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[TODO: Is the “department committee chairman” still Leslie A. Kolodziejski?]

Performance Engineering of Proof-Based Software Systems

by

Jason S. Gross

Submitted to the Department of Electrical Engineering and Computer
Science

in partial fulfillment of the requirements for the degree of

Doctor of Philosophy in Computer Science and Engineering

at the

MASSACHUSETTS INSTITUTE OF TECHNOLOGY

June 2020

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Abstract

[TODO: More formal wording] Formally verified proofs are important. Unfortunately, large-scale proofs, especially automated ones, in Coq, can be quite slow.

This thesis aims to be a partial guide to resolving the issue of slowness. We present a survey of the landscape of slowness in Coq, with a number of micro- and macro-benchmarks. We describe various metrics that allow prediction of slowness, such as term size, goal size, and number of binders, and note the occasional surprise-lack-of-bottleneck for some factors, such as total proof term size.

We identify three main categories of workarounds and partial solutions to slowness: design of APIs of Gallina libraries; changes to Coq's type theory, implementation, or tooling; and automation design patterns, including proof by reflection. We present lessons drawn from the case-studies of a category-theory library, a proof-producing parser-generator, and a verified compiler and code generator for low-level cryptographic primitives.

[TODO: Fix runon sentence] The central new contribution presented by this thesis, beyond hopefully providing a roadmap to avoid slowness in large Coq developments, is a reflective framework for partial evaluation and rewriting which, in addition to being used to compile a code-generator for field arithmetic cryptographic primitives which generates the code currently used in Google Chrome, can serve as a template for a possibly replacement for tactics such as `rewrite`, `rewrite_strat`, `autorewrite`, `simpl`, and `cbn` which achieves much better performance by running in Coq's VM while still allowing the flexibility of equational reasoning.

[TODO: Maybe instead use the alternative from the thesis-proposal?] The proposed research is a study of performance issues that come up in engineering large-scale proof-based systems in Coq. The thesis presents lessons learned about achieving acceptable performance in Coq in the course of case-studies on formalizing category theory, developing a parser synthesizer, and constructing a verified compiler for synthesizing efficient low-level cryptographic primitives. We also present a novel method of simple and fast reification, and a prototype tool for faster rewriting and customizable reduction which does not require extending Coq's trusted code base.

Thesis Supervisor: Adam Chlipala
Title: Associate Professor of Computer Science

Acknowledgments

[**TODO: uniform style**] [**TODO: rearrange**] Thank you, Mom, for encouraging me from my youth and supporting me in all that I do. Last, and most of all, thank you, Adam Chlipala, for your patience, guidance, advice, and wisdom, during the writing of this thesis, and through my research career. [**TODO: Add more acknowledgments**] I want to thank Andres Erbsen for pointing out to me some of the particular performance bottlenecks in Coq that I made use of in this thesis, including those of subsubsection Sharing in Section 2.2.1 and those of subsections Name Resolution, Capture-Avoiding Substitution, Quadratic Creation of Substitutions for Existential Variables, and Quadratic Substitution in Function Application in Subsection 2.2.3. [**TODO: cite various grants**] [**TODO: thank various other people**: Pierre-Marie Pédrot, Rajee Agrawal, friends, Allison Strandberg (incl. for chatting with me about thesis), Matthieu Sozeau, coqdevs more generally, maybe András Kovács on perf discussions] [**TODO: Rajee or Rajashree?**]

This work was supported in part by the MIT bigdata@CSAIL initiative, NSF grant CCF-1253229, ONR grant N000141310260, and AFOSR grant FA9550-14-1-0031. We also thank Benedikt Ahrens, Daniel R. Grayson, Robert Harper, Bas Spitters, and Edward Z. Yang for feedback on “Experience Implementing a Performant Category-Theory Library in Coq” [89].

[**TODO: rearrange**] [**TODO: make sure we didn’t miss anything**] A significant fraction of the text of this thesis is taken from papers I’ve co-authored during my PhD, sometimes with major edits, other times with only minor edits to conform to the flow of the thesis. In particular: [**TODO: how to format citations here**] Sections 3.3, 3.4, 3.5.1, 3.5.2 and 3.5.3 are taken from “Experience Implementing a Performant Category-Theory Library in Coq” [89]. Some of the text in Subsections 8.2.1 and 8.3.1 comes from “Experience Implementing a Performant Category-Theory Library in Coq” [89]

Chapter 5 is largely taken from [**TODO: cite rewriting**].

s ?? and 7.1 is based on the introduction to Gross, Erbsen, and Chlipala [90]. Chapter 7 is largely taken from Gross, Erbsen, and Chlipala [90].

[**TODO: should we include acknowledgements from various papers?**] Category theory paper: This work was supported in part by the MIT bigdata@CSAIL initiative, NSF grant CCF-1253229, ONR grant N000141310260, and AFOSR grant FA9550-14-1-0031. We also thank Benedikt Ahrens, Daniel R. Grayson, Robert Harper, Bas Spitters, and Edward Z. Yang for feedback on this paper.

Reification by parametricity paper: We would like to thank Hugo Herbelin for sharing

the trick with `type of` to propagate universe constraints¹ as well as useful conversations on Coq’s bug tracker that allowed us to track down performance issues.² We would like to thank Pierre-Marie Pédro for conversations on Coq’s Gitter and his help in tracking down performance bottlenecks in earlier versions of our reification scripts and in Coq’s tactics. We would like to thank Beta Ziliani for his help in using Mtac2, as well as his invaluable guidance in figuring out how to use canonical structures to reify to PHOAS. We also thank John Wiegley for feedback on the paper.

For those interested in history, our method of reification by parametricity was inspired by the `evm_compute` tactic [129]. We first made use of `pattern` to allow `vm_compute` to replace `cbv`-with-an-explicit-blacklist when we discovered `cbv` was too slow and the blacklist too hard to maintain. We then noticed that in the sequence of doing abstraction; `vm_compute`; application; β -reduction; reification, we could move β -reduction to the end of the sequence if we fused reification with application, and thus reification by parametricity was born.

This work was supported in part by a Google Research Award and National Science Foundation grants CCF-1253229, CCF-1512611, and CCF-1521584.

¹<https://github.com/coq/coq/issues/5996#issuecomment-338405694>

²<https://github.com/coq/coq/issues/6252>

[**QUESTION FOR ADAM:** Decide on whether to copyright myself or MIT; Adam, are there considerations to be aware of here?] [**TODO:** Download and fill out <https://libraries.mit.edu/wp-content/uploads/2019/08/umi-proquest-form.pdf>] [**TODO:** Change biblatex cite format to author-year or something so that it's easy to pick up references from the paper] [**TODO:** Share thesis with Eric Marion (cf 2020-05-14 15:53 Facebook Messenger)] [**TODO:** Share thesis with András Kovács (cf 2020-03-12 11:15 gmail <https://mail.google.com/mail/u/0/#search/andras/QgrcJHrnvrhWwVxKMQFsctxrGQx>)] [**TODO:** Share thesis with Talia Ringer and ppedrot] [**TODO:** single space or double space?] [**TODO:** maybe remove “introduction” section headings?]

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[**TODO:** decide on chapter/section capitalization and be consistent] [**TODO:** figure out if regression stats should be included, and, if so, how to calculate them] [**TODO:** make sure all appendices are at the very end, so we keep a linear progression] [**TODO:** update rewriting paper chapter with camera-ready file before editing] [**TODO:** add labels on every section, ensure no duplicate labels] [**TODO:** Decide whether or not to use contractions (less formal, but okay for thesis if I make a conscious decision)] [**TODO:** be consistent about $\lambda x.\text{body}$ vs. $\lambda x, \text{body}$ (note that Coq only accepts the latter)] [**TODO:** move any citations which point at people.csail.mit.edu to point at a github.io version, and move my personal website there] [**TODO:** maybe include all code in an appendix?] [**TODO:** be consistent about function extensionality vs functional extensionality] [**TODO:** include github code links for contributions]

Part I

Introduction

The most common mistake in performance engineering is to blindly optimize without profiling; you most often spend your time optimizing parts that aren't actually bottlenecks.

— Charles Leiserson (heavily paraphrased, reconstructed from memory of 6.172)

premature optimization is the root of all evil

— Donald Knuth [106, p. 671]

Testing shows the presence, not the absence of bugs

— Edsger Wybe Dijkstra, 1969 [41]

Chapter 1

Introduction

1.1 Introduction

As our world comes to rely more and more on large and complex software, the impact of bugs in our code grows ever more severe. Bugs in the software of the Therac-25, a radiation-therapy machine, killed three people and injured at least three more between 1985 and 1987 by delivering potentially lethal doses of radiation. [192] [QUESTION FOR ADAM: do I need a better citation than wikipedia?] [TODO: should I mention the pricing bug on amazon? <https://www.usatoday.com/story/tech/2019/07/19/amazon-prime-day-error-sold-13-000-camera-equipment-94-report/1775314001/> <https://www.theguardian.com/money/2014/dec/14/amazon-glitch-prices-penny-repricerexpress>, <http://www.michaeleisen.org/blog/?p=358>] More recently, unforeseen edge-cases in DAO smart-contracts nearly cost investors over 40 million USD. [91] [TODO: should I mention anything else? the pentium bug costing Intel \$475 million? <http://top-100s.blogspot.com/2009/05/10-most-expensive-software-blunders.html>, \$500 million Ariane Rocket disaster? , <https://www.olenick.com/blog/articles/infamous-software-bugs-fdiv-bug>] [TODO: should I pull from <https://xavierleroy.org/talks/Milner-award-lecture.pdf>? “Fiat-Chrysler recalled 1.5 M vehicles for software update.”]

While testing of various kinds is widely used to catch bugs, it’s quite costly [QUESTION FOR ADAM: should I email Xavier Leroy and ask if he has references on how much the airline industry spends on testing?] and does not guarantee a lack of bugs. In some adversarial contexts, random testing cannot hope to eliminate all bugs. For example, some bugs in cryptographic code only occur in as few as 19 out of 2^{255} cases. [118] Suppose you wanted to catch this bug in a “modest” twenty years of continuous random testing. You’d need to have over a thousand times as many computers as there are atoms in the solar system!

One appealing solution to this problem is to *prove* critical software correct. The idea is that if we can write down in formal mathematical language what we’d like our

software to do, then we can try to prove that the code that actually runs behaves in the same way that the math specifies. Ideally, the specification will be relatively simple, and much easier to trust, than the millions of lines of code making up the software we use every day.

While proofs of algorithms tend to be done with pen-and-paper (consider the ubiquitous proofs that various sorting algorithms are correct in introductory algorithms classes), proofs of actual code are much harder. For reasons of performance, historical accident, interfacing with real users, or otherwise, code in the real-world tends to be much more complicated than code in an algorithms textbook. Proofs of code-correctness thus tend to be filled with tedious case-analysis with only sparse mathematical insights, and attempts to create and check these proofs by hand are subject to the same issues of human fallibility as writing the programs in the first place.

Enter machine-checked proofs. Foundational tools which allow users to write down theorems and proofs to be checked by machine free the proof author from error-prone case analysis; the computer will check that no cases were missed. Although you can never gain complete confidence in your software—the layer underneath may be untrustworthy in ways you cannot see [193], you can never prove that the mathematical system you’re working in is correct [164], and most proof-checkers are not yet themselves proven correct—machine-checked proofs allow a drastic reduction in how much code needs to be trusted. Rather than trusting the millions of lines of code of the software being verified, you only need to trust the specifications of the software, and the likely-much-smaller codebase of the proof checker. Furthermore, it’s very unlikely that bugs in the proof checker will overlap with bugs in the proof in just the right way to allow bugs only in one particular piece of software to slip through. Said another way, we no longer need to write tests for every piece of software separately; writing tests for the proof-checker helps eliminate bugs in *all* software proven by that checker.

We come now to the main thrust of this thesis. Tools for machine-checked proofs have already had many successes, in both software verification and traditional mathematics proof formalization, from compilers [120] to microkernels [105] to web servers [49] to cryptography code [70], from the Four-Color Theorem [77] to the Odd-Order Theorem [79] to Homotopy Type Theory [194]. While compiler performance—both the time it takes to compile code and the time it takes to run the generated code—has long been an active field of study [1, 74, 139][**TODO: find more citations here**], to our knowledge there is no extant body of work systematically investigating the performance of proof assistants, nor even any work primarily focused on the problem of proof assistant performance writ large. [**TODO: integrate stuff from Boulton [36]** such as “For example, much of the research in the field of automated reasoning is concerned with producing proofs more efficiently”] [**TODO: Related to this is work on sound interfaces to external tools, also an enduring topic: Slind [176]**]

[**TODO: work on this paragraph**] Why is this problem worthy of study? This thesis argues that the problem of *compile-time performance* or *proof-checking performance*

is both significantly different from the problem of performance of more typical programs in industry and that it is nontrivial. While many papers mention performance, obliquely or otherwise, and some are even driven by performance concerns of a particular algorithm or part of the system, [77, p. 1382, 36, 86, 38, 28, 158, 45, 147, 85, 141, 38], we have not found any that investigate the performance problems that arise *asymptotically*, when proof assistants are used to verify programs at large scale.

Unlike many other performance domains, in our experience, the time it takes to check proofs as we scale the size of the input is almost always super-linear—quadratic at best, commonly cubic or exponential, occasionally even worse. Empirically, this might look like a proof script that checks in tens of seconds on the smallest of toy examples; takes about a minute on the smallest real-world example, which might be twice the size of the toy example; takes twenty hours to generate and check the proof of an example only twice that size; and, on a (perhaps not-quite-realistic) example twice the size of the twenty-hour example, the script might not finish within a year, or even a thousand years—see Section 2.1 for more details. In just three doublings of input size, we might go from tens of seconds to thousands of years. Moreover, this is not an isolated experience in proof assistants; this sort of performance behavior is *typical*.

Drawing experience from case studies in category theory [89], parsers [87], and generation of low-level cryptographic code [70], we investigate and seek to explicate a broad swath of performance issues and bottlenecks. While we propose some design principles for avoiding these performance issues in Part II and present a research prototype of a tool and methodology for achieving acceptable performance at scale in Part III—and we claim both of these are original contributions of this PhD [TODO: how to write PhD in text? with periods? not?]—the story this thesis aims to tell is that in order for proof assistants to scale to industrial uses, *we must get the basics of asymptotic performance right*.

1.1.1 What are proof assistants?

Before diving into the details of performance bottlenecks and solutions, we review the history of formal verification and proof assistants to bring the reader up to speed on the context of our work and investigation.

[TODO: wide look at various proof assistants (historical overview of how the field got to where it is today, maybe with weight according to importance) and successes (applications of proof assistants like compcert, sel4 (kernel), cakeml, nqthm, ACL2, crypto???, flyspeck, 4-color-theorem, odd-order-theorem)] [TODO: <https://www.eeweb.com/profile/ada/brief-history-of-formal-verification>] [TODO: look at OSDI submission (papers/light-bulb/) and mine related work] [TODO: read and cite Talia’s paper Ringer et al. [166]?]

1.2 Basic Design Choices

[TODO: introduce section where we talk about what some big design decisions are and why we might make them the way we do]

1.2.1 Dependent Types: What? Why? How?

[TODO: note: target audience: Saman (or POPL/PLDI)] [TODO: explain what dependent types are at all] [TODO: explain the idea of reducing proof checking to typechecking (with examples)] [TODO: explain idea of structuring programming so that some properties are true by construction] [TODO: maybe explain extensional vs intensional] [TODO: maybe mention lego as early implementation of dependent types] [TODO: explain dependent type theory, motivate using it] [TODO: read and cite Lampert and Paulson [117] and Paulson [148] h/t Karl Palmskog @palmskog on gitter <https://gitter.im/coq/coq?at=5e5ec0ae4eefc06dcf31943f>]

1.2.2 The De Bruijn Criterion

[TODO: explain the De Bruijn criterion] [TODO: explain why it's worth following this, and what goes wrong if you don't] [TODO: look into Selsam, Hudon, and Moura [170] Sealing Pointer-Based Optimizations Behind Pure Functions] [TODO: example: Ltac vs Gallina, kind-of]

1.3 Look ahead: Layout and contributions of the thesis

In the remainder of Part I, we will finish laying out the landscape of performance bottlenecks in dependently-typed proof assistants; Chapter 2 (The Performance Landscape in Type-Theoretic Proof Assistants) gives a more in-depth investigation into what makes performance optimization in dependent type theory hard, different, and unique, followed by describing major axes of super-linear performance bottlenecks in Section 2.2 (The Four Axes).

