In []:

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from future import print function
from keras.preprocessing import sequence
from keras.models import Sequential
from keras.layers import Dense, Embedding
from keras.layers import LSTM
from keras.datasets import imdb
max features = 20000
# обрезание текстов после данного количества слов (среди top max features наибол
ее используемые слова)
maxlen = 80
batch size = 128 # увеличьте значение для ускорения обучения
print('Загрузка данных...')
(x_train, y_train), (x_test, y_test) = imdb.load_data(num_words=max_features)
print(len(x_train), 'тренировочные последовательности')
print(len(x test), 'тестовые последовательности')
print('Pad последовательности (примеров в x единицу времени)')
x train = sequence.pad sequences(x train, maxlen=maxlen)
x test = sequence.pad sequences(x test, maxlen=maxlen)
print('x_train shape:', x_train.shape)
print('x test shape:', x test.shape)
print('Построение модели...')
model = Sequential()
model.add(Embedding(max features, 128))
model.add(LSTM(256, dropout=0.2, recurrent dropout=0.2))
model.add(Dense(1, activation='sigmoid'))
# стоит попробовать использовать другие оптимайзер и другие конфигурации оптимай
model.compile(loss='binary crossentropy',
              optimizer='adam',
              metrics=['accuracy'])
print('Процесс обучения...')
model.fit(x train, y train,
          batch_size=batch size,
          epochs=50, # увеличьте при необходимости
          validation data=(x test, y test))
score, acc = model.evaluate(x_test, y_test,
                            batch size=batch size)
print('Результат при тестировании:', score)
print('Тестовая точность:', acc)
```

Using TensorFlow backend.

/home/roman/anaconda3/lib/python3.7/site-packages/tensorflow_core/py thon/framework/indexed_slices.py:433: UserWarning: Converting sparse IndexedSlices to a dense Tensor of unknown shape. This may consume a large amount of memory.

"Converting sparse IndexedSlices to a dense Tensor of unknown shap e. "

```
Train on 25000 samples, validate on 25000 samples
Epoch 1/50
25000/25000 [============== ] - 260s 10ms/step - los
s: 0.4649 - accuracy: 0.7786 - val loss: 0.3627 - val accuracy: 0.84
34
Epoch 2/50
25000/25000 [============== ] - 266s 11ms/step - los
s: 0.3030 - accuracy: 0.8782 - val loss: 0.3941 - val accuracy: 0.81
88
Epoch 3/50
s: 0.2341 - accuracy: 0.9092 - val loss: 0.3932 - val accuracy: 0.82
Epoch 4/50
s: 0.1860 - accuracy: 0.9292 - val loss: 0.5500 - val accuracy: 0.81
81
Epoch 5/50
s: 0.1455 - accuracy: 0.9477 - val_loss: 0.6443 - val_accuracy: 0.81
06
Epoch 6/50
25000/25000 [============= ] - 265s 11ms/step - los
s: 0.1100 - accuracy: 0.9602 - val loss: 0.6045 - val accuracy: 0.81
47
Epoch 7/50
s: 0.0864 - accuracy: 0.9701 - val loss: 0.7593 - val accuracy: 0.81
17
Epoch 8/50
s: 0.0712 - accuracy: 0.9757 - val loss: 0.7251 - val accuracy: 0.81
17
Epoch 9/50
s: 0.0585 - accuracy: 0.9797 - val loss: 0.7361 - val accuracy: 0.81
25
Epoch 10/50
s: 0.0551 - accuracy: 0.9810 - val_loss: 0.8365 - val_accuracy: 0.81
Epoch 11/50
s: 0.0454 - accuracy: 0.9851 - val loss: 0.8115 - val accuracy: 0.80
19
Epoch 12/50
s: 0.0382 - accuracy: 0.9869 - val loss: 0.9943 - val accuracy: 0.80
42
Epoch 13/50
s: 0.0327 - accuracy: 0.9896 - val_loss: 0.9356 - val_accuracy: 0.79
90
Epoch 14/50
s: 0.0295 - accuracy: 0.9905 - val_loss: 1.0331 - val_accuracy: 0.80
Epoch 15/50
s: 0.0284 - accuracy: 0.9904 - val loss: 1.0826 - val accuracy: 0.79
87
```

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Epoch 16/50
s: 0.0249 - accuracy: 0.9922 - val loss: 0.9895 - val accuracy: 0.80
24
Epoch 17/50
s: 0.0207 - accuracy: 0.9939 - val_loss: 1.0285 - val_accuracy: 0.79
Epoch 18/50
25000/25000 [============= ] - 269s 11ms/step - los
s: 0.0166 - accuracy: 0.9947 - val_loss: 1.1477 - val_accuracy: 0.80
18
Epoch 19/50
s: 0.0133 - accuracy: 0.9960 - val loss: 1.1724 - val accuracy: 0.80
50
Epoch 20/50
s: 0.0150 - accuracy: 0.9950 - val loss: 1.0411 - val accuracy: 0.80
69
Epoch 21/50
20736/25000 [=============>.....] - ETA: 35s - loss: 0.01
40 - accuracy: 0.9962
```

In []: