dlnd_tv_script_generation

April 2, 2022

1 TV Script Generation

In this project, you'll generate your own Seinfeld TV scripts using RNNs. You'll be using part of the Seinfeld dataset of scripts from 9 seasons. The Neural Network you'll build will generate a new ,"fake" TV script, based on patterns it recognizes in this training data.

1.1 Get the Data

The data is already provided for you in ./data/Seinfeld_Scripts.txt and you're encouraged to open that file and look at the text. >* As a first step, we'll load in this data and look at some samples. * Then, you'll be tasked with defining and training an RNN to generate a new script!

1.2 Explore the Data

Play around with view_line_range to view different parts of the data. This will give you a sense of the data you'll be working with. You can see, for example, that it is all lowercase text, and each new line of dialogue is separated by a newline character \n.

```
In [2]: view_line_range = (0, 10)

"""

DON'T MODIFY ANYTHING IN THIS CELL THAT IS BELOW THIS LINE

"""

import numpy as np

print('Dataset Stats')

print('Roughly the number of unique words: {}'.format(len({word: None for word in text lines = text.split('\n')
```

```
print('Number of lines: {}'.format(len(lines)))
    word_count_line = [len(line.split()) for line in lines]
    print('Average number of words in each line: {}'.format(np.average(word_count_line)))

print('The lines {} to {}:'.format(*view_line_range))
    print('\n'.join(text.split('\n')[view_line_range[0]:view_line_range[1]]))

Dataset Stats
Roughly the number of unique words: 46367
Number of lines: 109233
Average number of words in each line: 5.544240293684143
The lines 0 to 10:
jerry: do you know what this is all about? do you know, why were here? to be out, this is out.
jerry: (pointing at georges shirt) see, to me, that button is in the worst possible spot. the seeres: are you through?
jerry: you do of course try on, when you buy?
george: yes, it was purple, i liked it, i dont actually recall considering the buttons.
```

1.3 Implement Pre-processing Functions

The first thing to do to any dataset is pre-processing. Implement the following pre-processing functions below: - Lookup Table - Tokenize Punctuation

1.3.1 Lookup Table

To create a word embedding, you first need to transform the words to ids. In this function, create two dictionaries: - Dictionary to go from the words to an id, we'll call vocab_to_int - Dictionary to go from the id to word, we'll call int_to_vocab

Return these dictionaries in the following tuple (vocab_to_int, int_to_vocab)

```
In [3]: import problem_unittests as tests

def create_lookup_tables(text):
    """
    Create lookup tables for vocabulary
    :param text: The text of tv scripts split into words
    :return: A tuple of dicts (vocab_to_int, int_to_vocab)
    """
    # TODO: Implement Function
    from collections import Counter
    counts = Counter(text)
```

```
vocab = sorted(counts,key=counts.get,reverse=True)
            vocab_to_int = {word:ii for ii, word in enumerate(vocab,1)}
            int_to_vocab = {ii:word for ii, word in enumerate(vocab,1)}
            # return tuple
            return (vocab_to_int, int_to_vocab)
        11 11 11
        DON'T MODIFY ANYTHING IN THIS CELL THAT IS BELOW THIS LINE
        tests.test_create_lookup_tables(create_lookup_tables)
Tests Passed
In [4]: vocab_to_int, int_to_vocab=create_lookup_tables(text)
In [5]: vocab_to_int
Out[5]: {' ': 1,
         'e': 2,
         't': 3,
         'o': 4,
         'a': 5,
         'r': 6,
         'i': 7,
         'n': 8,
         'h': 9,
         's': 10,
         '\n': 11,
         '1': 12,
         'y': 13,
         'g': 14,
         'u': 15,
         '.': 16,
         'd': 17,
         'm': 18,
         'w': 19,
         ':': 20,
         'c': 21,
         'k': 22,
         ',': 23,
         'p': 24,
         'f': 25,
         "'": 26,
         'b': 27,
         'j': 28,
         '?': 29,
```

```
'v': 30,
          '!': 31,
         '(': 32,
         ')': 33,
         '-': 34,
         'x': 35,
         '"': 36,
          'z': 37,
         'q': 38,
         '*': 39,
         '[': 40,
         ']': 41,
         '0': 42,
         '1': 43,
         '2': 44,
         '5': 45,
         '3': 46,
         ';': 47,
         '4': 48,
         '9': 49,
         '8': 50,
          '#': 51,
         '7': 52,
         '&': 53,
         '6': 54,
         '/': 55,
         '$': 56,
         '`': 57,
         '%': 58,
         '<': 59,
         '>': 60,
         '_': 61,
         '\x92': 62,
         '=': 63,
         '}': 64,
          '+': 65,
         '{': 66,
         '\x93': 67,
         '\x94': 68,
         '': 69,
         '\\': 70,
         '~': 71,
          '\x91': 72}
In [6]: int_to_vocab
Out[6]: {1: '',
         2: 'e',
```

- 3: 't',
- 4: 'o',
- 5: 'a',
- 6: 'r',
- 7: 'i',
- 8: 'n',
- 9: 'h',
- 10: 's',
- 11: '\n',
- 12: '1',
- 13: 'y',
- 14: 'g',
- 15: 'u',
- 16: '.',
- 17: 'd',
- 18: 'm',
- 19: 'w',
- 20: ':',
- 21: 'c',
- 22: 'k',
- 23: ',',
- 24: 'p',
- 25: 'f',
- 26: "'",
- 27: 'b',
- 28: 'j',
- 29: '?',
- 30: 'v',
- 31: '!', 32: '(',
- 33: ')',
- 34: '-',
- 35: 'x',
- 36: '"',
- 37: 'z',
- 38: 'q',
- 39: '*',
- 40: '[',
- 41: ']',
- 42: '0',
- 43: '1',
- 44: '2',
- 45: '5',
- 46: '3',
- 47: ';',
- 48: '4',
- 49: '9',
- 50: '8',