

Text Generation using Long-Short Term Memory (LSTM) Model

1. Prepare & Explore Dataset

```
In [2]: from tensorflow import keras
from keras.callbacks import LambdaCallback
from keras.models import Sequential
from keras.layers import Dense
from keras.layers import LSTM
from keras.optimizers import RMSprop
from keras.utils.data_utils import get_file
import numpy as np
import random
import sys
import io
```

```
In [3]: #read the corpus dataset
path = get_file(
    'nietzsche.txt',
    origin='https://s3.amazonaws.com/text-datasets/nietzsche.txt')
with io.open(path, encoding='utf-8') as f:
    text = f.read().lower()
print('corpus length:', len(text))

Downloading data from https://s3.amazonaws.com/text-datasets/nietzsche.txt (https://s3.amazonaws.com/text-datasets/nietzsche.txt)
606208/600901 [=====] - 1s 1us/step
corpus length: 600893
```

```
In [4]: #prepare charachter-based model
chars = sorted(list(set(text)))
print('total chars:', len(chars))
char_indices = dict((c, i) for i, c in enumerate(chars))
indices_char = dict((i, c) for i, c in enumerate(chars))

total chars: 57
```

```
In [5]: # cut the text in semi-redundant sequences of maxlen characters
maxlen = 40
step = 3
sentences = []
next_chars = []
for i in range(0, len(text) - maxlen, step):
    sentences.append(text[i: i + maxlen])
    next_chars.append(text[i + maxlen])
print('nb sequences:', len(sentences))

print('Vectorization...')
x = np.zeros((len(sentences), maxlen, len(chars)), dtype=np.bool)
y = np.zeros((len(sentences), len(chars)), dtype=np.bool)
for i, sentence in enumerate(sentences):
    for t, char in enumerate(sentence):
        x[i, t, char_indices[char]] = 1
    y[i, char_indices[next_chars[i]]] = 1

nb sequences: 200285
Vectorization...
```

2. Define the neural network architecture

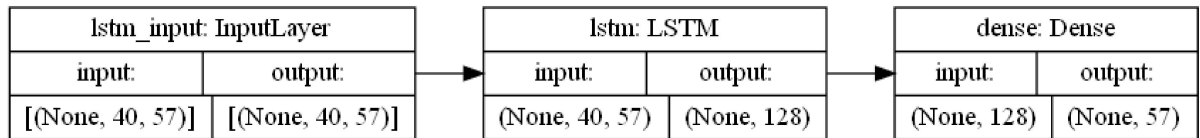
```
In [6]: # build the model: using single LSTM
model = Sequential()
model.add(LSTM(128, input_shape=(maxlen, len(chars))))
model.add(Dense(len(chars), activation='softmax'))
```

3. Compile the neural net

```
In [7]: optimizer = RMSprop()
model.compile(loss='categorical_crossentropy', optimizer=optimizer)
```

In [8]: `keras.utils.plot_model(model, show_shapes=True, rankdir="LR")`

Out[8]:



4. Fit / train the neural net

```

In [9]: def sample(preds, temperature=1.0):
        # helper function to sample an index from a probability array
        preds = np.asarray(preds).astype('float64')
        preds = np.log(preds) / temperature
        exp_preds = np.exp(preds)
        preds = exp_preds / np.sum(exp_preds)
        probas = np.random.multinomial(1, preds, 1)
        return np.argmax(probas)
  
```

```

In [10]: def on_epoch_end(epoch, _):
        # Function invoked at end of each epoch. Prints generated text.
        print()
        print('----- Generating text after Epoch: %d' % epoch)

        start_index = random.randint(0, len(text) - maxlen - 1)
        for diversity in [0.2, 0.5, 1.0, 1.2]:
            print('----- diversity:', diversity)
            generated = ''
            sentence = text[start_index: start_index + maxlen]
            generated += sentence
            print('----- Generating with seed: "' + sentence + '"')
            sys.stdout.write(generated)

            for i in range(400):
                x_pred = np.zeros((1, maxlen, len(chars)))
                for t, char in enumerate(sentence):
                    x_pred[0, t, char_indices[char]] = 1.

                preds = model.predict(x_pred, verbose=0)[0]
                next_index = sample(preds, diversity)
                next_char = indices_char[next_index]

                sentence = sentence[1:] + next_char

                sys.stdout.write(next_char)
                sys.stdout.flush()
            print()
  
```

```

In [11]: print_callback = LambdaCallback(on_epoch_end=on_epoch_end)

        model.fit(x, y,
                  batch_size=128,
                  epochs=12,
                  callbacks=[print_callback])
  
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