

CS 2051: Honors Discrete Mathematics

Spring 2023 Homework 9 Supplement

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The control of a large force is the same principle as the control of a few men: it is merely a question of dividing up their numbers.

Sun Tzu, *The Art of War*

Overview

One day when I was a student taking 2051, I was perusing an [article](#) with all the important results in discrete math. However, I noticed an interesting statement:

Noam Chomsky invents context-free grammars in his book “Syntactic Structures”. At the time, he called them phrase-structure grammars and explored them in the context of modeling the syntax of natural languages.

Initially, I was confused. Noam Chomsky may be many things, including a world-renowned linguist and psychologist, but it was difficult to see his correlation to discrete math. However, as I learned more, I realized the importance of that statement. His discovery took an algorithmic approach to language development whose effects are still being seen today.

In this supplement, you’ll learn about recursion, computational models, syntactical structures, and more!

Part 1: Way Down We Go (9 points)

Let’s motivate this section by recalling some explication from a prior supplement:

But, how can you evaluate arbitrary expressions without even knowing the names of the variables ahead of time? This question itself could be an entire supplemental, but luckily Python makes the job easy with the `eval()` function.

```
>>> eval('p and ((not p) or q)', {'p': True, 'q': False})
```

However, suppose we were now creating our own `eval()` function. This problem is not a trivial one, and requires careful consideration of various algorithmic techniques. Initially, one may assume that an iterative

approach would suffice to traverse the string representation of the proposition, evaluating sub-expressions and gradually building up to the final evaluation. However, such an approach would quickly lead to convoluted code that would be difficult to maintain and debug. Therefore, we must turn to a more elegant solution: recursion.

Recursion is a powerful tool used for solving algorithms, and is loosely given as follows:

- If the problem is small enough to solve directly, then do so.
- Otherwise, reduce it to one or more simpler instances of the same problem.

There are many advantages to this approach, the main one lying in the fact that the same code can be used to solve both the original problem and its sub-problems, leading to simpler, more modular code.

To apply recursion to the task of creating our `eval()` function, we can break the problem down into two parts. First, we must parse the string representation of the proposition into a data structure that can be easily evaluated. Second, we must recursively evaluate this data structure to arrive at the final evaluation of the proposition.

1. We first recurse on the string itself. The idea is to first read the first token in the string, where a token is a basic “word” of our language: either one of the single-letter tokens 'T', 'F', '(', ')', '~', '&', '|', or the two/three-letter “implies” and “iff” tokens '<->', '->', or a variable name like 'p' or 'q76'. This first token will tell you in a unique way how to continue reading the rest of the string, where this reading can be done recursively. For example, if the first token is an open parenthesis, '(', then we know that a formula ϕ must follow, which can be read by a recursive call. Once ϕ was recursively read, we know that the following token must be one of '&', '|', or '->', and once this token is read then a formula ψ must follow, and then a closing parenthesis, ')'. We can use this technique to encode our proposition into a data structure known as a tree, where we start with a root, and then branch off into different subtrees, each with their own roots. This will become concrete as you implement the below tasks.
2. Once we have constructed the tree data structure, we can move on to the second part of the problem: evaluating the proposition. This involves recursively evaluating the subtrees of the tree data structure, starting from the leaves and working our way up to the root. Each subtree represents a sub-expression of the original proposition, and its evaluation can be determined based on the logical structure of the subtree. Once we have evaluated all the subtrees, we can recombine them in such a way that the entire proposition is evaluated.

Note that there are two different types of recursion involved in this process. The first type of recursion involves parsing the string representation of the proposition and constructing the tree data structure, while the second type of recursion involves recursively evaluating the subtrees of the tree data structure.

These two types of recursion correspond to two different types of data structures: *lists* (in the form of strings) and *trees* (in the form of our constructed data structure). This will be explored in future coursework such as CS 351X.

In the very first coding supplement, you were tasked with evaluating a propositional formula given a supposed model. Fortunately, the built-in function `eval()` completed this task for us. *Unfortunately*, now you must do so without the use of that function.

