

Learning Support for Writing Proofs in Coq

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Learning Support for Writing Proofs in Coq

Context

Introduction

The goal of this project is to provide learning support for students enrolled in YSC3216: Functional Programming and Proving (FPP), by providing a tool that generates corrective suggestions for syntax issues, to be used by students to check their code.

FPP is a course in Yale-NUS College under the Mathematical, Computational and Statistical Sciences major, most recently offered in AY19/20 Semester 1 and taught by Professor Danvy. FPP introduces students to the Coq proof assistant, which is a system for writing and verifying formal proofs. Throughout the course, students learn that precision and orderliness in their code both reflects and encourages clarity of thought. Therefore, one of the primary learning goals for the first half of the course is to build muscle memory for basic proof techniques and programming habits.

To this end, I implement a program called the `proof-reader` that acts a set of 'safety rails' to guide students towards good proving and programming habits via automated intervention on syntax issues. In particular, this tool will help enforce prescribed techniques: using only tactics introduced in the course, and applying tactics explicitly. Since these issues are often highlighted in written feedback from the instructor, the tool also supports students' learning by greatly reducing the feedback cycle and allowing the instructor to focus on substantive rather than superficial feedback.

The program relies on a grammar specification of a subset of Coq as an input to the tool, which the instructor may modify as the course evolves. The program's interface is simple Emacs command, and is intended to be used interactively by students, both during proof editing and before submission.

Functional programming (FP)

Functional programming is a programming paradigm that models programs as mathematical functions. Students taking FPP are expected to have completed Intro to Computer Science taught in Yale-NUS, which trains them in functional programming with the language OCaml. Coq has a language of programs that is very similar to OCaml, and is in fact written in OCaml.

Proving

In mathematics, a proposition is a statement that either holds or does not hold; a proposition is also sometimes called a theorem or lemma. Proofs may be defined as a logical argument about whether a proposition holds. Proofs use logical rules to demonstrate that what we know or assume to be true – an axiom – implies the truth of something that we do not know - a proposition.

Propositions often contain equations, which are statements asserting the equality of two expressions containing variables. When writing proofs (including proofs in Coq), we exercise equational reasoning: we apply axioms to equations in order to incrementally transform them into something that is clearly true.

The Coq proof assistant

Many proofs in mathematics or computer science are natural language proofs - that is, they are written in a natural language, like English. Since natural languages are often ambiguous, natural language proofs are susceptible to misinterpretation or misconception.

On the other hand, just as there programming languages that express a set of instructions to be executed by a computer, there are also domain-specific languages for writing formal proofs that can be automatically, or mechanically, verified by a computer. Coq allows us to write formal, verifiable proofs in a structured logical language called Gallina, and will also automatically verify that our proofs are correct.

YSC3236: Functional Programming & Proving (FPP)

FPP is taken not only by Yale-NUS students, but also PhD and post-doctoral students from the National University of Singapore (NUS) School of Computing (SoC). Through the course, students gain an appreciation for the interconnectedness of computer programs and logical proofs - which have previously been presented to them as distinct domains of knowledge. For example, they are led to realize that a Coq proof exactly corresponds to an equivalent mathematical proof they have written in detail, by hand.

Students engage in weekly assignments consisting of rigorous, progressive exercises involving:

- writing mathematical proofs
- writing programs, and proofs about the properties of programs
- eventually, stating their own theorems and proving them.

By the end of the course, students will have independently written more proofs than they have ever written in their lives, all of which would have been formally verified.

The GNU Emacs text editor

Emacs is a family of real-time text editors characterized by their customizability and extensibility. GNU Emacs was written in 1984 by GNU Project founder Richard Stallman. At Yale-NUS College, GNU Emacs is used in Intro CS, Intro to Algos and Data Structures, and FPP. GNU Emacs provides a language called Emacs Lisp that is used to write programs run within Emacs. The `proof-reader` tool uses Emacs Lisp to provide a user interface.

Proof General

Proof General is an Emacs interface for proof assistants, developed at the University of Edinburgh since 1992. It provides a common interface across various proof assistants, including Coq, and allows users to interactively edit proof scripts. The `proof-reader` tool interacts with Proof General functions in order to provide its functionality.

Motivation

Building muscle memory

The learning philosophy of FPP is that programming and proving is similar to training in any skilled discipline such as martial arts, cooking, or dance: beginner training should build muscle memory for basic skills and habits. For example, if you are training to be a chef, but you don't develop proper knife skills early on, this will hurt you for the rest of your career.

Therefore, in the first half of the course, students complete rigorous, progressive exercises in order to practice specific proof techniques and programming habits. In the second half of the course, students can then rely on this muscle memory to write proofs with greater creativity and efficiency.