Part II (API Design) is devoted, by and large, to the performance bottlenecks that arise from the use of dependent types as the basis of a proof assistant as introduced in Subsection 1.2.1; in Chapter 3 (Design-based fixes), we discuss lessons on engineering libraries at scale drawn from our case study in formalizing category theory and augmented by our other experience. The category theory library formalized as part of this doctoral work, available at *HoTT/HoTT Categories* [97], is described briefly in this chapter; a more thorough description can be found in the paper we published on our experience formalizing this library. [89]

Part III (Program Transformation and Rewriting) is devoted, in some sense, to performance bottlenecks that arise from the de Bruijn criterion of Subsection 1.2.2. We

investigate one particular method for avoiding these performance bottlenecks. We introduce this method, variously called *proof by reflection* or *reflective automation*, in Chapter 4 (Reflective Program Transformation), with a special emphasis on a particularly common use case—transformation of syntax trees. Chapter 5 (A Framework for Building Verified Partial Evaluators) describes our original contribution of a framework for leveraging reflection to perform rewriting and program transformation at scale, driven by our need to synthesize efficient, proven-correct, low-level cryptographic primitives. [70][**TODO:** cite also the rewriter paper] Where Chapter 5 addresses the performance challenges of verified or proof-producing program transformation, Chapter 6 (Engineering Challenges in the Rewriter) is a deep-dive into the performance challenges of engineering the tool itself, and serves as a sort-of microcosm of the performance bottlenecks previously discussed and the solutions we’ve proposed to them. Unlike the other chapters of this thesis, Chapter 6 at times assumes a great deal of familiarity with the details of the Coq proof assistant. Finally, Chapter 7 (Reification by Parametricity) presents a way to efficiently, elegantly, and easily perform *reification*, the first step of proof by reflection, which is often a bottleneck in its own right. We discovered—or invented—this trick in the course of working on our library for synthesis of low-level cryptographic primitives. [70, 90]

Part IV (Conclusion) is in some sense the mirror image of Part I: Where Chapter 2 is a broad look at what is currently lacking and where performance bottlenecks arise, Chapter 8 (A Retrospective on Performance Improvements) takes a historical perspective on what advancements have already been made in the performance of proof assistants, and Coq in particular. Finally, while the present chapter which we are now concluding has looked back on the present state and history of formal verification, Chapter 9 (Concluding Remarks) looks forward to what we believe are the most important next steps in the perhaps-nascent field of proof-assistant performance at scale.

1.A TODOs from Section 1.1

[**TODO:** read and cite Ebner et al. [69]] [**TODO:** read Shao, League, and Monnier [172]] “It’s not directly related to proof assistants, but the techniques described can be applicable to proof assistants and the experience *may* be applicable to some extent.” [**TODO:** read Brady, McBride, and McKinna [39]] [**TODO:** read Nogin [141]] [**TODO:** read Grégoire and Leroy’s paper from ICFP 2002 Grégoire and Leroy [85]] [**TODO:** read Dirk Kleeblatt’s PhD thesis Kleeblatt [104].] “Both of these are about using compiled code in dependent type checkers instead of interpreters.” [**TODO:** read András Kovács has smalltt at Kovács [113]]“, which I don’t think has been written up anywhere but has nonetheless been influential, both on Idris 2 and on Olle Fredriksson’s reimplementation of Sixten at <https://github.com/ollef/sixty>” [**TODO:** look at <https://github.com/AndrasKovacs/normalization-bench>] [**TODO:** look at <https://github.com/AndrasKovacs/smalltt>] [**TODO:** “I haven’t yet updated the smalltt repo, but there’s a simplified (<https://gist.github.com/AndrasKovacs/>

a0e0938113b193d6b9c1c0620d853784) implementation of its evaluator, which seems to have roughly the same performance but which is much simpler to implement.”]
[TODO: [150] about why not LCF tactics in dependently typed setting]

András Kovács wrote:

The basic idea is that in elaboration there are two primary computational tasks, one is conversion checking and the other is generating solutions for metavariables. Clearly, we should use NbE/environment machines for evaluation, and implement conversion checking in the semantic domain. However, when we want to generate meta solutions, we need to compute syntactic terms, and vanilla NbE domain only supports quote/readback to normal forms. Normal forms are way too big and terrible for this purpose. Hence, we extend vanilla NbE domain with lazy non-deterministic choice between two or more evaluation strategies. In the simplest case, the point of divergence is whether we unfold some class of definitions or not. Then, the conversion checking algorithm can choose to take the full-unfolding branch, and the quoting operation can choose to take the non-unfolding branch. At the same time, we have a great deal of shared computation between the two branches; we avoid recomputing many things if we choose to look at both branches.

I believe that a feature like this is absolutely necessary for robust performance. Otherwise, we choke on bad asymptotics, which is surprisingly common in dependent settings. In Agda and Coq, even something as trivial as elaborating a length-indexed vector expression has quadratic complexity in the length of the vector.

It is also extremely important to stick to the spirit of Coquand’s semantic checking algorithm as much as possible. In summary: core syntax should support *no* expensive computation: no substitution, shifting, renaming, or other ad-hoc term massaging. Core syntax should be viewed as immutable machine code, which supports evaluation into various semantic domains, from which sometimes we can read syntax back; this also leaves it open to swap out the representation of core syntax to efficient alternatives such as bytecode or machine code.

Only after we get the above two basic points right, can we start to think about more specific and esoteric optimizations. I am skeptical of proposed solutions which do not include these. Hash consing has been brought up many times, but it is very unsatisfying compared to non-deterministic NbE, because of its large constant costs, implementation complexity, and the failure to handle sharing loss from beta-redexes in any meaningful way (which is the most important source of sharing loss!). I am also skeptical of exotic evaluators such as interaction nets and optimal beta reducers; there is a good reason that all modern functional languages run on environment machines instead of interaction nets.

If we want to support type classes, then tabled instance resolution Selsam, Ullrich, and Moura [171] is also a must, otherwise we are again smothered by bad asymptotics even in modestly complex class hierarchies. This can be viewed as a specific instance of hash-consing (or rather “memoization”), so while I think ubiquitous hash-consing is bad, some focused usage can do good.

Injectivity analysis is another thing which I believe has large potential impact. By this I mean checking whether functions are injective up to definitional equality, which is decidable, and can be used to more precisely optimize unfolding in conversion checking.

I’d be very interested in your findings about proof assistant performance. This has been a topic that I’ve been working on on-and-off for several years. I’ve recently started to implement a system which I intend to be eventually “production strength” and also as fast as possible, and naturally I want to incorporate existing performance know-how.

[**TODO:** look into “So technically, the lost sharing is the second-order sharing that is preserved in “optimal reduction” of lambda calculi [Levy-1980, Lamping-1990, Asperti-Laneve-1992], while hash consing normally is directly usable only for first-order sharing.”] [**TODO:** look at <https://math.stackexchange.com/questions/3466976/online-reference-book-for-implementing-concepts-in-type-theory>] [**TODO:** look at <https://github.com/AndrasKovacs/elaboration-zoo/blob/0c7f8a676c0964cc05AIMpres/AIMpres.pdf> or https://eotypes.cs.ru.nl/eotypes_pmwiki/uploads/Meetings/Kovacs_slides.pdf] Konrad Slind wrote: [**TODO:** read Barras [25]] [**TODO:** Boulton [36]]

1.B Coq’s design

[**TODO:** where does this description of Coq’s design go?] Coq is split into two parts.

There’s the part of the system that is called the kernel or the trusted code base. Once you get a proof this part will be like “yup, I believe the proof” or like “nope your proof is bad.” And then there’s the other part that is like “here’s magic and it will make proof for you.” You’re like “I have an arithmetic expression please prove that it’s true” and there’s a bit in this other part that’s like “I know how to prove arithmetic expressions” and it gets the arithmetic expression and then it generates a certificate or proof that this other trusted part checks. If the part generating the arithmetic proof is wrong then the users come complaining to you that you have a bug. If the part checking the proofs is wrong, then you don’t see the bug and now suddenly your users can prove anything they want. And the system is no longer trustworthy right and so for that bit of it you need to be very careful with any changes you make.

[**TODO:** this chapter]

Chapter 2

The Performance Landscape in Type-Theoretic Proof Assistants

2.1 The Story

[TODO: come up with a better name for this section] The purpose of this chapter is to convince the reader that the issue of performance in proof assistants is non-trivial in ways that differ from performance bottlenecks in non-dependently-typed languages. I intend to do this by first sketching out what I see as the main difference between performance issues in dependently-typed proof assistants vs performance issues in other languages, and then supporting this claim with a palette of real performance issues that have arisen in Coq.

The widespread commonsense in performance engineering[commonsense-perf-engineering-order-of] is that good performance optimization happens in a particular order: there is no use micro-optimizing code if you are implementing an algorithm with unacceptable performance characteristics; imagine trying to optimize the pseudorandom number generator used in bogosort [bogosort], for example.¹ Similarly, there is no use trying to find or create a better algorithm if the problem you’re solving is more complicated than it needs to be; consider, for example, the difference between ray tracers and physics simulators. Ray tracers determine what objects can be seen from a given point essentially by drawing lines from the viewpoint to the object and seeing if it passes through any other object “in front of” it. Alternatively, one could provide a source of light waves and simulate the physical interaction of light with the various objects, to determine what images remain when the light arrives at a particular point. There’s no use trying to find an efficient algorithm for simulating quantum electrodynamics, though, if all you need to know is “which parts of which objects need to be drawn on the screen?”

¹Bogosort, whose name is a portmanteau of the words bogus and sort [bogosort-name], sorts a list by randomly permuting the list over and over until it is sorted.

One essential ingredient to allowing this division of concerns—between specifying the problem, picking an efficient algorithm, and optimizing the implementation of the algorithm—is knowledge of what a typical set of input looks like, and what the scope looks like. In Coq, and other dependently-typed proof assistants, this ingredient is missing. When sorting a list, we know that the length of the list and the initial ordering matter; for sorting algorithms that work for sorting lists with any type of elements, it generally doesn’t matter, though, whether we’re sorting a list of integers or colors or names. Furthermore, randomized datasets tend to be reasonably representative for list ordering, though we may also care about some special cases, such as already-sorted lists, nearly sorted lists, and lists in reverse-sorted order. We can say that sorting is always possible in $\mathcal{O}(n \log n)$ time, and that’s a pretty good starting point.

In proof assistants, the domain is much larger: in theory, we want to be able to check any proof anyone might write. Furthermore, in dependently typed proof assistants, the worst-case behavior is effectively unbounded, because any provably terminating computation can be run at typechecking time. [TODO: cite <https://github.com/coq/coq/issues/12200>]

In fact, this issue already arises for compilers of mainstream programming languages. [TODO: literature search on perf of compiletime in mainstream compilers?] The C++ language, for example, has `constexpr` constructions that allow running arbitrary computation at compile-time, and it’s well-known that C++ templates can incur a large compile-time performance overhead.[TODO: cite?] However, I claim that, in most languages, even as you scale your program, these performance issues are the exception rather than the rule. Most code written in C or C++ does not hit unbounded compile-time performance bottlenecks. Generally if you write code that compiles in a reasonable amount of time, as you scale up your codebase, your compile time will slowly creep up as well.

In Coq, however, the scaling story is very different. Frequently, users will cobble together code that works to prove a toy version of some theorem, or to verify a toy version of some program. By virtue of the fact that humans are impatient, the code will execute in reasonable time on the toy version. The user will then apply the same proof technique on a slightly larger example, and the proof-checking time will often be pretty similar. After scaling the input size a bit more, the proof-checking time will be noticeably slow—maybe it now takes a couple of minutes. Scaling the input just a tiny bit more, though, will result in the compiler not finishing even if you let it run for a day or more. This is what working in an exponential performance domain is like.

To put numbers on this, a project I was working on[TODO: cite fiat-crypto] involved generating C code to do arithmetic on very large numbers. The code generation was parameterized on the number of machine words needed to represent a single big integer. Our smallest toy example used two machine words; our largest—slightly



Figure 2-1: Timing of synthesizing subtraction

unrealistic—example used 17. The smallest toy example—two machine words—took about 14 seconds. Based on the the compile-time performance of about a hundred examples, we expect the largest example—17 machine words—would have taken over four thousand *millenia!* See Figure 2-1 and Figure 2-2. (Our primary non-toy test example used four machine words and took just under a minute; the biggest realistic example we were targeting was twice that size, at eight machine words, and took about 20 hours.)

Maybe, you might ask, were we generating unreasonable amounts of code? Each example using n machine words generated $3n$ lines of code. Furthermore, the actual code generation took less than 0.002% of the total time on the largest examples we tested (just 14 seconds out of about 211 hours). How can this be?

Our method involved two steps: first generate the code, then check that the generated code matches with what comes out of the verified code generator. This may seem a bit silly, but this is actually somewhat common; if you have a theorem that says “any code that comes out of this code generator satisfies this property”, you need a

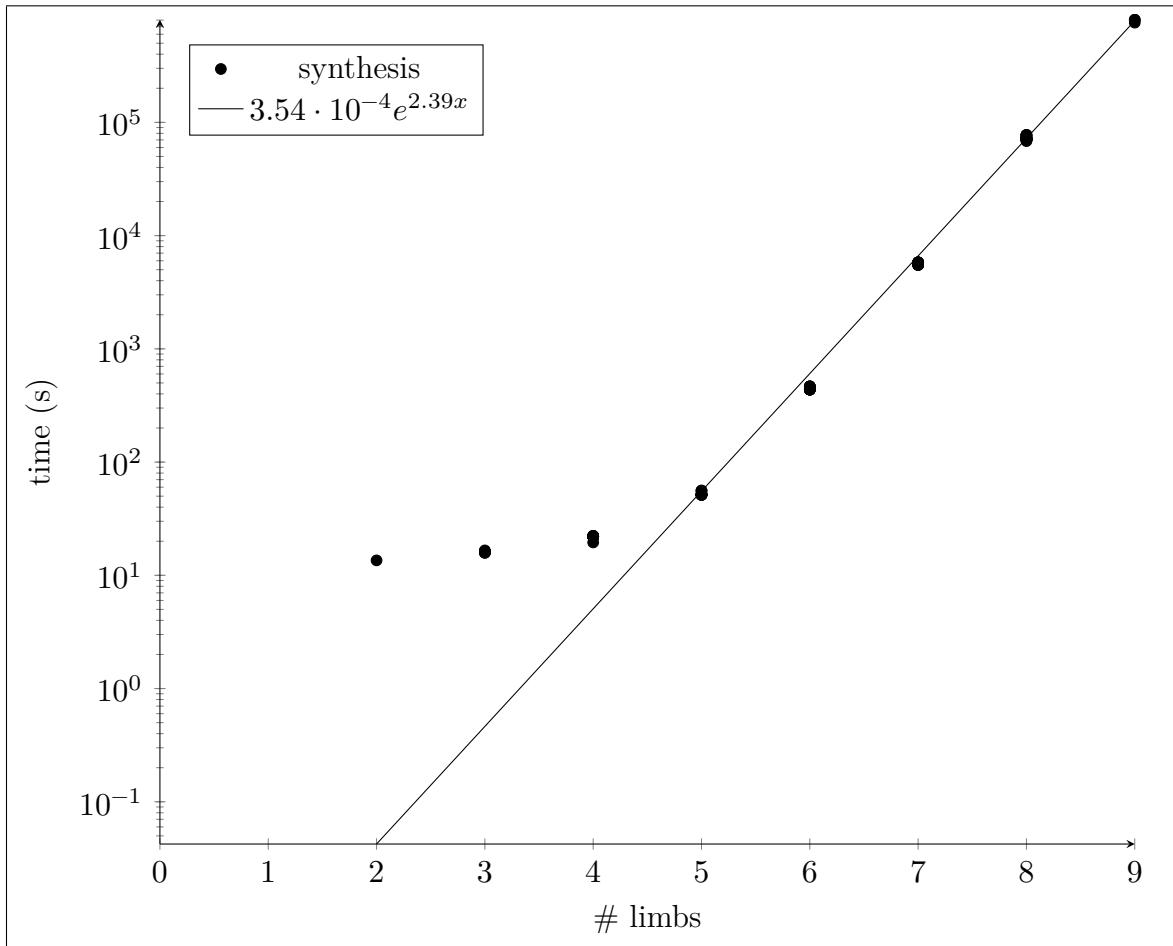


Figure 2-2: Timing of synthesizing subtraction (log-scale)

proof that the code you feed into the theorem actually came out of the specified code generator, and the easiest way to prove this is, roughly, to tell the proof assistant to just check that fact for you. (It's possible to be more careful and not do the work twice, but this often makes the code a bit harder to read and understand, and is oftentimes pointless; premature optimization is the root of all evil, as they say.) Furthermore, because you often don't want to fully compute results when checking that two things are equal—just imagine having to compute the factorial of 1000 just to check that $1000!$ is equal to itself—the default method for checking that the code came out of the code generator is different from the method we used to compute the code in the first place.