In this part, you'll implement the following functions:

- **parse:** The function takes in a string, represented as a propositional formula, and parses it to create an appropriate Formula object. You may assume that the input string is valid.
- **evaluate:** this function takes in a Formula object, equipped with a model (remember that term from hw2-suppl!?) and returns the evaluation.

Note: All functions must be implemented recursively, or call other recursive functions.

Part 2: Models of Computation (10 points)

In this section, we introduce an abstract representation for computational machines. *Models of computation* are abstract representations of how computation can be performed. They are commonly used to understand the limits and capabilities of computers and other computational devices. A computational model takes in a string as input, and returns a boolean value that represents whether or not to *accept* that string.

There are several models of computation, such as Turing machines, lambda calculus, and finite automata. In this section we present one such model, known as a *context-free-grammar*. We first introduce the concept through an example, and then formalize the definition.

Example 1: Binary strings

Let's say we were trying to create a computational model to *decide* the language $L = \{1^n 0^n, n \in \mathbb{N}\}$. What this means is that our computational model should take in a string, and *accept* the string (i.e. return `True`) if the string is in L , and *reject* otherwise.

One such model could be represented as follows: we would create a . We can write this down as some sort of "rule":

$$S \rightarrow 1S0 \mid \epsilon$$

Let's informally illustrate why this model decides the above language. Suppose we were trying to generate the string 111000. To do this, we can tell our machine to execute the following set of steps:

$$S \rightarrow 1S0 \rightarrow 11S00 \rightarrow 111S000 \rightarrow 111000.$$

As you can see, anytime we move one step to the right, we increase both the number of ones to the left and the number of zeros to the right by 1.

Context-Free Grammars

We now formalize the above concept. A **context-free grammar** is a 4-tuple (V, Σ, R, S) . We now explain each of these elements in detail.

1. V is a finite set called the **variables**. The elements in V are commonly represented using a capital letter (i.e. A, B, \dots). In the above example, the only variable was S .
2. Σ is a finite set, disjoint from V , called the **terminals**. These can be any symbol; in Example 1, they were the binary numbers 0, 1.

3. R is a finite set of **rules**. The left hand side of a rule is a single variable, and the right hand side can be any combination of variables and terminals.
4. $S \in V$ is the start variable.

Example 2: Union of Two Languages

Let's say we were trying to create a CFG for the language

$$L = \underbrace{\{0^n 1^n \mid n \geq 0\}}_{L_1} \cup \underbrace{\{1^n 0^n \mid n \geq 0\}}_{L_2}.$$

First, let's construct a grammar for L_1 . This is the same as in Example 1:

$$S_1 \rightarrow 0S_11 \mid \epsilon.$$

Similarly, we can generate a grammar for L_2 :

$$S_2 \rightarrow 1S_20 \mid \epsilon.$$

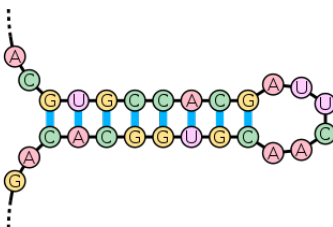
Then we can generate our desired language L with a CFG by 'combining' the two grammars, as follows:

$$\begin{aligned} S &\rightarrow S_1 \mid S_2 \\ S_1 &\rightarrow 0S_11 \mid \epsilon \\ S_2 &\rightarrow 1S_20 \mid \epsilon \end{aligned}$$

Example 3: RNA Secondary Structure

CFGs were applied successfully to RNA secondary structure prediction in the early 90s. We can get a taste of this application through the following task.

Each RNA string is a string over the alphabet $\{A, C, G, U\}$. The string GUGCCACGAUUCAACGUGGCAC can fold into a lollipop shape like this:



The prefix GUGCCACG and suffix CGUGGCAC can together form the handle of the lollipop because they line up in such a manner that As are across from Us (or vice versa) and Cs are across from Gs (or vice versa). The candy part of the lollipop consists of the string AUUCAA, which has length 6.

