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1 Part 1

1. The reason that the sentences are so long is because of the rule:

1 NP NP PP.

That actually translate with the rule of:

1 PP Prep NP.

To:

1 NP NP Prep NP.

Which meaning that NP is called twice. Since every sentence go thorough NP rule and only 2 of those exists. The sentences becoming long.

- 2. Every sentence go thorough NP rule and NP rule has 50% to go to the "finishing" rule: $1 \ NP \ Det \ Noun$. From this "finishing" rule, there is 6 rules of the kind: $1 \ Noun \ SOMETHING$, which only one of them have "Adj" in it. Therefore 16.67% to be chosen every time. The probability for For n "Adj" in a row is: $0.16^{n-1} \times 0.84$.
- 3. The changes:
 - (a) 9 NP Det Noun. So there is only 10% to duplicate the NP.
 - (b) 5 Noun Adj Noun. For 50% to add "Adj".
- 4. Change the weights of the adjectives and verbs to probabilistic create more realistic English.

The nouns are: president, sandwich, pickle, chief of staff, floor. In total 5 words.

- (a) The weight changes for the adjectives:
 - i. **fine** Can be in proper context with all the nouns. The weight is 1.
 - ii. **delicious** Can be in proper context on "sandwich" and "pickle" out of 5 noun words. The weight is 0.4.

- iii. **perplexed** Can be in proper context on "president" and "chief of staff" out of 5 noun words. The weight is 0.4.
- iv. **pickled** Can be in proper context on "pickle" out of 5 noun words. The weight is 0.2.

I normalize the four adjectives weights to 1:

- i. **fine** 0.5.
- ii. delicious 0.2.
- iii. **perplexed** 0.2.
- iv. **pickled** 0.1.
- (b) The verb come between two "NP" $(S \longrightarrow NP \ Verb \ NP)$, so the weights are actually capture the conditional probability. Given a noun, what the probability to choose a verb that correctly applied on the second noun. The weight changes for the verbs:
 - i. **ate** The verb can be apply in proper context on "sandwich" and "pickle" out of 5 noun words. The weight is 0.4.
 - ii. **wanted** The verb can be apply in proper context with all the nouns. The weight is 1.
 - iii. **kissed** The verb can be apply in proper context with all the nouns. The weight is 1.
 - iv. **understood** The verb can be apply in proper context on "president" and "chief of staff" out of 5 noun words. The weight is 0.4.
 - v. **pickled** The verb can be apply in proper context on "pickle" out of 5 noun words. The weight is 0.2.

Because the rule: "Noun Adj Noun" has weight of 5 (probability of 50%). I normalize the five verbs weights to 5:

- i. **ate** 0.66.
- ii. wanted 1.66.
- iii. kissed 1.66.
- iv. understood 0.66.
- v. **pickled** 0.33.

2 Part 2

All changes with explanations in the file: "grammer2". Before explain why (b) and (h)/(i) can interact in a bad way. First let's see an example: "Sally and the president wanted and is lazy." The problem is: $CC\ Verb$ (In the example translate to: and is), which not suitable verb.

3 Part 3

Tree structures in file "part3.gen".

4 Part 4

1. I chosen (a) "a" vs. "an". To logic was to split every rule with "det" into 3 rules. (1) rule that have all the "det" except "an" and 'a". (2) rule that have "detA" which is "a" that can become Nouns or adjectives that starts without vowel ("detA"/"adjA"). (3) rule that have "detAn" which is "an" that can become Nouns or adjectives that starts with vowel ("detAn"/"adjAn").

Examples:

- (a) every pickle wanted an actual desk.
- (b) a pickle is eating every sandwich.
- (c) is it true that a sandwich perplexed eating a sandwich?
- (d) an apple ate under a floor!
- 2. I chosen (b) Yes-no questions. To logic was to come up with multiple questions format in English and apply it to a rule. The rules added are: (1)ROOT Q?. (2)Q question NNP VP.(3)Q question NP VP. (4)question did.(5)question was.(6)question will

Examples:

- (a) was every president pickled an apple?
- (b) will an eel understood a proposal?
- (c) was a floor perplexed Sally?
- (d) did the umbrella understood eating an oven?

5 Part 5

Since in previous exercise, a POS dictionary already have been built on an input training file. I am that dictionary to built the vocabulary and grammar rules. Generally the dictionary record the POS for a sentence with a sliding window of size 3. The sliding window create rules of a structure: $POS_1 \longrightarrow POS_2 \ POS_3$. Where POS_1 is the POS of the most left word in the sliding window, POS_2 is the POS of the middle word in the sliding window and POS_3 is the POS of the most right word in the sliding window. In Addition structures like $POS \longrightarrow Word$ are saved.

Those structures are saved while traversing the train file, as well as their frequencies. Those rules are added to the initial rules to built a POS from a sentence.

For more details on creating the grammar or the dictionary, please look at the appendix.

Examples:

1. the trading correct the gain!

- 2. is it true that An department encourage the trade?
- 3. these consultant lead Both profitability .

6 Appendix

Running the code:

```
python create_grammar.py --input INPUT --output OUTPUT Code 1: Creating the grammar.
```

INPUT - Is the train POS file from previous task. OUTPUT - Is the location of the ouptput file.

sentences from grammar:

```
python generate.py OUTPUT
```

Code 2: Creating sentences from the grammar.

Dictionary implementation:

While passing over the sentence, word increase the dictionary counts by +1 in four nodes:

- Word POS counts. Number 1 in Fig. 1.
- Right POS counts for that word. Number 2 in Fig. 1.
- Left POS and Right POS counts for that word. Number 3 in Fig. 1.
- Left POS counts for that word. Number 4 in Fig. 1.

To get better understanding of the dictionary structure see Fig. 2. To how does it look in the code for the word "intends".

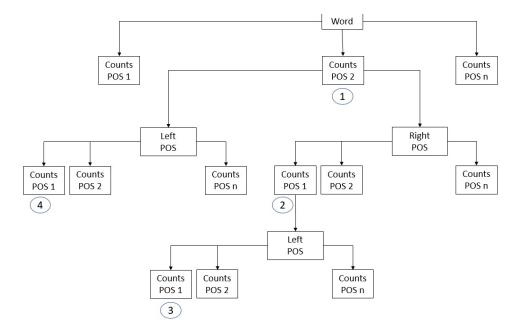


Figure 1: Illustration of the dictionary that saves the counts. Each circled number correspond to different scenario. 1 - POS of the word. 2 - POS of the word and right POS. 3 - left POS and Right POS counts for that word. 4 - POS of the word and left POS.

```
{
    "VBZ": {
        "left_pos": {
            "left_count": 1
        },
        "PRP": {
            "left_count": 1
        },
        "right_pos": {
            "left_pos": {},
            "right_count": 1
        },
        "TO": {
            "left_pos": {
                "left_count": 1
        },
        "right_count": 1
        }
        },
        "right_count": 43
        },
        "right_count": 1
        }
    },
        "right_count": 1
    }
    },
    "wp_count": 45
}
```

Figure 2: Example of the dictionary, as it look in the code, for the word "intends".