The fix itself is quite simple, only 21 characters long.² However, tracking down this solution was quite involved, requiring the following pieces:

1. A good profiling tool for proof scripts (see Subsection 8.1.3). This is a standard component of a performance engineer's toolkit, but when I started my PhD, there was no adequate profiling infrastructure for Coq. While such a tool is essential for performance engineering in all domains, what's unusual about dependently-typed proof assistants, I claim, is that essentially *every* codebase that needs to scale runs into performance issues, and furthermore these issues are frequently total blockers for development because so many of them are exponential in nature.
2. Understanding the details of how Coq works under-the-hood. Conversion, the ability to check if two types or terms are the same, is one of the core components of any dependently-typed proof assistant. Understanding the details of how conversion works is generally not something users of a proof assistant want to worry about; it's like asking C programmers to keep in mind the size of `gcc`'s maximum nesting level for `#include`'d files³ when writing basic programs. It's certainly something that advanced users need to be aware of, but it's not something that comes up frequently.
3. Being able to run the proof assistant in your head. When I looked at the conversion problem, I knew immediately what the most likely cause of the performance issue was. But this is because I've managed to internalize most of how Coq runs in my head. This might seem reasonable at a glance; one expects to have to understand the system being optimized in order to optimize it. But I've managed to learn the details of what Coq is doing—including performance characteristics—basically without having to read the source code at all! This is akin to, say, being able to learn how `gcc` represents various bits of C code, what transformations it does in what order, and what performance characteristics these transformations have, just from using `gcc` to compile C

²Strategy 1 [Let_In]. for those who are curious.

³It's 200, for those who are curious. [TODO: cite <https://gcc.gnu.org/onlinedocs/gcc-7.5.0/cpp/Implementation-limits.html>]

code and reading the error messages it gives you. These are details that should not need to be exposed to the user, but because dependent type theory is so complicated—complicated enough that it's generally assumed that users will get *line-by-line interactive feedback from the compiler* while developing, the numerous design decisions and seemingly reasonable defaults and heuristics lead to subtle performance issues. Note, furthermore, that this performance issue is essentially about the algorithm used to implement conversion, and is not even sensible when only talking about only the spec of what it means for two terms to be convertible. [TODO: incorporate Andres' suggestions:] you running the typechecker in your head is essentially the statement that if the entire implementation is part of the spec, it is possible to engineer better, and something close to this has been necessary in practice. the research direction you are advocating is finding a simpler performance-aware spec (perhaps by moving around interfaces or etc; my thought is that maybe we just want to get rid of the kernel and trust the proof engine).

4. Knowing how to tweak the built-in defaults for parts of the system which most users expect to be able to treat as black-boxes.

Note that even after this fix, the performance is *still* exponential! However, the performance is good enough that we deemed it not currently worth digging into the profile to understand the remaining bottlenecks. See Figure 2-3 and Figure 2-4.

[TODO: Some sort of summary of argument-so-far here]

To finish off the argument about slowness in dependently-typed proof assistants, I want to present four axes of performance bottlenecks. These axes are by no means exhaustive, but, in my experience, most interesting performance bottlenecks scale as a super-linear factor of one or more of these axes.

Misc Fragments

[TODO: Find a place for this (h/t conversation with Andres):] because we have a kernel and a proof engine on top of it, you need to simultaneously optimize the kernel and the proof engine to see performance improvements; if the kernel API doesn't give you good enough performance on primitives, then there's no hope to optimizing the proof engine, but at the same time if the proof engine is not optimized right, improvements in the performance of the kernel API don't have noticeable impact.

[TODO: find a place for this:] In many domains, good performance optimization can be done locally. It's rarely the case that disparate parts of the codebase must be simultaneously optimized to see any performance improvement. However, in proof assistants satisfying the de Bruijn criterion, there are many seemingly reasonable



Figure 2-3: Timing of synthesizing subtraction after fixing the bottleneck

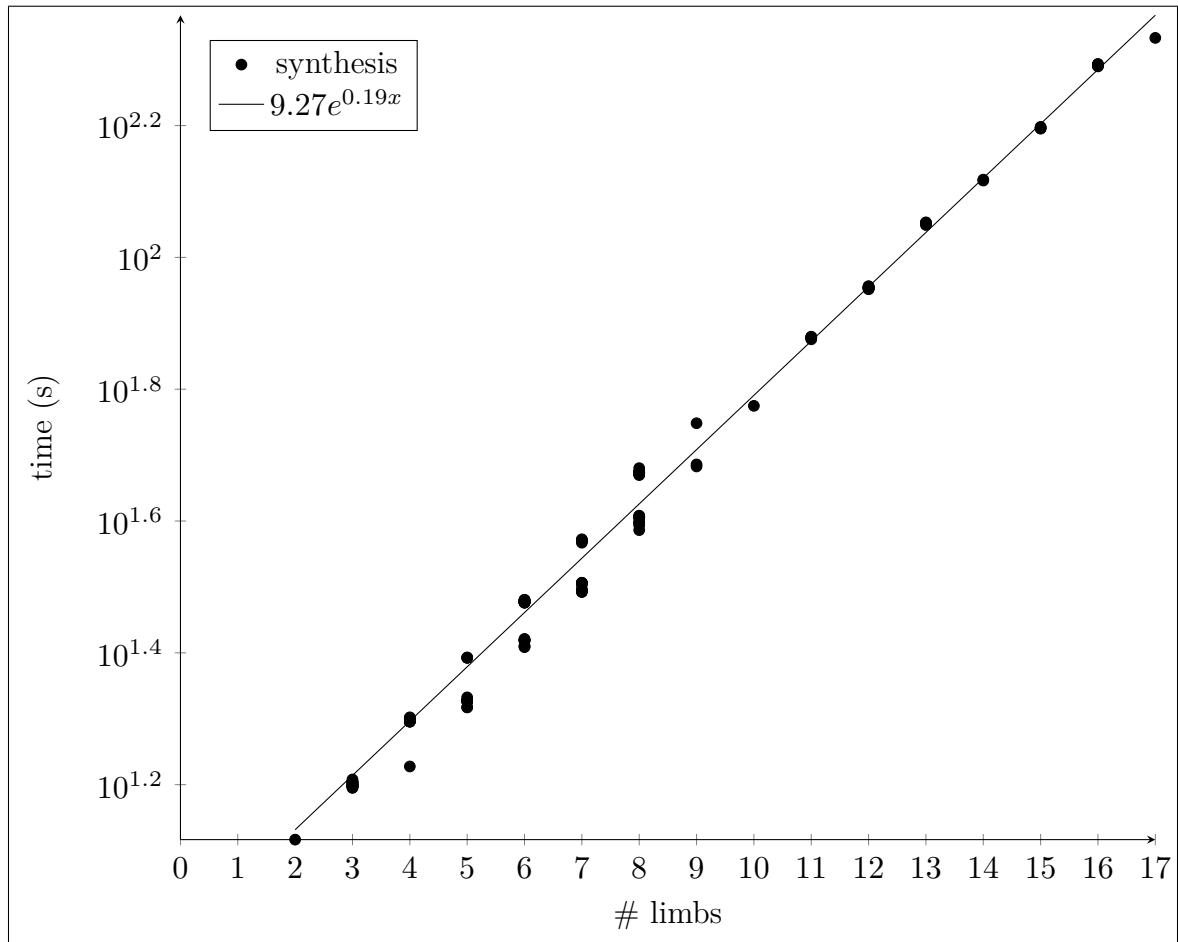


Figure 2-4: Timing of synthesizing subtraction after fixing the bottleneck (log-scale)

implementation choices that can be made for the kernel which make performance-optimizing the proof engine next to impossible. Worse, if performance optimization is done incrementally, to avoid needless premature optimization, then it can be the case that performance-optimizing the kernel has effectively no visible impact; the most efficient proof engine design for the slower kernel might be inefficient in ways that prevent optimizations in the kernel from showing up in actual use cases, because simple proof engine implementations tend to avoid the performance bottlenecks of the kernel while simultaneously shadowing them with bottlenecks with similar performance characteristics.

[**TODO:** incorporate Andres' suggestions] I like the last two sentences. I would instead lead with something along the lines of “in many domains, the performance challenges have been studied and understood, resulting in useful decompositions of the problem into subtasks that can be optimized independently.” “in proof assistants, it doesn’t look like anyone has even tried” :P. but e g signal processing was a huge mess too before the fast fourier transform. coq abstractions are mostly accidents of history. no other system has a clear performance-conscious story for how these interfaces should be designed either.

2.2 The Four Axes

I now present four major axes of performance. These are not comprehensive, but after extensive experience with Coq, most performance bottlenecks scaled super-linearly as a function of at least one of these axes. [**TODO:** introduce this section better]

2.2.1 The Size of the Type

We start with one of the simplest axes.

Suppose we want to prove a conjunction of n things, say, `True \wedge True $\wedge \dots \wedge$ True`. For such a simple theorem, we want the size of the proof, and the time- and memory-complexity of checking it, to be linear in n .

Recall from Subsection 1.2.2 that we want a separation between the small trusted part of the proof assistant and the larger untrusted part. The untrusted part generates certificates, which in dependently typed proof assistants are called terms, which the trusted part, the kernel, checks.

[**TODO:** Mention possibility of not building proof terms at all somewhere] Andrew Appel said via private correspondence on May 7, 2020, 7:39 PM:

There’s some work on typechecking LF, in the Twelf system, where there can be performance bottlenecks if you’re not careful.

The most glaringly obvious performance bottleneck in Coq is that it builds proof terms, when one should really use the futuristic technique of using data abstraction, in the type system, to distinguish “proposition” from “theorem”; as done in that state-of-the-art system, Edinburgh LCF. And presumably HOL, HOL light, Isabelle/HOL, etc.

The obvious certificate to prove a conjunction $A \wedge B$ is to hold a certificate a proving A and a certificate b proving B . In Coq, this certificate is called `conj` and it takes four parameters: A , B , $a : A$, and $b : B$. Perhaps you can already spot the problem.

To prove a conjunction of n things, we end up repeating the type n times in the certificate, resulting in a term that is quadratic in the size of the type. We see in Figure 2-5 the time it takes to do this in Coq’s tactic mode via `repeat constructor`. If we are careful to construct the certificate manually without duplicating work, we see that it takes linear time for Coq to build the certificate and quadratic time for Coq to check the certificate; see Figure 2-6. [TODO: improve data collection on perf test] [TODO: make a note about more complicated types causing scaling factors to be worse, and not just impacting the leaves]

Note that for small, and even medium-sized examples, it’s pretty reasonable to do duplicative work. It’s only when we reach very large examples that we start hitting non-linear behavior.

There are two obvious solutions for this problem:

1. We can drop the type parameters from the `conj` certificates.
2. We can implement some sort of sharing, where common subterms of the type only exist once in the representation.

Dropping Type Parameters: Nominal vs. Structural Typing

The first option requires that the proof assistant implement structural typing [structural-typing] rather than nominal typing [nominal-typing]. [TODO: Find a place for this note:] Note that it doesn’t actually require structural; we can do it with nominal typing if we enforce everywhere that we can only compare terms who are known to be the same type, because not having structural typing results in having a single kernel term with multiple non-unifiable types. [TODO: maybe look into TAPL] [TODO: explain structural and nominal typing more] Morally, the reason for this is that if we have an inductive record type [TODO: have we explained inductive types yet?] [TODO: have we explained records yet?] whose fields do not constrain the parameters of the inductive type family [TODO: have we explained parameters vs indices and inductive type families yet?], then we need to consider different instantiations of the same inductive type family to be convertible. That is, if we have a phantom record such as [TODO: mention where the name “phantom” comes from?]

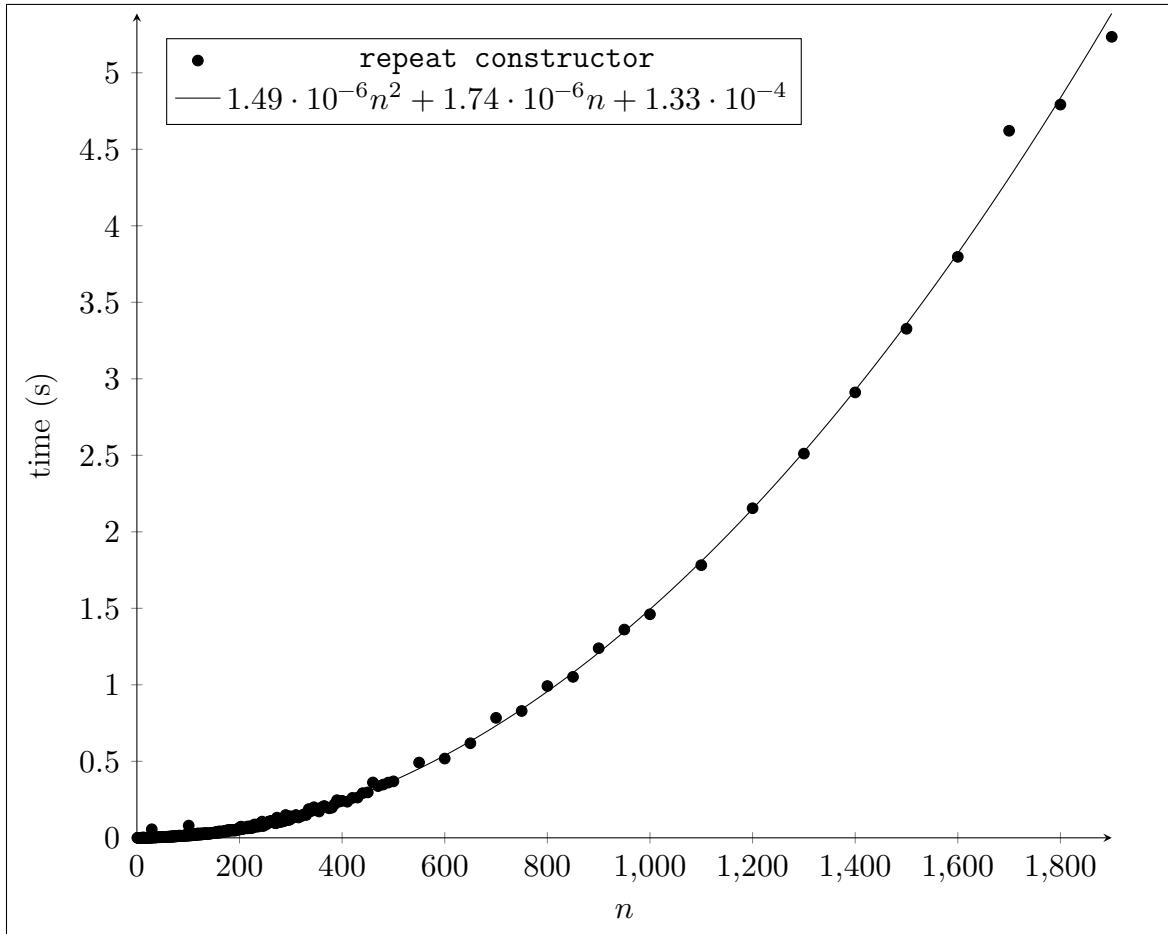


Figure 2-5: Timing of `repeat constructor` to prove a conjunction of n Trues

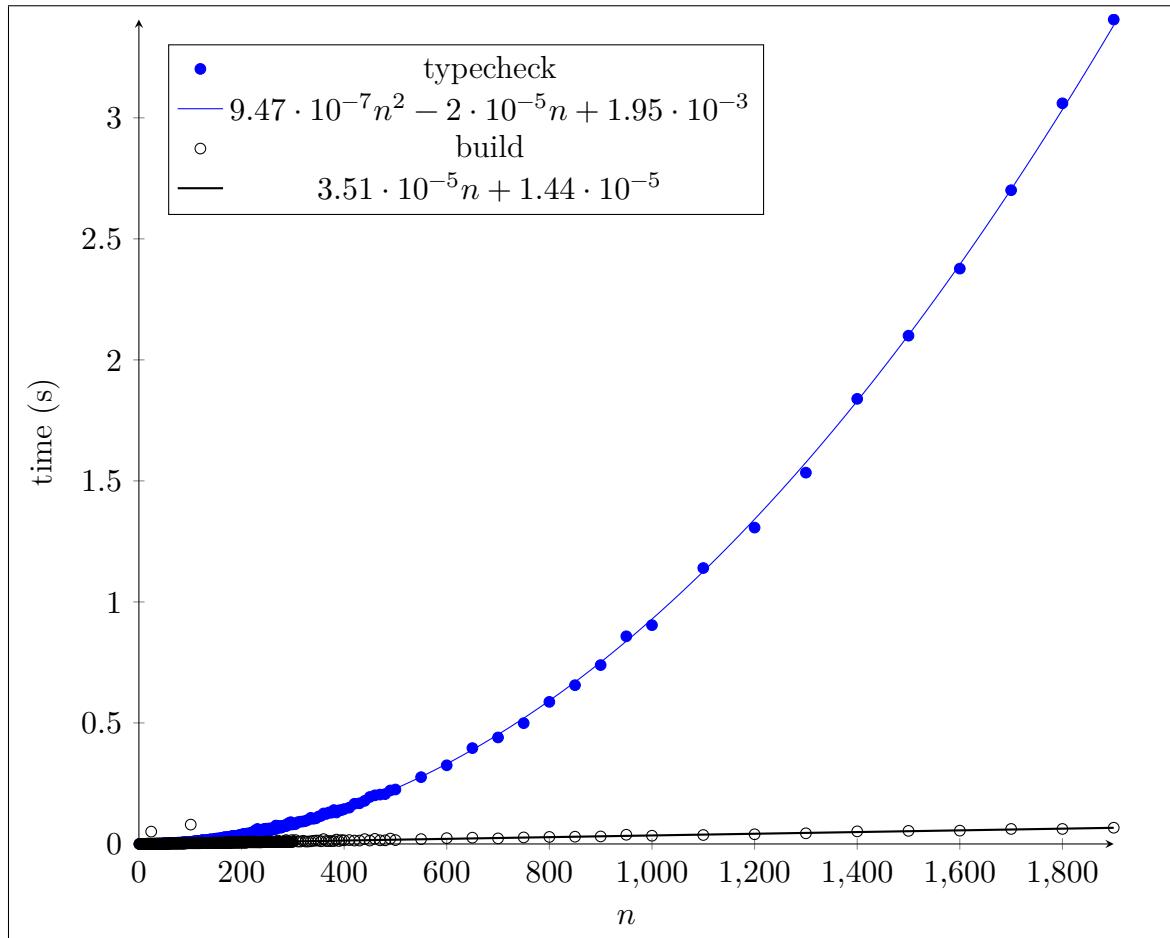


Figure 2-6: Timing of manually building and typechecking a certificate to prove a conjunction of n `Trues` using Ltac2

```
Record Phantom (A : Type) := phantom {}.
```

and our implementation does not include `A` as an argument to `phantom`, then we must consider `phantom` to be both of type `Phantom nat` and `Phantom bool`, even though `nat` and `bool` are not the same. I have requested this feature in [<https://github.com/coq/coq/issues>]. Note, however, that sometimes it is important for such phantom types to be considered distinct when doing type-level programming. [TODO: Come up with better justification for having nominal typing available?]