In general, an RNA string can fold into a lollipop shape if it can be written as xyz , where x and z contain at least two characters each, y contains at least four characters, and the characters of x match up with those of z^R , in the sense that that corresponding pairs of letters fall in the set $\{(A, U), (U, A), (C, G), (G, C)\}$. Can you find a CFG to successfully detect if a string can fold into such a shape?

Induction vs Recursion

As a side note, you might be wondering why we introduced recursion in this supplement, when the current focus of the course this week is induction. This is because the two are actually deeply intertwined. **Induction is often used to prove the validity of recursive algorithms¹.**

¹This fact is super important!

For example, let's try to prove Example 1 using induction. We wish to prove that

$$L = L(G),$$

where $L(G)$ is the language of G . We do so in two parts.

PART 1. We first prove $L \subseteq L(G)$; in other words we prove $P(n)$, the statement

Every string of the form $1^n 0^n$ can be generated by the CFG.

BASE CASE: $P(0)$ is true, since we can apply the rule $S \rightarrow \epsilon$ to generate the string $1^0 0^0 = \epsilon$.

INDUCTIVE STEP: Let $n \in \mathbb{N}$ such that $P(n)$ is true. To generate a string of the form $1^{n+1} 0^{n+1}$, apply the first production rule $S \rightarrow 1S0$ once, obtaining the string $1S0$. Now we can replace S with $1^k 0^k$, which we know can be generated by the CFG by the inductive hypothesis. Thus, we obtain the string $1^{k+1} 0^{k+1} \in L$, so $P(k+1)$ is true.

CONCLUSION: Therefore by induction, $P(n)$ is true for all $n \in \mathbb{W}$.

PART 2. Next, we prove $L \supseteq L(G)$; in other words we prove $P(n)$, the statement

Every string generated by the CFG in n productions is of the form $1^n 0^n$.

BASE CASE: $P(0)$ is true, since if a string in the CFG is generated in only one production it can only be the string ϵ , and $\epsilon = 1^0 0^0$.

INDUCTIVE STEP: Let $n \in \mathbb{N}^+$ such that $P(n)$ is true. For G to generate a nonempty word, the first rule it must apply is $S \rightarrow 1S0$. By inductive hypothesis, the S on the right produces a word of the form $1^n 0^n$ in n productions. Thus the string produced in $n+1$ productions is of the form $1^{n+1} 0^{n+1}$, so $P(n+1)$ is true as well.

CONCLUSION: Therefore by induction, $P(n)$ is true for all $n \in \mathbb{N}^+$.

This was an extremely tedious proof. The worst part? This grammar only had one production! Imagine if there were multiple production rules and multiple variable. A formal proof would take forever! For this reason, we try to generate CFGs simply through our intuition and examples.

In this part, you'll practice generating CFGs to create certain languages.

Implement the following functions:

- **generate_cfg_binary:** In Example 1 above, we presented a CFG to generate a language of the form $1^n 0^n$. Now let's weaken our restrictions even more. Create a CFG to decide the language

$$L_1 = \{w : w \in \{0, 1\}^*, \#1(w) = \#0(w)\};$$

in other words, L_1 represents all binary strings with the same number of 1's and 0's.

- **generate_cfg_union:** In Example 2, you learnt how to create CFGs for the union of multiple languages. Let's put that to the test. Implement a CFG to decide

$$L_2 = \{a^i b^j c^k \mid i, j, k \geq 0, \text{ and } i = j \text{ or } i = k\}$$

- **generate_cfg_rna:** Create a CFG to decide the set of all RNA secondary structures of the form of a "stem loop" as described above. Since terminals are typically lowercase characters, use 'u, g, c, u' instead of 'U, G, C, U'.
- **generate_cfg_tricky:** This one is tricky. Generate a CFG to decide the language

$$L_4 = \{1^i 0^j : 2i \neq 3j + 1\}$$

Part 3: Syntactical Structures (9 points)

This brings us to Noam Chomsky. He proposed that language is innate to humans and that we are born with a universal grammar that allows us to learn and produce language. He argued that this universal grammar consists of a set of **rules** or principles that are hard-wired into our brains, and that these rules underlie the structure of all human languages. Sound familiar? Indeed, Noam Chomsky’s rules are formulated in the exact structure of a context-free grammar.