Syntax issues

Just as there are many ways to write the same program, there are many equivalent versions of a Coq proof, especially because Coq is flexible and allows you to take shortcuts. However, for new learners, this power can be counterproductive. In the context of FPP, several issues arise.

1. Abuse of tactics

First, students may abuse tactics that have not been introduced in the course. When students get stuck on a proof, they might Google for related solutions or search the Coq documentation for anything that will 'solve' the proof. They might end up using a 'magical' tactic, for example the tactic 'trivial', as in the example proof below.

```
1 Lemma SSSn_is_3_plus_n :  
2   forall n : nat,  
3   S (S (S n)) = 3 + n.  
4 Proof.  
5   trivial.  
6 Qed.
```

Under the hood, the 'trivial' tactic iterates over various strategies to solve the current formula. However, in the first half of the course the focus is for students to understand every single proof step they write. Therefore, using a tactic like 'trivial' goes against the objective of the exercise. Instead, the proof should demonstrate every step:

```
1 ...  
2 Proof.  
3   intro n.  
4   rewrite <- (Nat.add_1_1 n).  
5   rewrite <- (plus_Sn_m 1 n).  
6   rewrite <- (plus_Sn_m 2 n).  
7   reflexivity.  
8 Qed.
```

Yet these tactics still appear in student submissions. This causes time between resubmissions to be wasted on superficial feedback.

2. Misuse of tactics

Second, even when students use tactics that have been introduced, they may misuse them. For instance, the rewrite tactic is used to apply a rewrite rule to the current goal. A rewrite rule is a function that expects arguments that refer to corresponding terms in the goal; Coq will rewrite the corresponding terms in the goal. For example, the rewrite rule `Nat.add_assoc` accepts three arguments, `n`, `m`, and `p`:

```
1 Check Nat.add_assoc.  
2 # Nat.add_assoc : forall n m p : nat, n + (m + p) = n + m + p.
```

However, Coq is flexible with the number of arguments given to terms. As the example proof below demonstrates, you could give the rewrite rule three, two, one or zero of the rewrite arguments required, and Coq will simply infer the intended application, by picking the first terms in the formula that it can apply the rule to.

```
1 Proposition add_assoc_nested :
2   forall a b c d e: nat,
3     a+b+c+d+e=
4     a + (b + (c + (d + e))).
5 Proof.
6   intros a b c d e.
7   rewrite -> (Nat.add_assoc a b).
8   rewrite -> (Nat.add_assoc (a + b)).
9   rewrite -> Nat.add_assoc.
10  reflexivity.
11 Qed.
```

However, in FPP, tactic applications should be explicit, i.e. rewrite rules should be supplied with the exact number of arguments required, so that it is clear which part of the goal is being rewritten:

```
1 ...
2 Proof.
3   intros a b c d e.
4   rewrite -> (Nat.add_assoc a b (c + (d + e))).
5   rewrite -> (Nat.add_assoc (a + b) c (d + e)).
6   rewrite -> (Nat.add_assoc (a + b + c) d e).
7   reflexivity.
8 Qed.
```

This issue arises for the tactics `rewrite`, `exact` and `apply`.

The goal: automated intervention on syntax issues

These two issues - abuse and misuse of tactics - correspond to issues of **abstract syntax** (what language constructs are represented in the grammar) and **concrete syntax** (what structures are used to represent language constructs) respectively. These issues are often highlighted in written feedback from the instructor, resulting in much 'superficial' feedback. Superficial feedback is comments on syntax issues, whereas substantive feedback is comments on the logical content of the proof.

`proof-reader` is a tool that anticipates and identifies both abstract and concrete syntax issues. By automatically intervening during the proof editing process, `proof-reader` guides students towards solutions that do not require superficial feedback, allowing the instructor to focus on substantive feedback.

Solution: the `proof-reader` tool

Parsing student submissions with `proof-reader`

The `proof-reader` tool is a parser that emits two types of warnings corresponding to the syntax issues described in the previous section:

1. Warns user of instances where unpermitted tactics are used.
2. Warns user of instances of incorrect arity in terms supplied to tactics such as "rewrite", "exact", "apply", etc.

`proof-reader` is intended to be used by students, both during proof editing and as a final check before submission. As students are writing proofs, the `proof-reader` keeps them on the right track by correcting issues that might be affecting their thought process. When used as a final check, it will help them correct proofs that might have been accepted by Coq but do not demonstrate the intended learning goals of the exercise. The `proof-reader` can be used directly on student submissions by the instructor as well.

