(char index["END"]))
             test X.append(tmp)
         for i in trunc test last:
             tmp = [set flag(char index[j]) for j in str(i)]
             for k in range(0,maxlen - len(str(i))):
                tmp.append(set flag(char index["END"]))
             test Y.append(tmp)
         for i in test['b or n']:
             if i == 1:
                test Z.append([1,0])
             else:
                test Z.append([0,1])
In [14]: print(np.asarray(test X).shape)
         print(np.asarray(test Y).shape)
         print(np.asarray(test Z).shape)
         (11941, 30, 39)
         (11941, 30, 39)
         (11941, 2)
In [15]:
        batch size = 1000
         model.fit([np.asarray(train X), np.asarray(train Y)], train Z, batch s
         ize=batch size, nb epoch=50, validation data=([np.asarray(test X), np.
         asarray(test Y)], test Z))
         /Users/DuaaTashkandi/anaconda/lib/python3.6/site-packages/keras/mode
         ls.py:939: UserWarning: The `nb epoch` argument in `fit` has been re
         named `epochs`.
          warnings.warn('The `nb epoch` argument in `fit` '
         Train on 48013 samples, validate on 11941 samples
         Epoch 1/50
         s: 0.6589 - acc: 0.5925 - val loss: 0.5259 - val acc: 0.7562
```

```
Epoch 2/50
s: 0.4934 - acc: 0.7720 - val loss: 0.5144 - val acc: 0.7621
Epoch 3/50
s: 0.4585 - acc: 0.7941 - val loss: 0.5232 - val acc: 0.7586
Epoch 4/50
s: 0.4429 - acc: 0.8017 - val_loss: 0.4991 - val acc: 0.7864
Epoch 5/50
s: 0.4213 - acc: 0.8171 - val loss: 0.4070 - val acc: 0.8232
Epoch 6/50
s: 0.3961 - acc: 0.8288 - val loss: 0.4094 - val acc: 0.8248
Epoch 7/50
s: 0.3776 - acc: 0.8404 - val_loss: 0.5977 - val_acc: 0.7334
Epoch 8/50
s: 0.3828 - acc: 0.8398 - val_loss: 0.4222 - val acc: 0.8276
Epoch 9/50
s: 0.3641 - acc: 0.8478 - val loss: 0.3531 - val acc: 0.8554
Epoch 10/50
s: 0.3330 - acc: 0.8620 - val loss: 0.3713 - val acc: 0.8446
Epoch 11/50
s: 0.3370 - acc: 0.8623 - val loss: 0.3468 - val acc: 0.8591
Epoch 12/50
s: 0.3242 - acc: 0.8682 - val loss: 0.3330 - val acc: 0.8610
Epoch 13/50
s: 0.3162 - acc: 0.8724 - val loss: 0.3369 - val acc: 0.8642
Epoch 14/50
s: 0.3041 - acc: 0.8775 - val loss: 0.3177 - val acc: 0.8733
Epoch 15/50
s: 0.3059 - acc: 0.8768 - val loss: 0.3424 - val acc: 0.8572
Epoch 16/50
s: 0.3067 - acc: 0.8746 - val loss: 0.3177 - val acc: 0.8715
Epoch 17/50
s: 0.2874 - acc: 0.8864 - val loss: 0.3299 - val acc: 0.8621
Epoch 18/50
```

```
s: 0.3203 - acc: 0.8711 - val loss: 0.3100 - val acc: 0.8757
Epoch 19/50
s: 0.2869 - acc: 0.8865 - val loss: 0.3051 - val acc: 0.8822
Epoch 20/50
s: 0.2834 - acc: 0.8887 - val loss: 0.3066 - val acc: 0.8776
s: 0.2846 - acc: 0.8869 - val loss: 0.2994 - val acc: 0.8844
Epoch 22/50
s: 0.2682 - acc: 0.8943 - val loss: 0.2956 - val acc: 0.8861
Epoch 23/50
s: 0.2622 - acc: 0.8983 - val loss: 0.2951 - val acc: 0.8843
Epoch 24/50
s: 0.2549 - acc: 0.9009 - val_loss: 0.2740 - val acc: 0.8946
Epoch 25/50
s: 0.2451 - acc: 0.9050 - val loss: 0.2854 - val acc: 0.8920
Epoch 26/50
s: 0.2544 - acc: 0.9025 - val loss: 0.2812 - val acc: 0.8916
Epoch 27/50
s: 0.2365 - acc: 0.9096 - val loss: 0.2833 - val acc: 0.8916
Epoch 28/50
s: 0.2315 - acc: 0.9114 - val loss: 0.2719 - val_acc: 0.8978
Epoch 29/50
s: 0.2280 - acc: 0.9130 - val loss: 0.2908 - val acc: 0.8881
Epoch 30/50
s: 0.2575 - acc: 0.8997 - val loss: 0.2714 - val acc: 0.8956
Epoch 31/50
s: 0.2325 - acc: 0.9121 - val loss: 0.2734 - val acc: 0.8952
Epoch 32/50
s: 0.2406 - acc: 0.9098 - val loss: 0.2907 - val acc: 0.8893
Epoch 33/50
s: 0.2200 - acc: 0.9174 - val loss: 0.2711 - val acc: 0.8967
Epoch 34/50
s: 0.2174 - acc: 0.9175 - val loss: 0.2534 - val acc: 0.9032
Epoch 35/50
```