Sharing

The alternative to eliminating the duplicative arguments is to ensure that the duplication is at-most constant sized. There are two ways to do this: either the user can explicitly share subterms so that the size of the term is in fact linear in the size of the goal, or the proof assistant can ensure maximal sharing of subterms [TODO: explain this better].

There are two ways for the user to share subterms: using let-binders, and using function abstraction. For example, rather than writing

```
@conj True (and True (and True True)) I (@conj True (and True True) I (@conj True True))
```

and having roughly n^2 occurrences⁴ of `True` when we are trying to prove a conjunction of n `True`s, the user can instead write

```
let T0 : Prop := True in
let v0 : T0    := I in
let T1 : Prop := and True T0 in
let v1 : T1    := @conj True T0 I v0 in
let T2 : Prop := and True T1 in
let v2 : T2    := @conj True T1 I v0 in
@conj True T2 I v2
```

which has only n occurrences of `True`. Alternatively, the user can write

```
(λ (T0 : Prop) (v0 : T0),
  (λ (T1 : Prop) (v1 : T1),
    (λ (T2 : Prop) (v2 : T2), @conj True T2 I v2)
      (and True T1) (@conj True T1 I v1))
    (and True T0) (@conj True T0 I v0)))
True I
```

⁴The exact count is $n(n + 1)/2 - 1$.

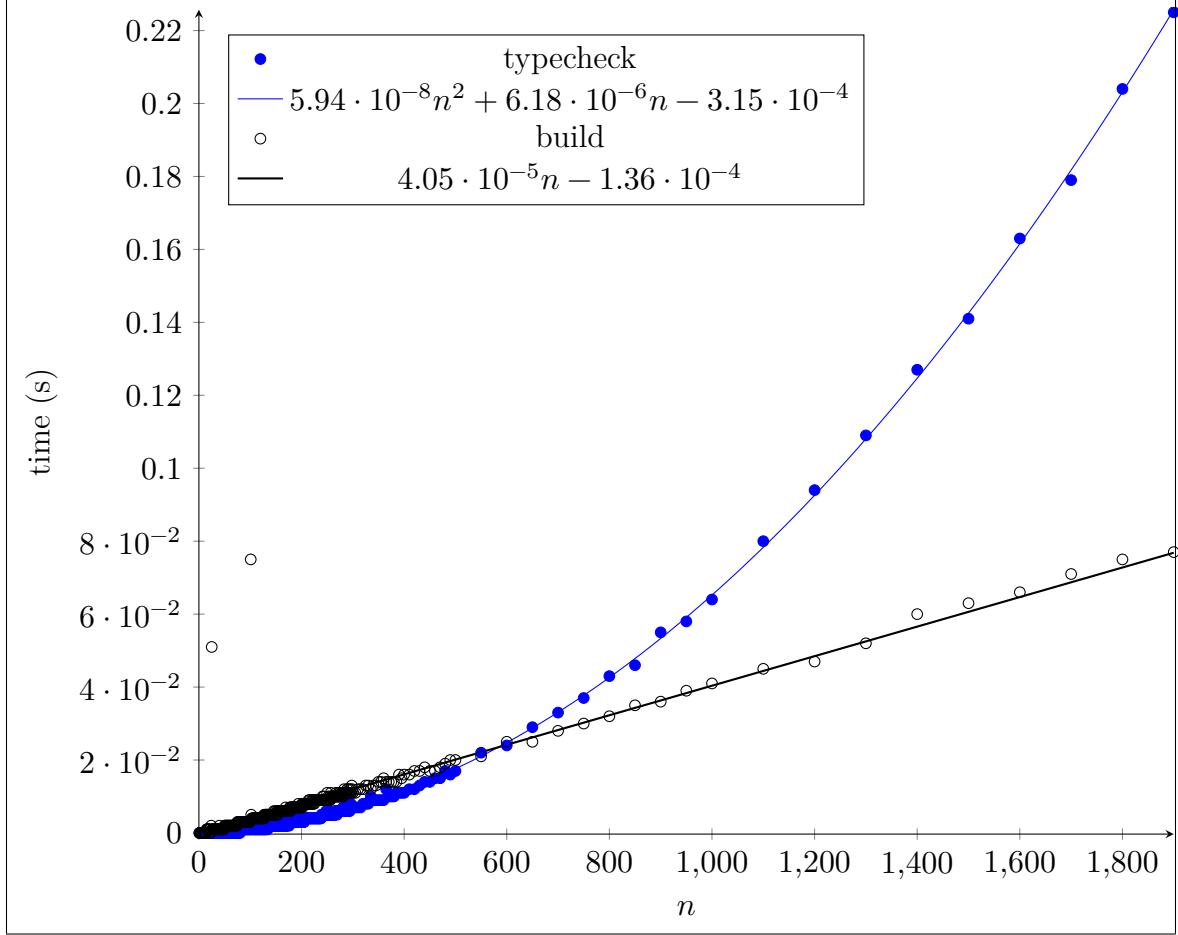


Figure 2-7: Timing of manually building and typechecking a certificate to prove a conjunction of n `True`s using `let`-binders using Ltac2 [TODO: find pdf of manual for styling]

Unfortunately, both of these incur quadratic typechecking cost, even though the size of the term is linear. See Figure 2-7 and Figure 2-8. [TODO: improve data collection on perf test]

Recall that the typing rules for λ and `let` are as follows: [TODO: cite appendix with typing rules of Coq?] [TODO: maybe look in <https://github.com/achlipala/frap> and/or TAPL by Benjamin C. Pierce for how to render typing rules] [QUESTION FOR ADAM: What's the suggested way of pretty-printing typing rules?] [QUESTION FOR ADAM: Which way do substitution brackets go?] [QUESTION FOR ADAM: What convention should we use for typing rules with regard to things being types? Maybe just copy the HoTT book?]

$$\begin{array}{c} \Gamma \vdash A \text{ type} \quad \Gamma, x:A \vdash B \text{ type} \\ \Gamma, x:A \vdash f:B \end{array}$$

$$\Gamma \vdash (\lambda (x:A), f) : \forall x:A, B$$

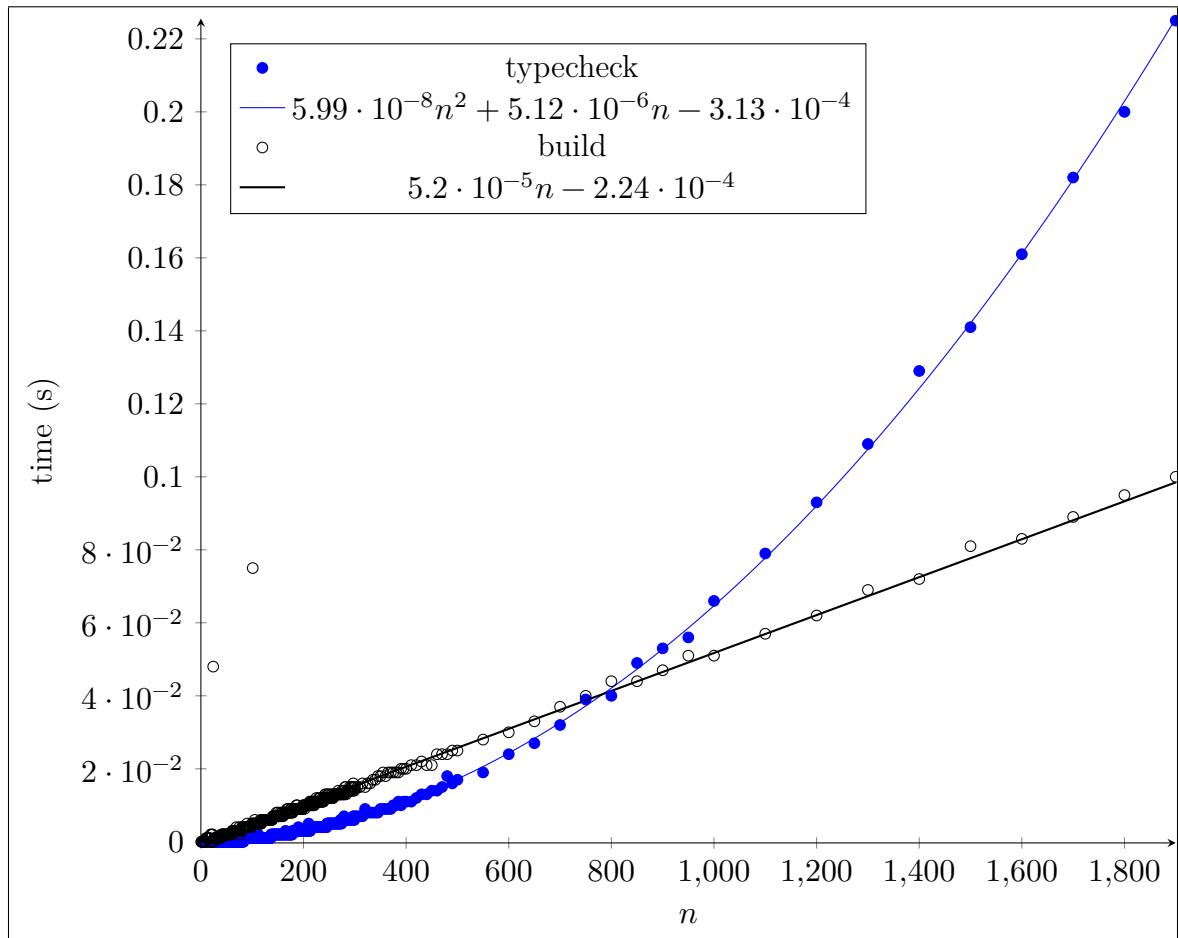


Figure 2-8: Timing of manually building and typechecking a certificate to prove a conjunction of n `Trues` using abstraction and application using Ltac2

$$\begin{array}{c} \Gamma \vdash A \text{ type} \quad \Gamma, x:A \vdash B \text{ type} \\ \Gamma \vdash f : \forall x:A, B \\ \Gamma \vdash y : A \end{array}$$

$$\Gamma \vdash f y : B[x/y]$$

$$\begin{array}{c} \Gamma \vdash A \text{ type} \\ \Gamma \vdash y : A \\ \Gamma, x:A := y \vdash B \text{ type} \\ \Gamma, x:A := y \vdash f : B \end{array}$$

$$\Gamma \vdash (\text{let } x : A := y \text{ in } f) : B[x/y]$$

Let us consider the inferred types for the intermediate terms when typechecking the **let** expression:

- We infer the type `and True T2` for the expression

```
@conj True T2 I v2
```

- We perform the no-op substitution of `v2` into that type to type the expression

```
let v2 : T2 := @conj True T1 I v0 in
@conj True T2 I v2
```

- We substitute `T2 := and True T1` into this type to get the type `and True (and True T1)` for the expression

```
let T2 : Prop := and True T1 in
let v2 : T2 := @conj True T1 I v0 in
@conj True T2 I v2
```

- We perform the no-op substitution of `v1` into this type to get the type for the expression

```
let v1 : T1 := @conj True T0 I v0 in
let T2 : Prop := and True T1 in
let v2 : T2 := @conj True T1 I v0 in
@conj True T2 I v2
```

- We substitute `T1 := and True T0` into this type to get the type `and True (and True T0)` for the expression

```
let T1 : Prop := and True T0 in
let v1 : T1 := @conj True T0 I v0 in
```

```

let T2 : Prop := and True T1 in
let v2 : T2    := @conj True T1 I v0 in
@conj True T2 I v2

```

- We perform the no-op substitution of `v0` into this type to get the type for the expression

```

let v0 : T0    := I in
let T1 : Prop := and True T0 in
let v1 : T1    := @conj True T0 I v0 in
let T2 : Prop := and True T1 in
let v2 : T2    := @conj True T1 I v0 in
@conj True T2 I v2

```

- Finally, we substitute `T0 := True` into this type to get the type `and True (and True True)` for the expression

```

let T0 : Prop := True in
let v0 : T0    := I in
let T1 : Prop := and True T0 in
let v1 : T1    := @conj True T0 I v0 in
let T2 : Prop := and True T1 in
let v2 : T2    := @conj True T1 I v0 in
@conj True T2 I v2

```

Note that we have performed linearly many substitutions into linearly-sized types, so unless substitution is constant time in size of the term being substituted, we incur quadratic overhead here. The story for function abstraction is similar. [TODO: cite <https://github.com/coq/coq/issues/8232> maybe?]

[TODO: Should we run though typechecking in more detail here?]

We again have two choices to fix this: either we can change the typechecking rules (which work just fine for small-to-medium-sized terms), or we can adjust typechecking to deal with some sort of pending substitution data, so that we only do substitution once.

[TODO: maybe cite <https://github.com/coq/coq/issues/11838>?]

[TODO: reference quadratic cbv here, which had a similar issue?]

[TODO: some sort of section division marker here?]

The proof assistant can also try to heuristically share subterms for us. Many proof assistants do some version of this, called *hashconsing*. [TODO: explain and cite hashconsing?]

However, hashconsing loses a lot of its benefit if terms are not maximally shared (and they almost never are), and can lead to very unpredictable performance when transformations unexpectedly cause a loss of sharing. [TODO: cite hashconsing needing to be full to get perf benefit] Furthermore, it's an open problem how to efficiently persist full hashconsing to disk in a way that allows for diamond dependencies. [TODO: explain this more, find citation for hashconsing being hard with disk] [TODO: flesh out hashconsing section more] [TODO: maybe cite <https://github.com/coq/coq/issues/9028#issuecomment-600013284> about hashconsing being slow]

2.2.2 The Size of the Term

Recall that Coq (and dependently typed proof assistants in general) have *terms* which serve as both programs and proofs. The essential function of a proof checker is to verify that a given term has a given type. We obviously cannot type-check a term in better than linear time in the size of the representation of the term.

Recall that we cannot place any hard bounds on complexity of typechecking a term, as terms as simple as `@eq_refl bool true` proving that the boolean `true` is equal to itself can also be typechecked as proofs of arbitrarily complex decision procedures returning success.

We might reasonably hope that typechecking problems which require no interesting computation can be completed in time linear in the size of the term and its type.

However, some seemingly reasonable decisions can result in typechecking taking quadratic time in the size of the term, as we saw in Section 2.2.1.

[TODO: maybe move some text from the sharing section to here?]

Even worse, typechecking can easily be unboundedly large in the size of the term when the typechecker chooses the wrong constants to unfold, even when very little work ought to be done.

[TODO: discussion of conversion checking, and conversion modulo delta-beta]

Consider the problem of typechecking `@eq_refl nat (fact 100) : @id nat (fact 100) = fact 100`, where `fact` is the factorial function on natural numbers and `id` is the polymorphic identity function. [TODO: should we define polymorphic identity function somewhere?] If the typechecker either decides to unfold `id` before unfolding `fact`, or if it performs a breath-first search, then we get speedy performance. However, if the typechecker instead unfolds `id` *last*, then we end up computing the normal form of $100!$, which takes a long time and a lot of memory. See Figure 2-9.