Usage

To run `proof-reader` on your proof script, simply execute the following Emacs command while in Proof General, with the editor focused on the buffer containing the script:

```
1 | M-x jeremy-proof-reader
```

The script will first be re-run from the beginning by Proof General. `proof-reader` will then evaluate the script and display any relevant warnings in the Emacs response buffer, for example:

```
1 | WARNING: In tactic invocation REWRITE on parent term (Nat.add_assoc a b):  
2 | Term "Nat.add_assoc" with arity 3 incorrectly applied to 2 terms (a),(b).
```

Or, if there are no warnings:

```
1 | No warnings.
```

Setup

Simply download the source code or binary package and follow the installation steps from this repository: <https://github.com/jeremyyew/ync-capstone>.

Examples

Example 1: Warning user of instances where unpermitted tactics are used

When `proof-reader` is applied to the example in chapter XX, the output is:

```
1 | Parser error: Could not parse the substring "trivial.". "trivial" may be an  
   | unpermitted tactic, please only use tactics that have been introduced in the  
   | course.
```

Example 2: Warning user of instances of incorrect arity

When `proof-reader` is applied to the example in chapter XX, the output is:

```
1 | WARNING: In tactic invocation REWRITE on parent term (Nat.add_assoc a b):
2 |   Term "Nat.add_assoc" with arity 3 incorrectly applied to 2 terms (a),(b).
3 |
4 | WARNING: In tactic invocation REWRITE on parent term (Nat.add_assoc (a+b)):
5 |   Term "Nat.add_assoc" with arity 3 incorrectly applied to 1 terms (a+b).
6 |
7 | WARNING: In tactic invocation REWRITE on parent term (Nat.add_assoc):
8 |   Term "Nat.add_assoc" with arity 3 incorrectly applied to 0 terms.
```

Possible errors

`proof-reader` only accepts syntactically correct Coq code. It will first trigger Proof General to reevaluate the entire buffer. As long as Proof General accepts the script without error, `proof-reader` will process it.

If there are Coq syntax errors, `proof-reader` will display:

```
1 | Coq error raised. Please correct and try again.
```

The parser will then terminate without evaluating the script. The Coq errors will be in the response buffer, as usual.

Furthermore, `proof-reader` only accepts a subset of Coq syntax, which has been pre-defined by the instructor (See "Appendix/Supported syntax"). Therefore, if the script contains unsupported syntax, it is likely either a command that has been introduced in the course but not accounted for, or a bug. `proof-reader` will display:

```
1 | Parser error: Could not parse "XXX". This syntax may not be currently supported.
```

To extend the supported syntax or modify the parser behaviour, see "Design and implementation/Extensibility".

Design and implementation

Backus-Naur Form (BNF)

Before writing a parser for a language we must understand the specification of the language. Coq consists of various sublanguages for different purposes, each with their own specifications available in the documentation. The language used to write Coq proofs is the Gallina specification language, which interacts with the language of tactics. These sublanguages are specified in Backus-Naur Form.

The Backus-Naur Form is a notation describing a context-free grammar. A BNF specification consists of expressions on the right - consisting of both 'terminals' (literals) and 'non-terminals' (variables) - represented by non-terminals on the left.

BNF specifications for different languages will have particular conventions but follow the same overall structure. For example, here is a fragment of the BNF specification for the vernacular of Gallina:

```

1  assertion      ::=  assertion_keyword ident [binders] : term .
2  assertion_keyword ::= Theorem | Lemma
3                      Remark | Fact
4                      Corollary | Property | Proposition
5                      Definition | Example
6  proof          ::= Proof . ... Qed .
7                      Proof . ... Defined .
8                      Proof . ... Admitted .

```

The grammar module: a BNF-inspired data structure

The code referenced in this section can be found in `jeremy-parser/grammar.py` unless otherwise specified.

`proof-reader` parses a given input string into a syntax tree which can be easily traversed and thereby evaluated. In parsing a language, there are many levels of abstraction that we can utilize. The highest level of abstraction would be a parser generator that directly accepts a BNF specification and returns a fully functioning parser with zero or minimal lines of code needed (I explain why this path is not taken in the "Discussion" chapter.) Through trial and error we eventually arrived at a mid-level abstraction. We define a data structure that resembles BNF and dictates the flow of the parser, provided by the `grammar` module.

Instead of hard-coding conditional branches to determine matching patterns at each step, the `grammar` module provides the `GRAMMAR` variable, which is a map (Python dictionary) from a grammar rule to a tuple containing a regular expression and a list of child grammar rules. As the parser constructs a syntactical unit and its children, it performs lookups to the data structure to understand what rules to apply at each step.