```
s: 0.2082 - acc: 0.9226 - val loss: 0.2710 - val acc: 0.9029
Epoch 36/50
s: 0.2007 - acc: 0.9251 - val loss: 0.2933 - val acc: 0.8962
Epoch 37/50
s: 0.2033 - acc: 0.9256 - val loss: 0.2641 - val acc: 0.9003
Epoch 38/50
s: 0.1980 - acc: 0.9271 - val loss: 0.2706 - val acc: 0.9018
Epoch 39/50
s: 0.1824 - acc: 0.9331 - val loss: 0.2745 - val acc: 0.8991
Epoch 40/50
s: 0.1763 - acc: 0.9358 - val loss: 0.2661 - val acc: 0.9056
Epoch 41/50
s: 0.1752 - acc: 0.9357 - val loss: 0.2678 - val acc: 0.9034
Epoch 42/50
s: 0.1765 - acc: 0.9355 - val loss: 0.2645 - val acc: 0.9046
Epoch 43/50
s: 0.1676 - acc: 0.9381 - val loss: 0.2742 - val acc: 0.9008
Epoch 44/50
s: 0.1660 - acc: 0.9396 - val loss: 0.2794 - val acc: 0.8977
Epoch 45/50
s: 0.1781 - acc: 0.9335 - val loss: 0.3036 - val acc: 0.8916
Epoch 46/50
s: 0.2114 - acc: 0.9201 - val_loss: 0.3214 - val acc: 0.8620
Epoch 47/50
s: 0.1970 - acc: 0.9245 - val loss: 0.2628 - val acc: 0.9011
Epoch 48/50
s: 0.1753 - acc: 0.9349 - val loss: 0.2637 - val acc: 0.9067
Epoch 49/50
s: 0.1546 - acc: 0.9436 - val loss: 0.2890 - val acc: 0.9005
Epoch 50/50
s: 0.1543 - acc: 0.9425 - val loss: 0.2924 - val acc: 0.9005
```

Out[15]: <keras.callbacks.History at 0x182ea6588>

```
In [17]:
        score, acc = model.evaluate([np.asarray(test X), np.asarray(test Y)],
         test Z)
         print('Test score:', score)
         print('Test accuracy:', acc)
         Test score: 0.292373455935
         Test accuracy: 0.900510844993
In [18]: #save our model and data
         model.save weights('model', overwrite=True)
         train.to_csv("train_split.csv")
         test.to_csv("test_split.csv")
In [19]: evals = model.predict([np.asarray(test_X), np.asarray(test_Y)])
         prob m = [i[0] \text{ for } i \text{ in } evals]
        out = pd.DataFrame(prob m)
In [20]:
         out['first'] = test['first'].reset index()['first']
         out['last'] = test['last'].reset index()['last']
         out['b or n'] = test['b or n'].reset index()['b or n']
In [21]: out.head(10)
         out.columns = ['prob b','first', 'last', 'actual']
         out.head(10)
         out.to_csv("pred_out.csv")
```

```
In [45]: # small test. first 3 names are spanish and last 3 names are Brazilian
         first = ["manuel","luis","ezequil","michel","luiz","alvaro"]
         last = ["garcia", "sanchez", "gomez", "caetano", "guilherme", "lime"]
         X = []
         Y = []
         trunc first = [i[0:maxlen] for i in first]
         trunc last = [i[0:maxlen] for i in last]
         for i in trunc first:
             tmp = [set flag(char index[j]) for j in str(i)]
             for k in range(0,maxlen - len(str(i))):
                 tmp.append(set_flag(char_index["END"]))
             X.append(tmp)
         for i in trunc last:
             tmp = [set flag(char index[j]) for j in str(i)]
             for k in range(0,maxlen - len(str(i))):
                 tmp.append(set flag(char index["END"]))
             Y.append(tmp)
         pred = model.predict([np.asarray(X), np.asarray(Y)])
         pred
```