Note that it is by no means obvious that the typechecker can meaningfully do anything about this. Breath-first search is significantly more complicated than depth-first, is

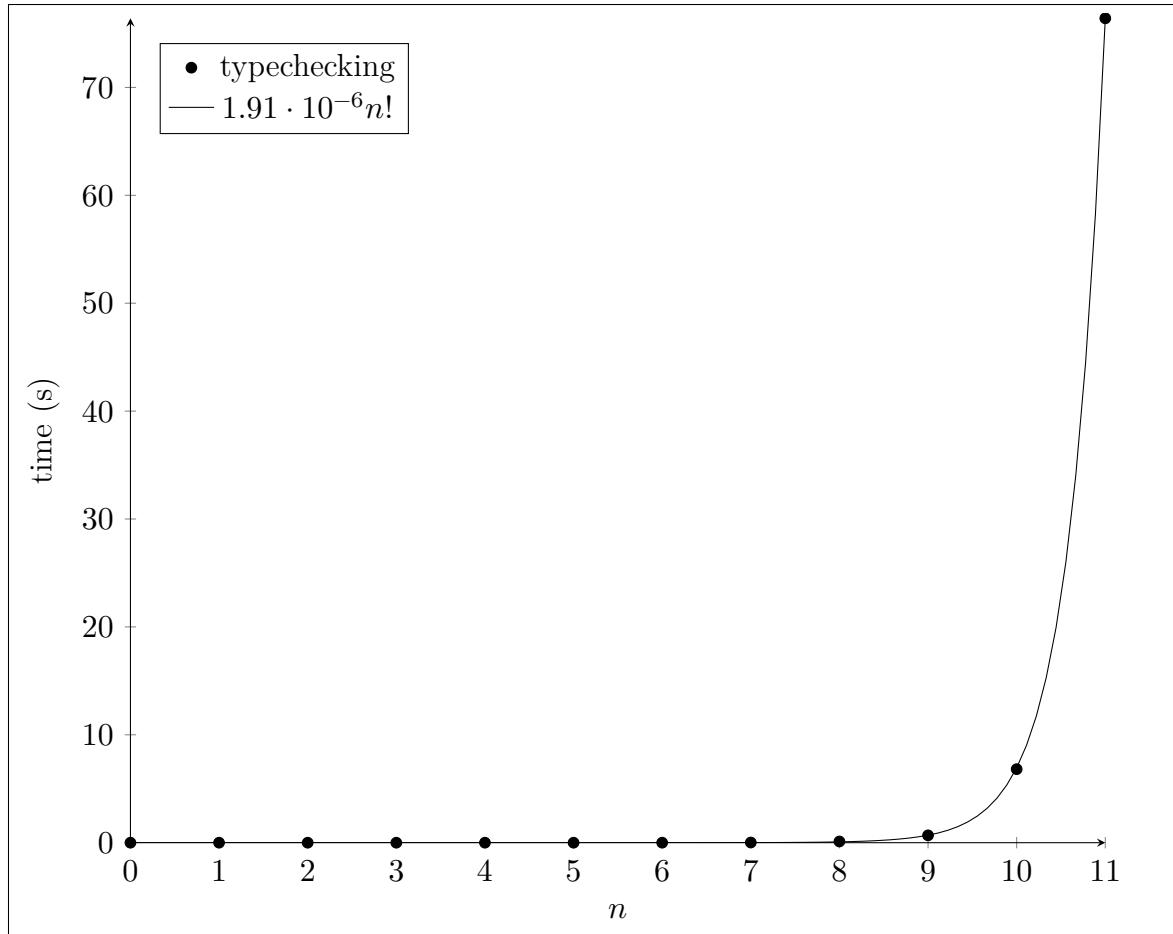


Figure 2-9: Timing of typechecking `@eq_refl nat (fact n) : @id nat (fact n) = fact n`

harder to write good heuristics for, can incur enormous space overheads, and can be massively slower in cases where there are many options and the standard heuristics for depth-first unfolding in conversion-checking are sufficient. Furthermore, the more heuristics there are to tune conversion-checking, the more “magic” the algorithm seems, and the harder it is to debug when the performance is inadequate.

As described in [TODO: reference above example about fiat-crypto], in fiat-crypto, we got exponential slowdown due to this issue, with an estimated overhead of over four millenia of extra typechecking time in the worst examples we were trying to handle.

[TODO: maybe forward reference to the number of abstraction barriers] [TODO: maybe include more about real-world fiat-crypto example here?]

2.2.3 The Number of Binders

This is a particular subcase of the above sections that we call out explicitly. Often there will be some operation (for example, substitution, lifting, context-creation) that needs to happen every time there is a binder, and which, when done naively, is linear in the size of the term or the size of the context. As a result, naïve implementations will often incur quadratic—or worse—overhead in the number of binders.

[TODO: make sure we’ve explained proof engine and Ltac by here]

Similarly, if there is any operation that is even linear rather than constant in the number of binders in the context, then any user operation in proof mode which must be done, say, for each hypothesis, will incur an overall quadratic-or-worse performance penalty.

The claim of this subsection is not that any particular application is inherently constrained by a performance bottleneck in the number of binders, but instead that it’s very, very easy to end up with quadratic-or-worse performance in the number of binders, and hence that this forms a meaningful cluster for performance bottlenecks in practice.

I will attempt to demonstrate this point with a palette of actual historical performance issues in Coq—some of which persist to this day—where the relevant axis was “number of binders.” None of these performance issues are insurmountable, but all of them are either a result of seemingly reasonable decisions, have subtle interplay with seemingly disparate parts of the system, or else are to this day still mysterious despite the work of developers to investigate them.

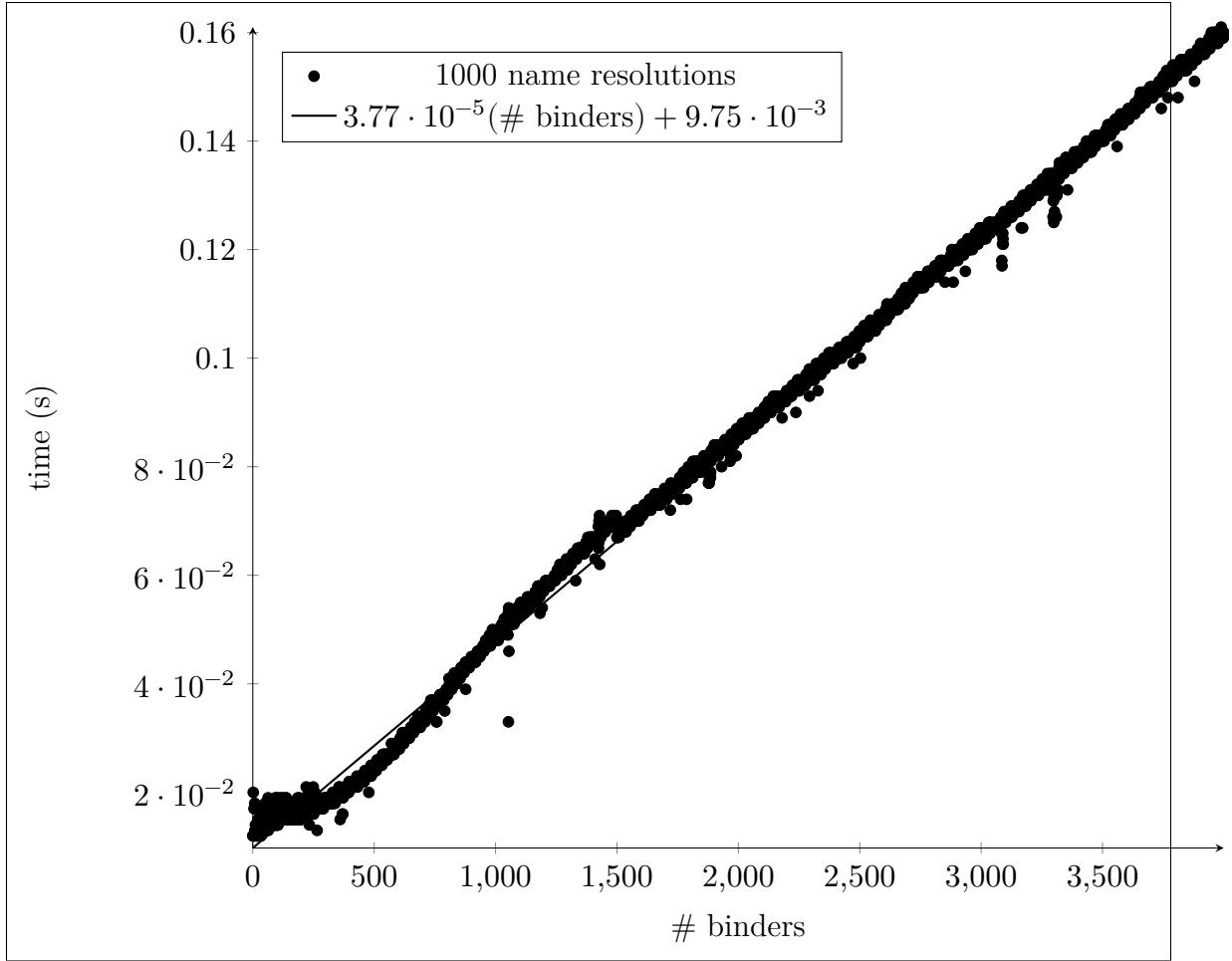


Figure 2-10: Timing of internalizing a name 1000 times under n binders

Name Resolution

One key component of interactive proof assistants is figuring out which constant is referred to by a given name. It may be tempting to keep the context in an array or linked list. However, if looking up which constant or variable is referred to by a name is $\mathcal{O}(n)$, then internalizing a term with n typed binders is going to be $\mathcal{O}(n^2)$, because we need to do name lookups for each binder. See #9582 and #9586.

See Figure 2-10 for the timing of name resolution in Coq. See Figure 2-11 for the effect on internalizing a lambda with n arguments.

[TODO: mention Coq version automatically, mention why we're using a different Coq version]

[TODO: Is this called internalization or is it called elaboration]

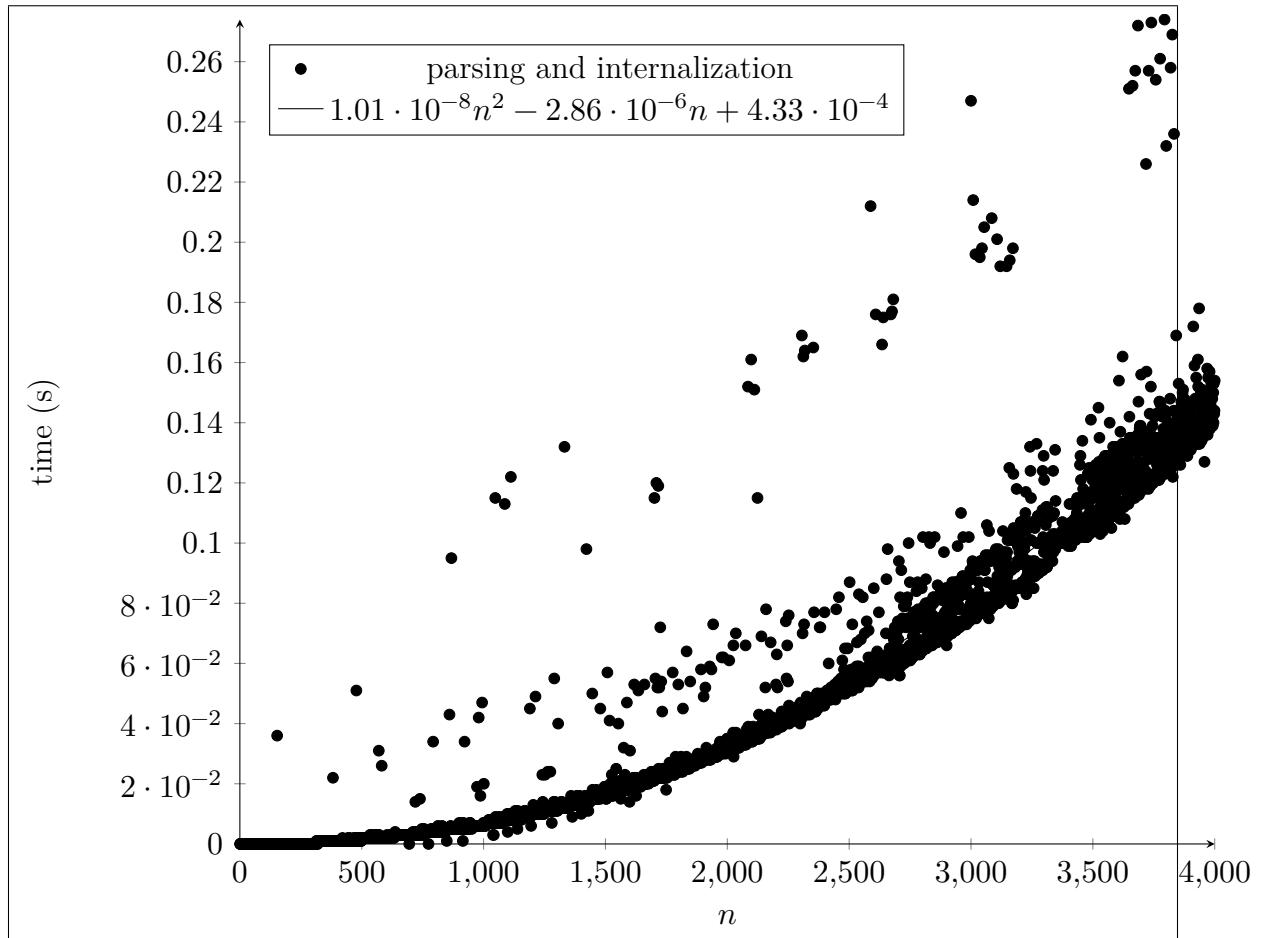


Figure 2-11: Timing of internalizing a function with n differently-named arguments of type `True`

Capture-Avoiding Substitution

If the user is presented with a proof engine interface where all context variables are named, then in general the proof engine must implement capture-avoiding substitution. For example, if the user wants to operate inside the hole in $(\lambda x, \text{let } y := x \text{ in } \lambda x, _)$, then the user needs to be able to talk about the body of y , which is not the same as the innermost x . However, if the α -renaming is even just linear in the existing context, then creating a new hole under n binders will take $\mathcal{O}(n^2)$ time in the worst case, as we may have to do n renamings, each of which take time $\mathcal{O}(n)$. See #9582, perhaps also #8245 and #8237 and #8231.

This might be the cause of the difference in Figure 2-13 between having different names (which do not need to be renamed) and having either no name (requiring name generation) or having all binders with the same name (requiring renaming in evar substitutions).

[**TODO:** ADD PLOT: try to come up with a graph for renaming stuff] [**TODO:** ADD PLOT: check if confounders come up]

Quadratic Creation of Substitutions for Existential Variables

Recall [**TODO:** make sure that this is mentioned previously, and that we're not rehashing things too much] that when we separate the trusted kernel from the untrusted proof engine, we want to be able to represent not-yet-finished terms in the proof engine. The standard way to do this is to enrich the type of terms with an “existential variable” node, which stands for a term which will be filled later. [**TODO:** cite original idea for evars? (what is it?)] Such existential variables, or evars, typically exist in a particular context. That is, you have access to some hypotheses but not others when filling an evar.

Sometimes, reduction results in changing the context in which an evar exists. For example, if we want to β -reduce $(\lambda x, ?e_1) (S y)$, then the result is the evar $?e_1$ with $S y$ substituted for x .

There are a number of ways to represent substitution, and the choices are entangled with the choices of term representation.