Here is a truncated version of the actual `GRAMMAR` map (the full structure is in the `grammar.py` file). As you can see, each rule mirrors a non-terminal on the left, and its regular expression and children represent its expression on the right. (Note the indentation does not correspond to actual nesting of data, but is intended to visually reflect the nested definitions).

```

1  ...
2  GRAMMAR = {
3      LABEL_DOCUMENT: (None,
4          [...
5              LABEL_PROOF,
6              LABEL_REQUIRE_IMPORT,
7              LABEL_SEARCH]),
8      ...
9      LABEL_PROOF:
10         (fr"{KW_PROOF}\.(.*?)(?:{KW_QED}|{KW_ADMITTED}|{KW_ABORT})\.",
11         [LABEL_APPLY,
12         LABEL_ASSERT,
13         ...]),
14         ...
15         LABEL_REWRITE:
16         (fr"{KW_REWRITE}({REGEXP_REWRITE_ARROW}?
17         {REGEXP_TERM_OPTIONAL_SPACE}){REGEXP_IN_OCCURRENCE}{REGEXP_AT_OCCURRENCE}
18         {REGEXP_TACTIC_END}{REGEXP_TACTIC_LOOKAHEAD}",
19         [LABEL_REWRITE_ARROW,
20         LABEL_TERM]),
21         LABEL_REWRITE_ARROW: (fr"({REGEXP_REWRITE_ARROW})", []),

```



```

20 | ...
21 | ...
22 | }

```

For example, the `PROOF` rule states: "a proof is a lazily matched substring beginning with the keyword 'Proof' plus a period, and ends with either the keyword 'Qed', 'Admitted', or 'Abort', plus a period. The inner string must consist of any number of child components matching the rules `APPLY`, `ASSERT`, etc".

Observe:

- For readability and code reuse, we factor out constants (keywords, regular expressions) into the `constants.py` module.
- A terminal node has an empty rule list. For example, `LABEL_REWRITE_ARROW` is a terminal.
- This data structure is simple. While it serves the purpose of this project, see "Limitations of the `grammar` module".

Having generated a syntax tree, we are now in a position to traverse and evaluate it.

Parsing input to generate a syntax tree

The code referenced in this section can be found in `jeremy-parser/proof_reader.py` unless otherwise specified.

Constructing syntax tree nodes

First, we define a `Node` object.

```

1 class Node:
2     def __init__(self, label: str, val: str = None, children: list = None):
3         # Each node's label identifies its "type", i.e. which syntactical unit
4         # it represents.
5         self.label = label
6         # Each node's value is the contents of the substring it matched on. This
7         # is useful for logging and constructing helpful warning messages.
8         self.val = val
9         # Each node has a list of children, since each syntactical unit may be
10        # comprised of sub-components. Hence a node with no children is a terminal node.
11        self.children = children or []

```

The main parser function `construct_node` is responsible for recursively constructing a syntax tree composed of `Node` objects, and returning it. It accepts an input string `s` and `rule`, the label of the rule that the string has been matched on.

```

1 def construct_node(s: str, rule: str) -> Node:
2     def construct_children(s: str, parent: str, acc: list) -> List[Node]:
3         if not s:
4             return None, acc, ""
5         _, expected = grammar.GRAMMAR[parent]
6         if expected == []:
7             return s, acc, ""
8         exception = None
9         for item in expected:
10            pattern, _ = grammar.GRAMMAR[item]

```

```

11         match = re.match(pattern, s)
12         if not match:
13             continue
14         if item == LABEL_TERM:
15             term_s, remaining_s = get_next_subterm(match.group(1))
16             term = construct_term(term_s)
17             return term_s, acc+[term], remaining_s
18         try:
19             child, remaining_s = construct_node_helper(
20                 match.group(1), item)
21             remaining_s = remaining_s + s[match.end():]
22             return construct_children(remaining_s, parent, acc + [child])
23         except (UnmatchedToken, UnmatchedTactic) as e:
24             exception = e
25     if exception:
26         raise exception
27     if parent == LABEL_PROOF:
28         raise UnmatchedTactic(s)
29     raise UnmatchedToken(s)
30
31     def construct_node_helper(s: str, rule:str) -> (Node, str):
32         term_s, children, remaining_s = construct_children(s, rule, [])
33         node = Node(rule, term_s or s)
34         node.children = children
35         return node, remaining_s
36
37     node, _ = construct_node_helper(s, rule)
38     return node

```

It is only called once by the main program, with the entire script as input. It initiates the construction of the tree by calling `construct_node_helper` and assumes there will be no remaining string to parse (since the root node 'DOCUMENT' accepts the entire script).

`construct_node_helper` is responsible for recursively constructing the current subtree, which comprises of both the current node as well as all its children. It accepts an input substring `s` and `rule`, the label of the rule that the substring has been matched on. Only after successfully constructing all its children will it then create the current node and assign the children. It returns both a syntax subtree as well as the remaining string that contains sibling subtrees to be further parsed.