Even English can be (somewhat) modeled as a context-free grammar. In other words, the English language can be generated by a set of rules that define the relationship between words, phrases, and clauses, without considering the context in which they appear. In this section, we provide a brief overview of how this modeling works.

Consider the following sentence:

“The cat sat on the mat.”

This simple sentence contains a subject (“cat”) and a predicate (“sat on the mat”), which are both necessary components for a complete sentence in English. However, English also allows for more complex structures by using phrases and clauses. For example,

“The cat, which was jet black, sat on the mat.”

In this sentence, we have added a relative clause (“which was jet black”) to provide more information about the subject.

Personally, I think one of the strongest examples of CFGs in English arises when combining sentences using conjunctions, as in the following:

$$\underbrace{\text{The cat sat on the mat}}_{\text{sentence}} \text{ and } \underbrace{\text{the dog sat on the log}}_{\text{sentence}}$$

This is a form of “recursion” in English – enabled by our ability to multiply meanings by a kind of mental recycling of structures.

While these simple examples provide a cursory introduction, [further resources](#) should be consulted to gain a deeper understanding of the complex syntactical structures needed for the English language (and to complete the implementation requirements).

Implement the following functions:

- `generate_cfg_logic(atoms)` In Part 1, we created a parser for propositional strings. Since parsers are typically created for context free grammars, it is natural to assume there exists a CFG for the language of propositions, which there is.
- `generate_cfg_english(parts_of_speech)`: Create a CFG for a reasonable implementation of the english language. Your implementation should cover the main parts of speech: nouns, verbs, adjectives, adverbs, prepositions, articles, pronouns, conjunctions. An example of `parts_of_speech` is as follows:

```
parts_of_speech = {
    "singular_noun": {"dumbbell", "barbell", "ab roller", "treadmill",...}
    "plural_noun": {"dumbbells", "barbells", "ab rollers", "treadmills",...}
    "proper_noun": {"Reese", "Paul", "Sofia", "Saloni", "Ananya", "Ron"...}
    "intransitive verb": {"exercise", "run", "swim", "deadlift", "bench",...}
    "transitive verb": {"lift", "carry", "deadlift", "bench", ...}
    "adjective": {"fit", "athletic", "healthy", "motivated", "resilient"...}
    "adverb": {"quickly", "slowly", "eagerly", "steadily", "loudly",...},
    "preposition": {"in", "on", "with", "at", "from", "over", "under",...},
    "article": {"the", "a", "an", ...},
    "pronoun": {"he", "she", "they", "it", ...},
    "conjunction": {"and", "or", "but", ...},
}
```

For full credit, your CFG must be able to generate sentences of the following structures (ignoring tense, punctuation and capitalization):

- The trainer carried the dumbbells.
- She ran on the treadmill quickly and enthusiastically.
- They exercised with the ab roller and the jump rope.
- The man swims.
- Reese gave her pull-up bar to Paul.
- Ron ate his cold, delicious protein shake.
- The very extremely tall, intelligent woman began to deadlift.
- The motivated fellow lifted the weights, but the unmotivated fellow dropped them.
- With the ab roller, you can strengthen your core.
- A fit and healthy lifestyle is a goal worth pursuing.

Note: Your CFG also does not have to generate semantically correct sentences, for example since we did not distinguish between common and nouns, it is possible to generate the sentence “She benched Dave enthusiastically.”, which semantically does not make sense.

Submission Instructions (10 pts)

After you fill the appropriate functions, submit the following files to Gradescope.

- `proposition_parser.py`

- `generate_grammar.py`

Most of the autograder is given to you this time, so you should be able to check your work locally.

References

- [1] Gonczarowski, Y. A., & Nisan, N. (2022). *Mathematical Logic through Python*. Cambridge University Press.
- [2] Sag, I. A., Bender, E. M., & Wasow, T. (2006). *Syntactic theory: A formal introduction*. CSLI.