Note that most substitutions are either identity or lifting substitutions. [**TODO:** define identity and lifting substitutions]

One popular representation is the locally nameless representation. [**locally-nameless**] [**TODO:** Justify it? Discuss other representations? Say why and how it's convenient?] However, if we use a locally nameless term representation, then finding a compact representation for identity and lifting substitutions is quite tricky. If the substitution representation takes $\mathcal{O}(n)$ time to create in a context of size n , then having a λ with n

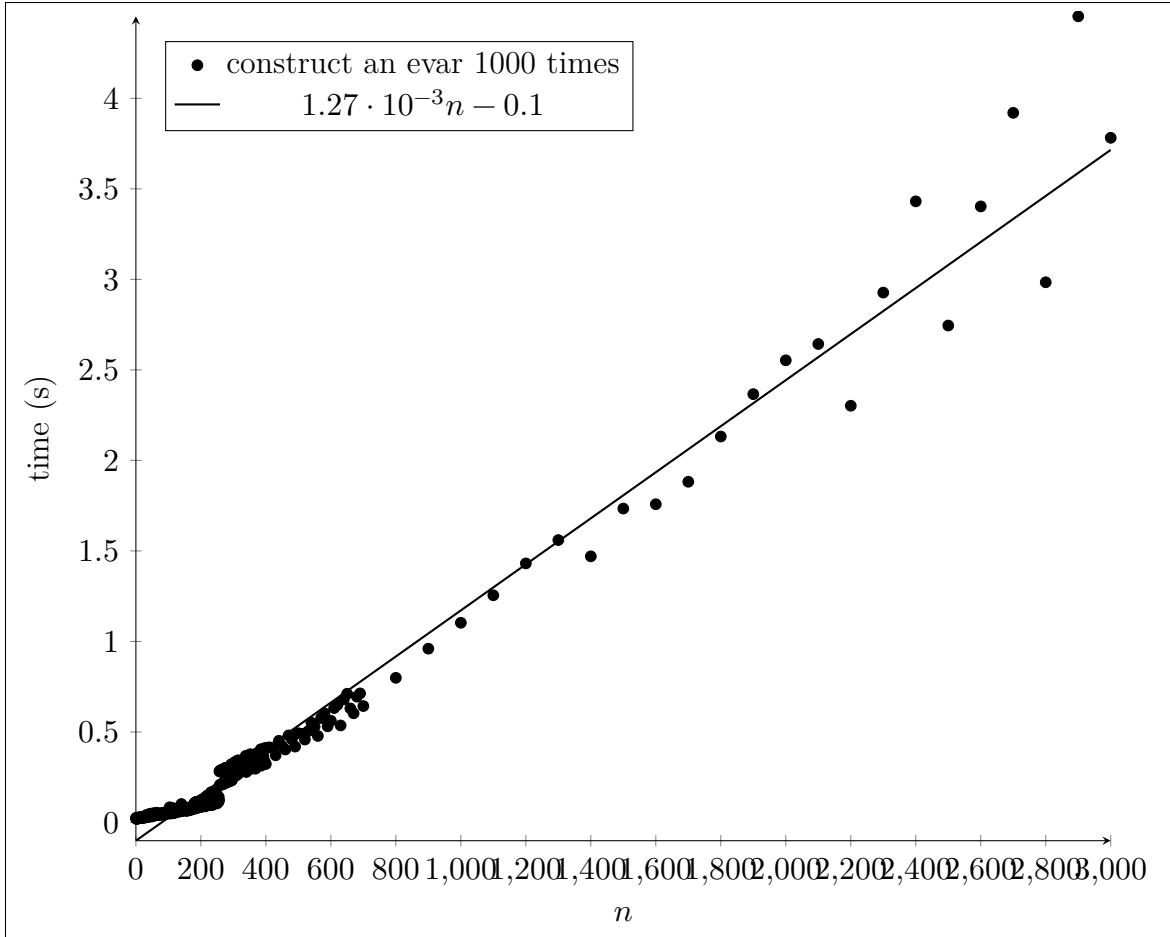


Figure 2-12: Timing of generating 1000 evars in a context of size n

arguments whose types are not known takes $\mathcal{O}(n^2)$ time, because we end up creating identity substitutions for n holes, with linear-sized contexts.

Note that fully nameless, i.e., de Bruijn term representations, do not suffer from this issue.

See #8237 and #11896 for a mitigation of some (but not all) issues.

See also Figure 2-12 and Figure 2-13.

Quadratic Substitution in Function Application

Consider the case of typechecking a non-dependent function applied to n arguments. If substitution is performed eagerly, following directly the rules of the type theory, [TODO: cite the rules / reference an appendix], then typechecking is quadratic. This is because the type of the function is $\mathcal{O}(n)$, and doing substitution n times on a term of size $\mathcal{O}(n)$ is quadratic.

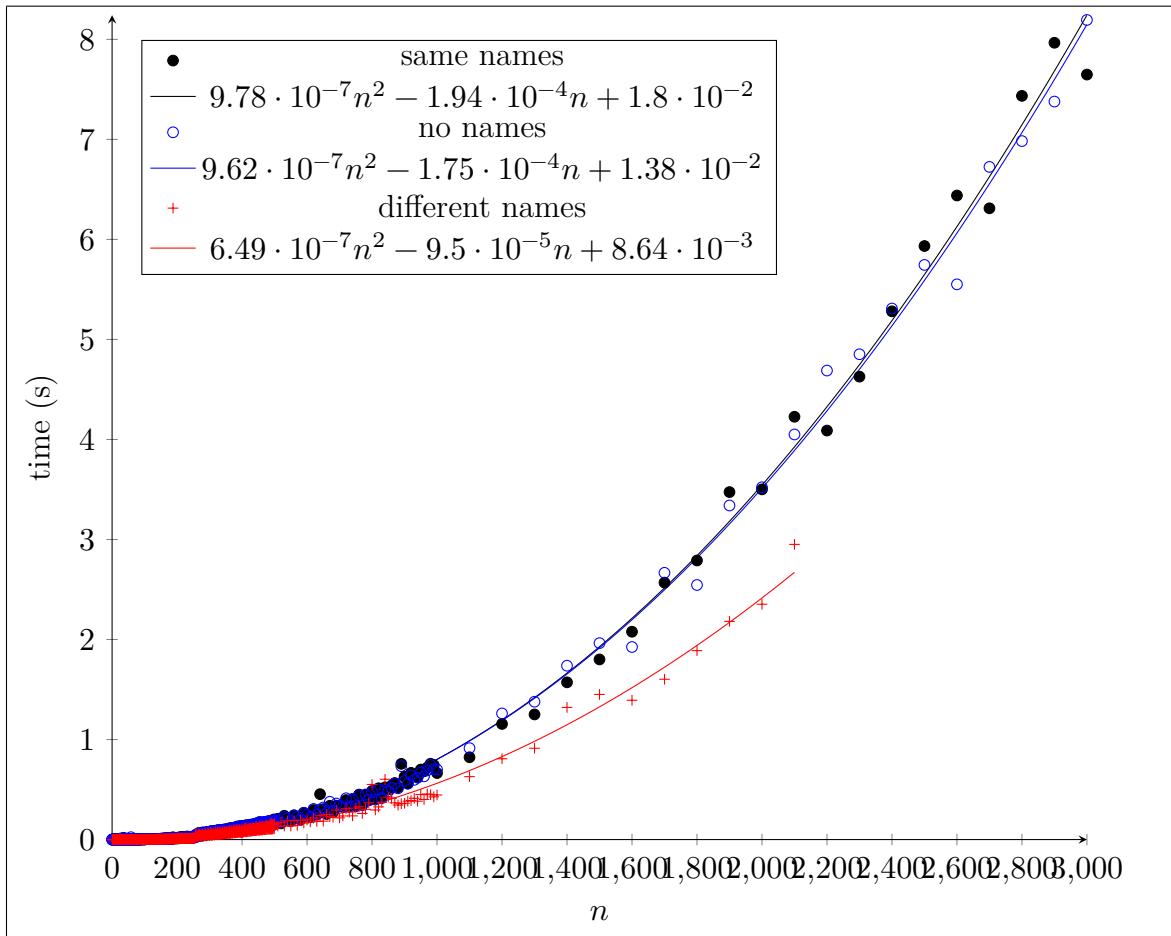


Figure 2-13: Timing of generating a λ with n binders of unknown/evar type, all of which have either no name, the same name, or different names

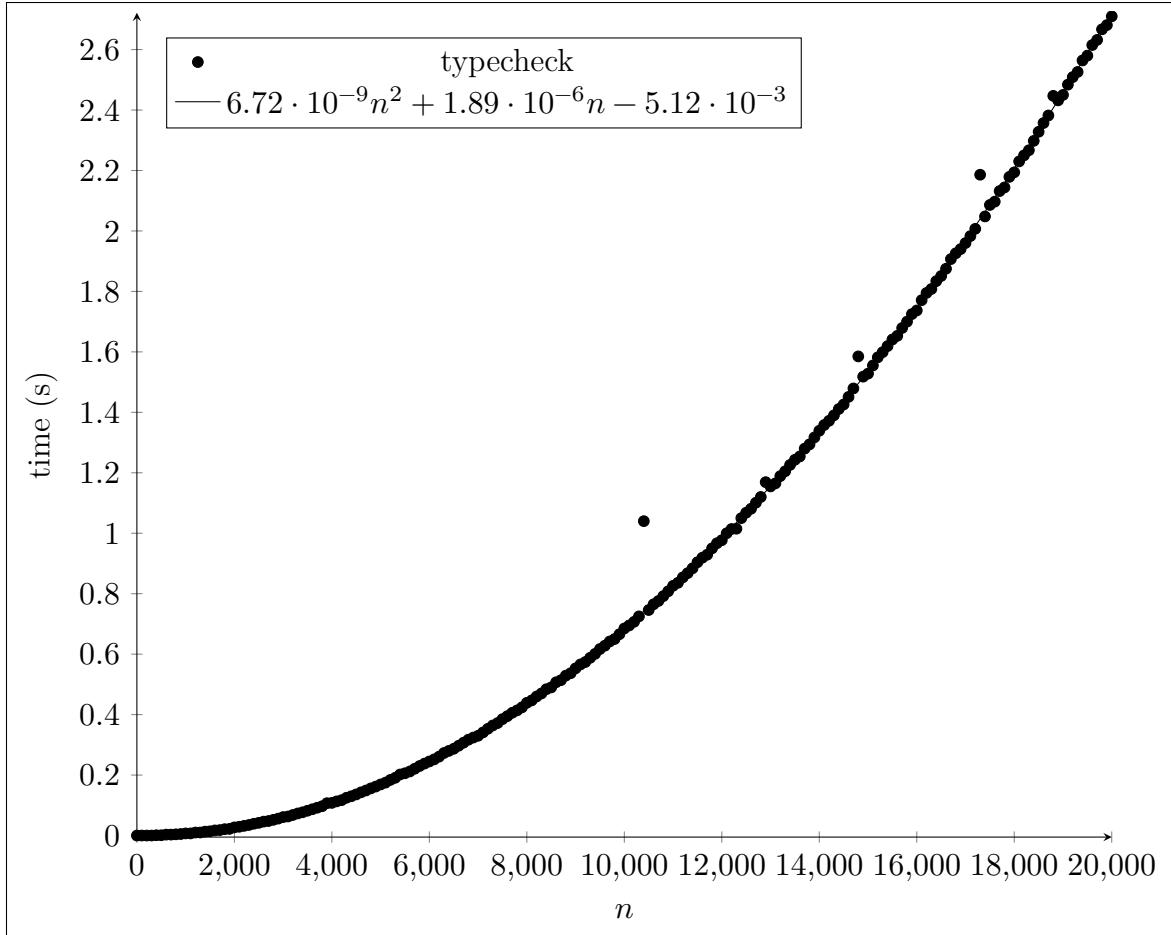


Figure 2-14: Timing of typechecking a function applied to n arguments

If the term representation contains n -ary application nodes, it's possible to resolve this performance bottleneck by delaying the substitutions. If only unary application nodes exist, it's much harder to solve.

Note that this is important, for example, if you want to avoid the problem of quadratically-sized certificates by making a n -ary conjunction-constructor which is parameterized on a list of the conjuncts. Such a function could then be applied to the n proofs of the conjuncts.

See #8232 and #12118 and #8255.

See Figure 2-14.

Quadratic Normalization by Evaluation

Normalization by evaluation (NbE) is a nifty way to implement reduction where function abstraction in the object language is represented by function abstraction in the metalanguage. [TODO: explain or forward-reference NbE] [TODO: Read stuff in

<https://github.com/HoTT/book/issues/995#issuecomment-418825844> for NbE stuff] [TODO: Say more about it than that it's "nifty"] Coq uses NbE to implement two of its reduction machines (`lazy` and `c bv`).

The details of implementing NbE depend on the term representation used. If a fancy term encoding like PHOAS [TODO: reference explanation of PHOAS] is used, then it's not hard to implement a good NbE algorithm. However, such fancy term representations incur unpredictable and hard-to-deal-with performance costs. Most languages do not do any reduction on thunks until they are called with arguments, which means that forcing early reduction of a PHOAS-like term representation requires round-tripping through another term representation, which can be costly on large terms if there is not much to reduce. On the other hand, other term representations need to implement either capture-avoiding substitution (for named representations) or index lifting (for de Bruijn and locally nameless representations).

The sort-of obvious way to implement this transformation is to write a function that takes a term and a binder, and either renames the binder for capture-avoiding substitution or else lifts the indices of the term. The problem with this implementation is that if you call it every time you move a term under a binder, then moving a term under n binders traverses the term n times. If the term size is also proportional to n , then the result is quadratic blowup in the number of binders.

See #11151 for an occurrence of this performance issue in the wild in Coq. See also Figure 2-15.

Quadratic Closure Compilation

It's important to be able to perform reduction of terms in an optimized way. [TODO: maybe give a better introductory justification to the vm and native compiler than "it's important"?] When doing optimized reduction in an imperative language, we need to represent closures—abstraction nodes—in some way. Often this involves associating to each closure both some information about or code implementing the body of the function, as well as the values of all of the free variables of that closure. [TODO: cite <https://flint.cs.yale.edu/shao/papers/escc.html> or its references for closure compilation] In order to have efficient lookup, we need to know the memory location storing the value of any given variable statically at closure-compilation time. The standard way of doing this [TODO: cite something for flat closure compilation] is to allocate an array of values for each closure. If variables are represented with de Bruijn indices, for example, it's then a very easy array lookup to get the value of any variable. Note that this allocation is linear in the number of free variables of a term. If we have many nested binders and use all of them underneath all the binders, then every abstraction node has as many free variables as there are total binders, and hence we get quadratic overhead.

See #11151 and #11964 and #7826 for an occurrence of this issue in the wild. Note

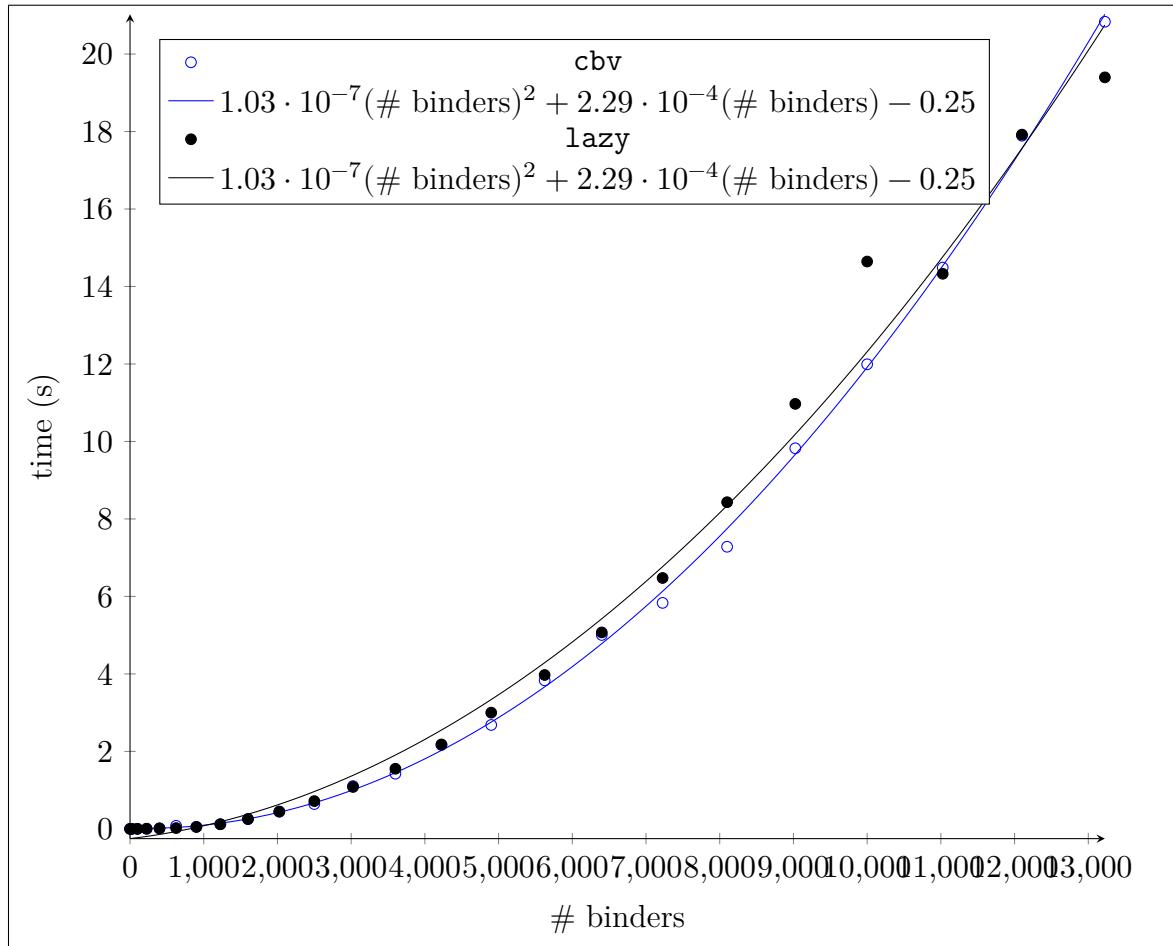


Figure 2-15: Timing of running `cbv` and `lazy` reduction on interpreting a PHOAS expression as a function of the number of binders