`construct_children` accepts an input substring `s`, its parent `rule`, and an accumulator `acc` containing the child nodes it will return. It returns `term_s` (the value of the parent substring, should this be different from the original substring `s`), a list of child nodes `children`, and the remaining string that contains sibling subtrees to be further parsed, `remaining_s`. It proceeds as such:

- `construct_children` first performs a lookup in the grammar module to obtain the expected children of `rule`.
- For each rule, it performs a lookup in the grammar module to obtain the corresponding regular expression. It attempts to match the expression against the beginning of `s`.
- On a match, it calls `construct_node_helper` to construct the current child's subtree, and then recurses on the remaining string to construct the rest of the children.

Observe that in `construct_children`, instead of recursing on the rest of the substring following the matched capture group `s[match.end():]`, we return `remaining_s` from `construct_node_helper`, which always demands it from `construct_children`, which is ultimately returned as an empty string in `construct_children`'s base cases, or extracted by `construct_term`. We then recurse on `remaining_s + s[match.end():]`. This is because we cannot rely solely on regular expressions to correctly capture the exact substring containing the next subtree.

When parsing a tactic invocation containing a term, we need to count the parentheses in the term to determine its endpoint. Afterwards, we can continue parsing the rest of the substring, prefixed with any extraneous characters the regular expression might have captured. Therefore, if the current substring matches a term, we make a call to `construct_term`.

Constructing terms and subterms

In order to validate the arity of terms supplied to tactics such as `rewrite`, `exact` and `apply`, we need to parse a term into its subterms, which are grouped in nested parenthesis. Regular expressions are not expressive enough to capture nested patterns. Here is an example substring:

```
1 | exact (my_lemma_1 (my_lemma_2 n1) n2).
2 | exact (my_lemma_3 n3).
```

Suppose we have constructed the first `exact` node, and now we need to capture the parent term `(my_lemma_1 (my_lemma_2 n1) n2)`, before trying to capture its children `my_lemma_1`, `(my_lemma_2 n1)` and `n2`.

- A lazy pattern on opening and closing parenthesis, such as `\(.*\)`, would capture:
 - `(my_lemma_1 (my_lemma_2 n1))`.
- On the other hand, a greedy pattern like `\(.*+\\)` would capture everything until the last parenthesis in the substring:
 - `(my_lemma_1 (my_lemma_2 n1) n2).exact (my_lemma_3 n3).`

Hence the regular expressions used in the generic `construct_node` function cannot construct this subtree, and we need a specialized `construct_term` function which is able to count parentheses:

```
1 | def construct_term(term: str) -> Node:
2 |     def construct_subterms(s: str) -> List[Node]:
3 |         if s == "":
4 |             return []
5 |         subterm, remaining = get_next_subterm(s)
6 |         child = construct_term(subterm)
7 |         children = [child] + construct_subterms(remaining)
8 |         return children
9 |     if term and term[0] == "(" and term[-1] == ")":
10 |         term = term[1:-1]
11 |     node = Node(LABEL_TERM, term)
12 |     if re.fullmatch(r"^[^\\s]+", term):
13 |         return node
14 |     node.children = construct_subterms(term)
15 |     return node
```

The familiar problem of counting parenthesis is implemented iterative-style in `get_next_subterm`:

```

1 def get_next_subterm(s: str) -> (str, str):
2     k = 0
3     for i, c in enumerate(s):
4         if c == " " and k == 0:
5             return s[:i], s[i+1:]
6         elif c == '(':
7             k += 1
8         elif c == ')':
9             k -= 1
10    if k != 0:
11        logger.info("Invalid parentheses.")
12        raise Exception("Invalid parentheses.")
13    return s, ""

```

Example syntax tree

Here is an example proof from the lecture notes of week 2 of FPP AY 19/20 Semester 1.

```

1 Require Import Arith Bool.
2 Check Nat.add_0_r.
3 Proposition first_formal_proof :
4     forall n : nat,
5         n + 0 = 0 + n.
6 Proof.
7     intro n.
8     Check (Nat.add_0_r n).
9     rewrite -> (Nat.add_0_r n).
10    Check (Nat.add_0_l n).
11    rewrite -> (Nat.add_0_l n).
12    reflexivity.
13 Qed.

```

Here is the resulting syntax tree, pretty-printed:

```

1 |- DOCUMENT:
2   "Require Import Arith Bool...Qed."
3   |- REQUIRE_IMPORT:
4     "Require Import Arith Bool."
5   |- CHECK:
6     "Check Nat.add_0_r."
7   |- ASSERTION:
8     "Proposition first_formal_proof : forall n : nat, n + 0 = 0 + n"
9     |- ASSERTION_KEYWORD:
10      "Proposition"
11     |- ASSERTION_IDENT:
12      "first_formal_proof"
13     |- FORALL:
14      "n : nat"
15      |- BINDER:
16        "n"
17      |- TYPE:
18        "nat"
19     |- ASSERTION_TERM:

```

```

20         "n + 0 = 0 + n"
21     |- PROOF:
22         "intro n.Check (Nat.add_0_r n).rewrite -> (Nat.add_0_r n).Check
(Nat.add_0_l n).rewrite -> (Nat.add_0_l n).reflexivity."
23     |- INTRO:
24         "n"
25     |- CHECK:
26         "Check (Nat.add_0_r n).\"
27     |- REWRITE:
28         \"(Nat.add_0_r n)\"
29         |- REWRITE_ARROW:
30             \" -> \"
31     |- TERM:
32         \"Nat.add_0_r n\"
33         |- TERM:
34             \"Nat.add_0_r\"
35         |- TERM:
36             \"n\"
37     |- CHECK:
38         \"Check (Nat.add_0_l n).\"
39     |- REWRITE:
40         \"(Nat.add_0_l n)\"
41         |- REWRITE_ARROW:
42             \" -> \"
43     |- TERM:
44         \"Nat.add_0_l n\"
45         |- TERM:
46             \"Nat.add_0_l\"
47         |- TERM:
48             \"n\"
49     |- REFLEXIVITY:
50         \"reflexivity.\"

```

Observe:

- `ASSERTION` subtrees have `ASSERTION_IDENT` as well as `BINDER` arguments so the `collect_arity` function simply has to store the identifier with the number of arguments.
- `TERM` subtrees have subterms as children, so the `check_arity` function simply has to validate the number of terms at each depth.

Feature 1: Recognizing unpermitted tactics

We define two custom exceptions in `jeremy-parser/proof_reader.py`:

```

1  class UnmatchedTactic(Exception):
2      def __init__(self, remaining):
3          self.remaining = remaining
4          self.tactic = None
5          match = re.match(fr"(.+?){REGEXP_TACTIC_END}{REGEXP_TACTIC_LOOKAHEAD}",
6                          self.remaining)
7          if match:
8              self.tactic = match.group(1)
9
10 class UnmatchedToken(Exception):
11     def __init__(self, remaining):
12         self.remaining = remaining

```

In general, when `proof-reader` encounters unsupported syntax, it will not continue parsing since it is impossible to differentiate the unsupported syntactical unit and the rest of the script. Thus it raises the `UnmatchedToken` exception (see "Constructing syntax tree nodes"), which will be used to display the remaining substring.

When it encounters an unrecognized token that might be an unpermitted tactic invocation – i.e. any unsupported syntax within a `PROOF` node – it raises `UnmatchedTactic` instead, which attempts to extract a specific tactic invocation from the remaining substring. This is used to display a more precise warning message, such as in "Example 1: Warning user of instances where unpermitted tactics are used".

Feature 2: Verifying arity

The code referenced in this section can be found in `jeremy-parser/proof_reader.py` unless otherwise specified.

As we evaluate the syntax tree that we have generated, we need to compare the arity of its assertions with assertions that have been defined in the environment – both assertions that have been defined earlier in the script, as well as predefined theorems from built-in libraries that have been imported.

Collecting arity signatures of built-in library theorems

We must have a database of arities for built-in library theorems. This only needs to be done once, or whenever new modules are added to the course (see Appendix for modules used in the course).

For each module, we manually issue a Search command, e.g. `Search _ inside Nat` to list all theorem definitions in the response buffer. We save each list as a string, process the string so it resembles Coq code, and generate a syntax tree containing those definitions using the same `construct_tree` used for input code. We then run `collect_arity`, defined below, returning the dictionary `arity_db`, a map from assertion names to arities. We save this dictionary as a file, to be loaded whenever `proof-reader` is run. Later, we add user-defined lemmas to this dictionary we parse the input syntax tree.

This procedure is performed by the script `jeremy-parser/script_parse_theory_lib.py`.

```

1  def collect_arity(t: Node, arity_db : dict) -> dict:
2      assert(t.label == LABEL_DOCUMENT)
3      for child in t.children:
4          if child.label != LABEL_ASSERTION:
5              continue
6          assertion = child
7          if len(assertion.children) < 3:

```

```

8         continue
9         ident = assertion.children[1]
10        forall = assertion.children[2]
11        binders = [c for c in forall.children if c.label == LABEL_BINDER]
12        arity = len(binders)
13        arity_db[ident.val] = arity
14    return arity_db

```

Collecting arity signatures of the syntax tree

In `check_arity`, before evaluating the input syntax tree, we use the same `collect_arity` function to collect arity signatures of assertions that have been defined in the script.

```

1 def check_arity(t: Node, arity_db: dict) -> (list, list):
2     ...
3     def check_subterms(subterms: list, parent_term: Node, parent_tactic_label:
4         str):
5         ...
6         def traverse(t: Node):
7             ...
8             arity_db = collect_arity(t, arity_db)
9             traverse(t)
10        return warnings, ...