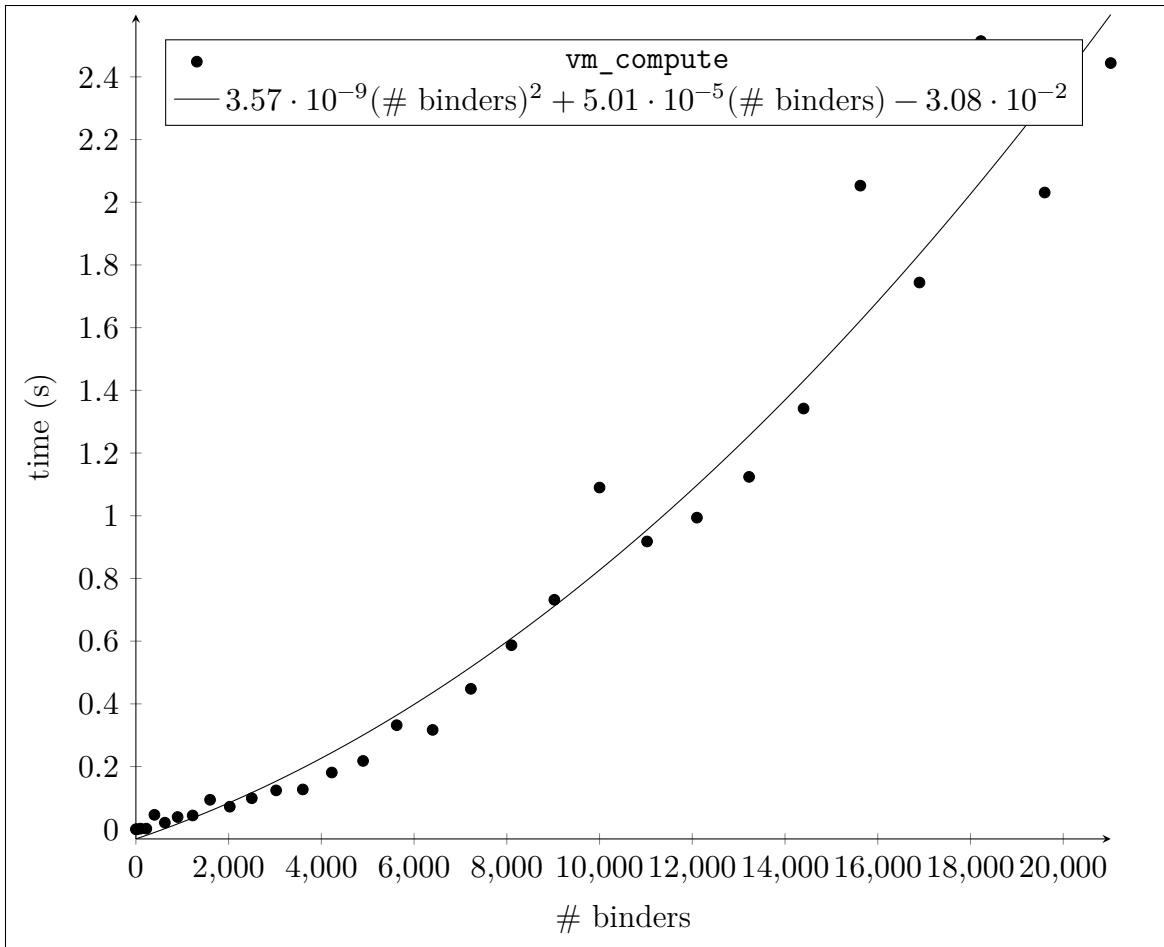


Figure 2-16: Timing of running `vm_compute` reduction on interpreting a PHOAS expression as a function of the number of binders

that this issue rarely shows up in hand-written code, only in generated code, so developers of compilers such as `ocamlc` and `gcc` might be uninterested in optimizing this case. However, it's quite essential when doing meta-programming involving large generated terms. It's especially essential if we want to chain together reflective automation passes that operate on different input languages and therefore require denotation and reification between the passes. In such cases, unless our encoding language uses named or de Bruijn variable encoding, there's no way to avoid large numbers of nested binders at compilation time while preserving code sharing. Hence if we're trying to reuse the work of existing compilers to bootstrap good performance of reduction (as is the case for the native compiler in Coq), we have trouble with cases such as this one. [TODO: reorganize this paragraph, improve it]

See also Figure 2-16 and Figure 2-17.

[TODO: maybe more examples here?]

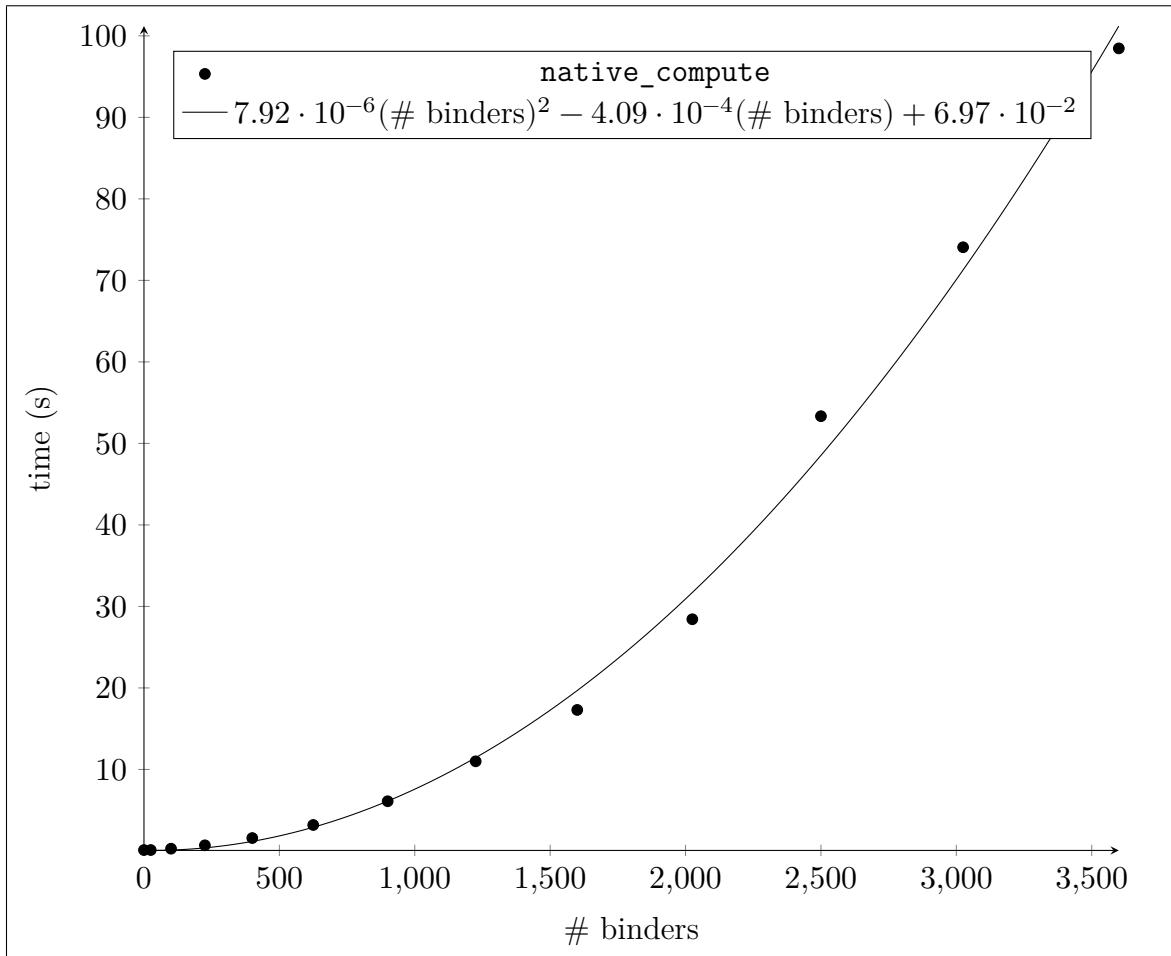


Figure 2-17: Timing of running **native_compute** reduction on interpreting a PHOAS expression as a function of the number of binders

2.2.4 The Number of Nested Abstraction Barriers

[TODO: ADD PLOT: This section requires digging into the historical performance issues around this to find convincing stand-alone examples so we can have graphs here.]

This axis is the most theoretical of the axes. An abstraction barrier is an interface for making use of code, definitions, and theorems. For example, you might define non-negative integers using a binary representation, and present the interface of zero, successor, and the standard induction principle, along with an equational theory for how induction behaves on zero and successor. [TODO: should I spell this example out more?] You might use lists and non-negative integers to implement a hash-set datatype for storing sets of hashable values, and present the hash-set with methods for empty, add, remove, membership-testing, and some sort of fold. Each of these is an abstraction barrier.

There are three primary ways that nested abstraction barriers can lead to performance bottlenecks: one involving conversion missteps and two involving exponential blow-up in the size of types.

Conversion Troubles

If abstraction barriers are not perfectly opaque—that is, if the typechecker ever has to unfold the definitions making up the API in order to typecheck a term—then every additional abstraction barrier provides another opportunity for the typechecker to pick the wrong constant to unfold first. [TODO: add example?] In some typecheckers, such as Coq, it’s possible to provide hints to the typechecker to inform it which constants to unfold when. In such a system, it’s possible to carefully craft conversion hints so that abstraction barriers are always unfolded in the right order. Alternatively, it might be possible to carefully craft a system which picks the right order of unfolding by using a dependency analysis.

However, most users don’t bother to set up hints like this, and dependency analysis isn’t sufficient to determine which abstraction barrier is “higher up” when there are many parts of it, only some of which are mentioned in any given part of the next abstraction barrier. The reason users don’t set up hints like this is that usually it’s not necessary. There’s often minimal overhead, and things just work, even when the wrong path is picked—until the number of abstraction barriers or the size of the underlying term gets large enough. Then we get noticeable exponential blowup, and everything is sad. Furthermore, it’s hard to know which part of conversion is incurring exponential blowup, and thus one has to basically get all of the conversion hints right, simultaneously, without any feedback, to see any performance improvement.

Type Size Blowup: Abstraction Barrier Mismatch

When abstraction barriers are leaky or misaligned, there's a cost that accumulates in the size of the types of theorems. Consider, for example, the two different ways of using tuples: (1) we can use the projections `fst` and `snd`; or (2) we can use the eliminator $\text{pair_rect} : \forall A B (P : A \times B \rightarrow \text{Type}), (\forall a b, P(a, b)) \rightarrow \forall x, P x$. The first gets us access to one element of the tuple at a time, while the second has us using all elements of the tuple simultaneously.

Suppose now there is one API defined in terms of `fst` and `snd`, and another API defined in terms of `pair_rect`. To make these APIs interoperate, we need to explicitly convert from one representation to another. Furthermore, every theorem about the composition of these APIs needs to include the interoperation in talking about how they relate.

If such API mismatches are nested, or if this code size blowup interacts with conversion missteps, then the performance issues compound.

[**TODO:** maybe be more concrete here?]

Type Size Blowup: Packed vs. Unpacked Records

When designing APIs, especially of mathematical objects, one of the biggest choices is whether to pack the records, or whether to pass arguments in as fields. That is, when defining a monoid, for example, there are five ways to go about specifying it:

1. (packed) A *monoid* consists of a type A , a binary operation $\cdot : A \rightarrow A \rightarrow A$, an identity element e , a proof that e is a left- and right-identity $e \cdot a = a \cdot e = a$ for all a , and a proof of associativity that $(a \cdot b) \cdot c = a \cdot (b \cdot c)$.
2. A *monoid on a carrier type A* consists of a binary operation $\cdot : A \rightarrow A \rightarrow A$, an identity element e , a proof that e is a left- and right-identity, and a proof of associativity.
3. A *monoid on a carrier type A under the binary operation* $\cdot : A \rightarrow A \rightarrow A$ consists of an identity element e , a proof that e is a left- and right-identity, and a proof of associativity.
4. (mostly unpacked) A *monoid on a carrier type A under the binary operation* $\cdot : A \rightarrow A \rightarrow A$ *with identity element e* consists of a proof that e is a left- and right-identity and a proof of associativity.
5. (fully unpacked) A monoid on a carrier type A under the binary operation $\cdot : A \rightarrow A \rightarrow A$ with identity element e using a proof p that e is a left- and right-identity and a proof of q of associativity consists of an element of the one-element unit type.

[**TODO:** cite math-classes and HoTT for past design work here]

If we go with anything but the fully packed design, then we incur exponential overhead as we go up abstraction layers, as follows. A *monoid homomorphism* from a monoid A to a monoid B consists of a function between the carrier types, and proofs that this function respects composition and identity. If we use an unpacked definition of monoid with n type parameters, then a similar definition of a monoid homomorphism involves at least $2n + 2$ type parameters. In higher category theory, it's common to talk about morphisms between morphisms, and every additional layer here doubles the number of type arguments, and this can quickly lead to very large terms, resulting in major performance bottlenecks. Note that number of type parameters determines the constant factor out front of the exponential growth in the number of layers of mathematical constructions.

How much is this overhead concretely? When developing a category theory library [89], we sped up overall compilation time by approximately a factor of two, from around 16 minutes to around 8 minutes, by changing one of the two parameters to a field in the definition of a category.⁵

[**TODO:** maybe figure out some examples and include perf data?] [**TODO:** does this need more exposition?]

2.3 Conclusion of this Chapter

[**TODO:** How should this chapter be concluded?] [**TODO:** Maybe another look-forward at what comes next?]

⁵See commit 209231ae3e94d5dbbc678c94930dcf585d53d555 of JasonGross/catdb on GitHub for details.

Part II

API Design

Chapter 3

Design-based fixes

3.1 Introduction

In Chapters 1 and 2, we talked about two different fundamental sources of performance bottlenecks in proof assistants: the power that comes from having dependent types, in Subsection 1.2.1; and the de Bruijn criterion of having a small trusted kernel, in Subsection 1.2.2. In this chapter, we will dive further into the performance issues arising from the first of these design decisions, expanding on Subsection 2.2.4 (The Number of Nested Abstraction Barriers), and proposing some general guidelines for handling these performance bottlenecks.

This chapter is primarily geared at the users of proof assistants, and especially at proof-assistant library developers.

We saw in The Number of Nested Abstraction Barriers three different ways that design choices for abstraction barriers can impact performance: We saw in Type Size Blowup: Abstraction Barrier Mismatch that API mismatch results in type-size blowup. We saw in Conversion Troubles that imperfectly opaque abstraction barriers result in slowdown due to needless calls to the conversion checker. We saw in Type Size Blowup: Packed vs. Unpacked Records how the choice of whether to use packed or unpacked records impacts performance.

In this chapter, we will focus primarily on the first of these ways; while it might seem like a simple question of good design, it turns out that good API design in dependently-typed programming languages is significantly harder than in simply-typed programming languages. Mitigating the second source of performance bottlenecks, imperfectly opaque abstraction barriers, on the other hand, is actually just a question of meticulous tracking of how abstraction barriers are defined and used, and designing them so that they all unfolding is explicit. However, we will present an exception to the rule of opaque abstraction barriers in Section 3.4 in which deliberate

breaking of all abstraction barriers in a careful way can result in performance gains of up to a factor of two: Section 3.4 presents one of our favorite design patterns for categorical constructions: a way of coaxing Coq’s definitional equality into implementing *proof by duality*, one of the most widely known ideas in category theory. [TODO: adjust the tone of what comes after the colon in the previous sentence to fit with this paper] Finally, the question of whether to use packed or unpacked records is actually a genuine trade-off in both design-space and performance, as far as I can tell; the non-performance design considerations have been discussed before in [TODO: cite math classes, etc], while the performance implications are relatively straightforward. As far as I’m aware, there’s not really a good way to get the best of all worlds.

Much of this chapter will draw on examples and experience from a category theory library we implemented in Coq [89], which we introduce in Section 3.3.

3.2 When And How To Use Dependent Types Painlessly

The extremes are relatively easy:

- Total separation between proofs and programs, so that programs are simply typed, works relatively well
- Pre-existing mathematics, where objects are fully bundled with proofs and never need to be separated from them, also works relatively well
- The rule of thumb in the middle: it is painful to recombine proofs and programs after you separate them; if you are doing it to define an opaque transformation that acts on proof-carrying code, that is okay, but if you cannot make that abstraction barrier, enormous pain results.
- For example, if you have length-indexed lists and want to index into them with elements of a finite type, things are fine until you need to divorce the index from its proof of finiteness. If you, for example, want to index into, say, the concatenation of two lists, with an index into the first of the lists, then you will likely run into trouble, because you are trying to consider the index separately from its proof of finitude, but you have to recombine them to do the indexing.

[TODO: flesh out this section] [TODO: More examples from the rewriter?]

3.3 A Brief Introduction To Our Category Theory Library

3.3.1 Introduction

Category theory [124] is a popular all-encompassing mathematical formalism that casts familiar mathematical ideas from many domains in terms of a few unifying

concepts. A *category* can be described as a directed graph plus algebraic laws stating equivalences between paths through the graph. Because of this spartan philosophical grounding, category theory is sometimes referred to in good humor as “formal abstract nonsense.” Certainly the popular perception of category theory is quite far from pragmatic issues of implementation. Our implementation of category theory has run squarely into issues of design and efficient implementation of type theories, proof assistants, and developments within them.