```

We could collect arity signatures and check arity at the same time instead of two separate steps, therefore performing a single traversal instead of two traversals, since the tree traversal is left-to-right, and assertions must be declared before they are referenced. This was the approach in an earlier implementation. However, we implement two separate traversals for the sake of readability. Since each traversal is done in constant time, there is insignificant difference in actual performance. Furthermore, Coq does not allow repeat definitions so we will not have to deal with multiple signatures for the same assertion.

Checking arity of the syntax tree

The `check_arity` function can be found in `jeremy-parser/proof_reader.py`. It is responsible for traversing the input syntax tree `t` given a map of arity signatures `arity_db`, and returning a list of warnings indicating instances where the expected arity is not equal to the actual arity. Expected arity is the integer value found in `arity_db` if the assertion has been defined. Actual arity is the number of subterms following the first term. `check_arity` simply explores relevant nodes – `REWRITE`, `EXACT`, and `APPLY` – and counts the number of subterms at each level, generating a warning when arity is incorrect.

Extending proof-reader

The grammar module abstraction enables syntax to be extended conveniently. To extend the supported syntax - for example, adding a permitted tactic - the instructor simply has to add a rule definition to the grammar module, comprising of a regular expression and the expected child rules. Not all desired syntax may be expressible in the current framework; refer to "Limitations of the `grammar` module".

Unit and acceptance testing

The code referenced in this section can be found in `jeremy-parser/tests.py` unless otherwise specified.

`TestParser` performs unit tests to verify that `construct_node` fulfils its specifications. For each test input it compares the generated syntax tree with the expected syntax tree. Each unit test verifies that variations of a particular syntactical unit is correctly parsed.

`TestParserAcceptance` also performs acceptance tests for `construct_node`, running it on sample student submissions which have provided by the instructor.

`TestParityCheck` performs unit tests to verify that `collect_arity` and `arity_check`. It generates and evaluates the syntax tree for each test input, and compares the output warnings with the expected warnings. It performs both positive tests (inputs that should trigger no warnings) and negative tests (inputs that should trigger warnings), and each input contains variations of `exact`, `rewrite` and `apply` syntax.

Discussion

Time and space complexity

Note that the generated syntax trees are expected to be shallow with some constant depth, with most branches having only a few levels, since the grammar contains non-recursive definitions. The only source of arbitrary depth is `TERM`, but the expected depth of `TERM` subtrees for actual student submissions is around 1-3 levels, since highly nested terms are quite rare in simple proofs, and might call for a goal to be factored into a standalone lemma.

Regular expression matching in `construct_node`

Regular expressions can be computationally expensive - for example, they might grow exponentially in complexity when catastrophic backtracking occurs. However, all the regex patterns in the `grammar` module use lazy quantifiers, thus avoiding backtracking. Furthermore, almost all matches are performed on Coq sentences or fragments of sentences, which are very short substrings. Lastly, since we only accept input that is syntactically validated by `coqtop`, the input is quite predictable.

Therefore we can reasonably assume total $O(N)$ time where N is the length of the input string; the tree is expected to be shallow, so the same substring will only be processed some constant number of times. We also have $O(1)$ space complexity. Counting parentheses in `get_next_subterm` incurs $O(N)$ time complexity as well.

Constructing and traversing the syntax tree

The `construct_node` function takes $O(N)$ time to construct the entire tree, and $O(N)$ space on the callstack (since `construct_node` is recursive and Python does not have tail-call optimization), where N is the length of the input string and the total number of nodes constructed is proportional to N . A more precise estimate might be $O(\log_k N)$ where k is the average number of nodes generated for each level of the tree. Due to the space consumption, Python's recursion limit should be increased if long proofs are expected (currently set to 10000).

`collect_arity` and `arity_check` will explore nodes only at the first and second level, incurring a $O(N)$ time complexity and $O(1)$ space complexity.

Overall complexity

Therefore, the overall complexity for constructing and evaluating the syntax tree is $O(N)$ time and $O(N)$ space. We also run performance tests on real input as well as large input, the results of which can be found in the Appendix on the repository.

Limitations of the `grammar` module

The grammar module is limited in its expressiveness because of its simple structure. Firstly, rules do not directly express patterns with distinct subcomponents. For example, an assertion is broken down into a keyword, identifier, 'for all' statement, and a term:

```
1 LABEL_ASSERTION:
2     (fr"({REGEXP_ASSERTION} .+?:.+?)\.{REGEXP_DOC_LOOKAHEAD}" ,
3     [LABEL_ASSERTION_KEYWORD,
4     LABEL_ASSERTION_IDENT,
5     LABEL_FORALL,
6     LABEL_ASSERTION_TERM]),
```

However, since `construct_node` only expects one capture group, it must iteratively match all subcomponent rules on that single captured substring after the substring has been matched on the assertion. The pattern cannot express subcomponents that appear in different locations of the substring. Furthermore, the structure does not express which subcomponents are required. Each subcomponent rule is treated as optional; it could match on only one of them once, or one of them repeatedly, and as long as the string is eventually consumed the assertion subtree will be treated as a successful match. One solution is for the parser to iterate over a list of capture groups and a list of expected patterns, or a map of named capture groups and their expected patterns.

Secondly, rules do not express alternate patterns. A single regex pattern defines the acceptable structure for each rule. So far we are able to express alternate patterns within the regular expressions as a single pattern with alternating parts. However, a better solution is to define a rule as a list of pattern/children pairs instead of a single pair, with each pair representing an additional matching option.

Fortunately, these issues do not seem to pose a problem for the input we have tested on, but making the module more expressive would make matching more precise, and improve extensibility.

Alternative approaches

Modifying Coqtop source code

Given that Coq's ability to infer missing arguments is an additional feature, it seems natural to modify the source code to provide an option to turn the feature off, as opposed to building an external parser from scratch. There are two reasons why this approach was not taken:

Firstly, and most importantly, a source code approach reduces usability and maintainability of the tool. Unless my changes are accepted into the master branch of the open-source Coq codebase on Github, students will be locked in to the Coq version I worked with. They would not benefit from updates to Coq and might have limited access to the Coq ecosystem. The features described in this report are quite prescriptive and highly specific to FPP's learning goals. While the tool has broader applications as an approach to educators with similar goals, its features may not be relevant to the average Coq user. Granted, we could maintain a modified branch in a separate repository, into which we merge Coq updates. But this involves repeated reintegration and might also introduce dependency or installation issues.

Secondly, a source code approach did not seem feasible. I judged that it was out of scope for this project due to my limited experience with large-scale Ocaml applications. Even if I managed to achieve my desired functionality, refactoring source code might introduce invisible bugs in other components.

Using a parser generator

Writing parsers for programming language is a well-understood problem, and parser generators automate the implementation of parsing algorithms. A parser generator accepts a grammar specification and produces a parser that can evaluate the specified language. I spent a significant amount of time exploring the use of parser generators to build my tool. Using a parser generator was appealing because I did not want to 're-invent the wheel', and it promised quick development, high-level abstraction, and high performance.

I tried using several parser generators, for example CEDET's built-in Wisent, and the python package Lark. Each had their own issues. For example, I had some success specifying the grammar of my Coq sublanguage with Lark, but there were often bugs that I had to find workarounds for because I did not understand the error messages. The algorithms were quite complex and I did not have full visibility or understanding of the underlying processes. Furthermore, there were certain functionalities (e.g. collecting and storing previous arity signatures) that did not seem possible in the existing frameworks - modifying the source code was possible but complicated.

Ultimately, for the purposes of my project, writing a parser from scratch gave me granular control of my development process and allowed me to make informed decisions on the level of abstraction to use for different components.

Future work

Firstly, implementing a more expressive grammar structure would improve extensibility and precision, as detailed in "Limitations of the grammar module".

Secondly, implementing the entire program in Emacs-lisp would allow the program to be run directly in Emacs instead of as a child process. This would likely improve performance and eliminate any installation or interoperability issues.

Thirdly, modifications or extensions to the grammar should be made possible at runtime instead of requiring source-code modification. This would avoid rebuilding the project and improve the user experience for the instructor. In this implementation, this can be done by importing the grammar module from outside the binary package, similar to how a parser generator might take a specification as input. Closer integration with Proof General would also allow new library modules to be added via an Emacs command.

Fourthly, performance profiling can be run to identify bottlenecks - string slicing should be eliminated to avoid unnecessary copying, and an iterative implementation might be faster.

Reflections

In the process of developing `proof-reader` I gained some insights on development in general. First and foremost, we need to have clear specifications from which we can write unit tests. In this project I had a set of example inputs and expected warnings, but would have saved time if I expanded on them by running more acceptance tests earlier in order to discover more edge cases.

Secondly, we should always take advantage of any features that ensure type correctness, as early as possible (I used type annotations from Python's built-in `typing` module). They help us think more precisely especially in the planning phase, which is really where most of the work happens.

Lastly, in designing a program, we need to find the right level of abstraction. Even though using a parser generator was appealing, like most 'automagical' tools, it provided a high level of abstraction and fast development, at the cost of flexibility and understanding. I learned this the hard way after spending an entire semester on it only to switch approaches – but it was a lesson worth learning.