One might presume that it is a routine exercise to transliterate categorical concepts from the whiteboard to Coq. Most category theorists would probably be surprised to learn that standard constructions “run too slowly,” but in our experience that is exactly the result of experimenting with naïve first Coq implementations of categorical constructs. It is important to tune the library design to minimize the cost of manipulating terms and proving interesting theorems.

Category theory, said to be “notoriously hard to formalize” [94], provides a good stress test of any proof assistant, highlighting problems in usability and efficiency.

Formalizing the connection between universal morphisms and adjunctions provides a typical example of our experience with performance. A *universal morphism* is a construct in category theory generalizing extrema from calculus. An *adjunction* is a weakened notion of equivalence. In the process of rewriting our library to be compatible with homotopy type theory, we discovered that cleaning up this construction conceptually resulted in a significant slow-down, because our first attempted rewrite resulted in a leaky abstraction barrier and, most importantly, large goals (Subsection 3.5.2). Plugging the holes there reduced goal sizes by two orders of magnitude¹, which led to a factor of ten speedup in that file (from 39s to 3s), but incurred a factor of three slow-down in the file where we defined the abstraction barriers (from 7s to 21s). Working around slow projections of Σ types (Subsection 3.5.4) and being more careful about code reuse each gave us back half of that lost time.

Although pre-existing formalizations of category theory in proof assistants abound [11, 135, 145, 155, 169, 183, 72], we chose to implement our library [97] from scratch. **[TODO: reword this paragraph a bit]** Beginning from scratch allowed the first author to familiarize himself with both category theory and Coq, without simultaneously having to familiarize himself with a large pre-existing code base.

We begin our discussion in ?? considering a mundane aspect of type definitions that has large consequences for usability and performance. With the expressive power of Coq’s logic Gallina, we often face a choice of making *parameters* of a type family explicit arguments to it, which looks like universal quantification; or of including them within values of the type, which looks like existential quantification. As a general principle, we found that the universal or *outside* style improves the user experience

¹The word count of the larger of the two relevant goals went from 7,312 to 191.

modulo performance, while the existential or *inside* style speeds up type checking. The rule that we settled on was: *inside* definitions for pieces that are usually treated as black boxes by further constructions, and *outside* definitions for pieces whose internal structure is more important later on.

Section 3.4 presents one of our favorite design patterns for categorical constructions: a way of coaxing Coq’s definitional equality into implementing *proof by duality*, one of the most widely known ideas in category theory. In ??, we describe a few other design choices that had large impacts on usability and performance, often of a few orders of magnitude. [TODO: this is duplicative with text above]

3.4 Internalizing Duality Arguments in Type Theory

In general, we tried to design our library so that trivial proofs on paper remain trivial when formalized. One of Coq’s main tools to make proofs trivial is the definitional equality, where some facts follow by computational reduction of terms. We came up with some small tweaks to core definitions that allow a common family of proofs by *duality* to follow by computation.

Proof by duality is a common idea in higher mathematics: sometimes, it is productive to flip the directions of all the arrows. For example, if some fact about least upper bounds is provable, chances are that the same kind of fact about greatest lower bounds will also be provable in roughly the same way, by replacing “greater than”s with “less than”s and vice versa.

Concretely, there is a dualizing operation on categories that inverts the directions of the morphisms:

Notation “ \mathcal{C}^{op} ” := ($\{ \mid \text{Ob} := \text{Ob } \mathcal{C}; \text{Hom } x \ y := \text{Hom } \mathcal{C} \ y \ x; \dots \mid \}$).

Dualization can be used, roughly, for example, to turn a proof that Cartesian product is an associative operation into a proof that disjoint union is an associative operation; products are dual to disjoint unions.

One of the simplest examples of duality in category theory is initial and terminal objects. In a category \mathcal{C} , an initial object 0 is one that has a unique morphism $0 \rightarrow x$ to every object x in \mathcal{C} ; a terminal object 1 is one that has a unique morphism $x \rightarrow 1$ from every object x in \mathcal{C} . Initial objects in \mathcal{C} are terminal objects in \mathcal{C}^{op} . The initial object of any category is unique up to isomorphism; for any two initial objects 0 and $0'$, there is an isomorphism $0 \cong 0'$. By flipping all of the arrows around, we can prove, by duality, that the terminal object is unique up to isomorphism. More precisely, from a proof that an initial object of \mathcal{C}^{op} is unique up to isomorphism, we get that any two

terminal objects $1'$ and 1 in \mathcal{C} , which are initial in \mathcal{C}^{op} , are isomorphic in \mathcal{C}^{op} . Since an isomorphism $x \cong y$ in \mathcal{C}^{op} is an isomorphism $y \cong x$ in \mathcal{C} , we get that 1 and $1'$ are isomorphic in \mathcal{C} .

It is generally straightforward to see that there is an isomorphism between a theorem and its dual, and the technique of dualization is well-known to category theorists, among others. We discovered that, by being careful about how we defined things, we could make theorems be judgmentally equal to their duals! That is, when we prove a theorem

```
initial_ob_unique : ∀ C(x y : Ob C),
  is_initial_ob x → is_initial_ob y → x ≈ y,
```

we can define another theorem

```
terminal_ob_unique : ∀ C(x y : Ob C),
  is_terminal_ob x → is_terminal_ob y → x ≈ y
```

as

```
terminal_ob_unique C x y H H' := initial_ob_unique Cop y x H' H.
```

Interestingly, we found that in proofs with sufficiently complicated types, it can take a few seconds or more for Coq to accept such a definition; we are not sure whether this is due to peculiarities of the reduction strategy of our version of Coq, or speed dependency on the size of the normal form of the type (rather than on the size of the unnormalized type), or something else entirely.

In contrast to the simplicity of witnessing the isomorphism, it takes a significant amount of care in defining concepts, often to get around deficiencies of Coq, to achieve *judgmental* duality. Even now, we were unable to achieve this ideal for some theorems. For example, category theorists typically identify the functor category $\mathcal{C}^{\text{op}} \rightarrow \mathcal{D}^{\text{op}}$ (whose objects are functors $\mathcal{C}^{\text{op}} \rightarrow \mathcal{D}^{\text{op}}$ and whose morphisms are natural transformations) with $(\mathcal{C} \rightarrow \mathcal{D})^{\text{op}}$ (whose objects are functors $\mathcal{C} \rightarrow \mathcal{D}$ and whose morphisms are flipped natural transformations). These categories are canonically isomorphic (by the dualizing natural transformations), and, with the univalence axiom [194], they are equal as categories! However, to make these categories definitionally equal, we need to define functors as a structural record type (see Section 2.2.1) rather than a nominal one.

3.4.1 Duality Design Patterns

One of the simplest theorems about duality is that it is involutive; we have that $(\mathcal{C}^{\text{op}})^{\text{op}} = \mathcal{C}$. In order to internalize proof by duality via judgmental equality, we sometimes need this equality to be judgmental. Although it is impossible in general

in Coq 8.4 (see dodging judgmental η on records below), the latest version of Coq available when we were creating this library, we want at least to have it be true for any explicit category (that is, any category specified by giving its objects, morphisms, etc., rather than referred to via a local variable).

Removing Symmetry

Taking the dual of a category, one constructs a proof that $f \circ (g \circ h) = (f \circ g) \circ h$ from a proof that $(f \circ g) \circ h = f \circ (g \circ h)$. The standard approach is to apply symmetry. However, because applying symmetry twice results in a judgmentally different proof, we decided instead to extend the definition of `Category` to require both a proof of $f \circ (g \circ h) = (f \circ g) \circ h$ and a proof of $(f \circ g) \circ h = f \circ (g \circ h)$. Then our dualizing operation simply swaps the proofs. We added a convenience constructor for categories that asks only for one of the proofs, and applies symmetry to get the other one. Because we formalized 0-truncated category theory, where the type of morphisms is required to have unique identity proofs, asking for this other proof does not result in any coherence issues.

Dualizing the Terminal Category

To make everything work out nicely, we needed the terminal category, which is the category with one object and only the identity morphism, to be the dual of itself. We originally had the terminal category as a special case of the discrete category on n objects. Given a type T with uniqueness of identity proofs, the discrete category on T has as objects inhabitants of T , and has as morphisms from x to y proofs that $x = y$. These categories are not judgmentally equal to their duals, because the type $x = y$ is not judgmentally the same as the type $y = x$. As a result, we instead used the indiscrete category, which has `unit` as its type of morphisms.

Which Side Does the Identity Go On?

The last tricky obstacle we encountered was that when defining a functor out of the terminal category, it is necessary to pick whether to use the right identity law or the left identity law to prove that the functor preserves composition; both will prove that the identity composed with itself is the identity. The problem is that dualizing the functor leads to a road block where either concrete choice turns out to be “wrong,” because the dual of the functor out of the terminal category will not be judgmentally equal to another instance of itself. To fix this problem, we further extended the definition of category to require a proof that the identity composed with itself is the identity.

Dodging Judgmental η on Records

The last problem we ran into was the fact that sometimes, we really, really wanted judgmental η on records. The η rule for records says any application of the record constructor to all the projections of an object yields exactly that object; e.g. for pairs,

$x \equiv (x_1, x_2)$ (where x_1 and x_2 are the first and second projections, respectively). For categories, the η rule says that given a category \mathcal{C} , for a “new” category defined by saying that its objects are the objects of \mathcal{C} , its morphisms are the morphisms of \mathcal{C} , ..., the “new” category is judgmentally equal to \mathcal{C} .

In particular, we wanted to show that any functor out of the terminal category is the opposite of some other functor; namely, any $F : 1 \rightarrow \mathcal{C}$ should be equal to $(F^{\text{op}})^{\text{op}} : 1 \rightarrow (\mathcal{C}^{\text{op}})^{\text{op}}$. However, without the judgmental η rule for records, a local variable \mathcal{C} cannot be judgmentally equal to $(\mathcal{C}^{\text{op}})^{\text{op}}$, which reduces to an application of the constructor for a category, unless the η rule is built into the proof assistant. To get around the problem, we made two variants of dual functors: given $F : \mathcal{C} \rightarrow \mathcal{D}$, we have $F^{\text{op}} : \mathcal{C}^{\text{op}} \rightarrow \mathcal{D}^{\text{op}}$, and given $F : \mathcal{C}^{\text{op}} \rightarrow \mathcal{D}^{\text{op}}$, we have $F^{\text{op}'} : \mathcal{C} \rightarrow \mathcal{D}$. There are two other flavors of dual functors, corresponding to the other two pairings of ${}^{\text{op}}$ with domain and codomain, but we have been glad to avoid defining them so far. As it was, we ended up having four variants of dual natural transformation, and are very glad that we did not need sixteen. When Coq 8.5 was released, we no longer needed to pull this trick, as we could simply enable the η rule for records judgmentally.

3.4.2 Moving Forward: Computation Rules for Pattern Matching

While we were able to work around most of the issues that we had in internalizing proof by duality, things would have been far nicer if we had more η rules. The η rule for records is explained above. The η rule for equality says that the identity function is judgmentally equal to the function $f : \forall x y, x = y \rightarrow x = y$ defined by pattern matching on the first proof of equality; this rule is necessary to have any hope that applying symmetry twice is judgmentally the identity transformation.

Subsection 3.5.1 will give more examples of the pain of manipulating pattern matching on equality. Homotopy type theory provides a framework that systematizes reasoning about proofs of equality, turning a seemingly impossible task into a manageable one. However, there is still a significant burden associated with reasoning about equalities, because so few of the rules are judgmental.

We are currently attempting to divine the appropriate computation rules for pattern matching constructs, in the hopes of making reasoning with proofs of equality more pleasant.²

3.5 A Sampling of Abstraction Barriers

[TODO: maybe this section should come first?]

²See https://coq.inria.fr/bugs/show_bug.cgi?id=3179 and https://coq.inria.fr/bugs/show_bug.cgi?id=3119.

We acknowledge that the concept of performance issues arising from choices of abstraction barriers may seem a bit counter-intuitive. After all, abstraction barriers generally live in the mind of the developer, in some sense, and it seems a bit insane to say that performance of the code depends on the mental state of the programmer.

Therefore, we will describe a sampling of abstraction barriers and the design choices that went into them, drawn from real examples [**QUESTION FOR ADAM: how much should I talk about the category theory library itself in my thesis?**], as well as the performance issues that arose from these choices.

A few other pervasive strategies made non-trivial differences for proof performance or simplicity.

3.5.1 Identities vs. Equalities; Associators

There are a number of constructions that are provably equal, but which we found more convenient to construct transformations between instead, despite the increased verbosity of such definitions. This is especially true of constructions that strayed towards higher category theory. For example, when constructing the Grothendieck construction of a functor to the category of categories, we found it easier to first generalize the construction from functors to pseudofunctors. The definition of a pseudofunctor results from replacing various equalities in the definition of a functor with isomorphisms (analogous to bijections between sets or types), together with proofs that the isomorphisms obey various coherence properties. This replacement helped because there are fewer operations on isomorphisms (namely, just composition and inverting), and more operations on proofs of equality (pattern matching, or anything definable via induction); when we were forced to perform all of the operations in the same way, syntactically, it was easier to pick out the operations and reason about them.

Another example was defining the (co)unit of adjunction composition, where instead of a proof that $F \circ (G \circ H) = (F \circ G) \circ H$, we used a natural transformation, a coherent mapping between the actions of functors. Where equality-based constructions led to computational reduction getting stuck at casts, the constructions with natural transformations reduce in all of the expected contexts.

3.5.2 Opacity; Linear Dependence of Speed on Term Size

Coq is slow at dealing with large terms. For goals around 175,000 words long³, we have found that simple tactics like `apply f_equal` take around 1 second to execute, which makes interactive theorem proving very frustrating.⁴ Even more frustrating is the fact that the largest contribution to this size is often arguments to irrelevant functions, i.e., functions that are provably equal to all other functions of the same

³When we had objects as arguments rather than fields (see ??), we encountered goals of about 219,633 words when constructing pointwise Kan extensions.

⁴See also https://coq.inria.fr/bugs/show_bug.cgi?id=3280.

type. (These are proofs related to algebraic laws like associativity, carried inside many constructions.)

Opacification helps by preventing the type checker from unfolding some definitions, but it is not enough: the type checker still has to deal with all of the large arguments to the opaque function. Hash-consing might fix the problem completely.

Alternatively, it would be nice if, given a proof that all of the inhabitants of a type were equal, we could forget about terms of that type, so that their sizes would not impose any penalties on term manipulation. One solution might be irrelevant fields, like those of Agda, or implemented via the Implicit CiC [26, 136].

3.5.3 Abstraction Barriers

In many projects, choosing the right abstraction barriers is essential to reducing mistakes, improving maintainability and readability of code, and cutting down on time wasted by programmers trying to hold too many things in their heads at once. This project was no exception; we developed an allergic reaction to constructions with more than four or so arguments, after making one too many mistakes in defining limits and colimits. Limits are a generalization, to arbitrary categories, of subsets of Cartesian products. Colimits are a generalization, to arbitrary categories, of disjoint unions modulo equivalence relations.

Our original flattened definition of limits involved a single definition with 14 nested binders for types and algebraic properties. After a particularly frustrating experience hunting down a mistake in one of these components, we decided to factor the definition into a larger number of simpler definitions, including familiar categorical constructs like terminal objects and comma categories. This refactoring paid off even further when some months later we discovered the universal morphism definition of adjoint functors. With a little more abstraction, we were able to reuse the same decomposition to prove the equivalence between universal morphisms and adjoint functors, with minimal effort.

Perhaps less typical of programming experience, we found that picking the right abstraction barriers could drastically reduce compile time by keeping details out of sight in large goal formulas. In the instance discussed in the introduction, we got a factor of ten speed-up by plugging holes in a leaky abstraction barrier!⁵

3.5.4 Nested Σ Types

In Coq, there are two ways to represent a data structure with one constructor and many fields: as a single inductive type with one constructor (records), or as a nested Σ type. For instance, consider a record type with two type fields A and B and a

⁵See <https://github.com/HoTT/HoTT/commit/eb0099005171> for the exact change.

function f from A to B . A logically equivalent encoding would be $\Sigma A. \Sigma B. A \rightarrow B$. There are two important differences between these encodings in Coq.

The first is that while a theorem statement may abstract over all possible Σ types, it may not abstract over all record types, which somehow have a less first-class status. Such a limitation is inconvenient and leads to code duplication.

The far more pressing problem, overriding the previous point, is that nested Σ types have horrendous performance, and are sometimes a few orders of magnitude slower. The culprit is projections from nested Σ types, which, when unfolded (as they must be, to do computation), each take almost the entirety of the nested Σ type as an argument, and so grow in size very quickly.

Let's consider a toy example to see the asymptotic performance. To construct a nested Σ type with three fields of type `unit`, we can write the type:

```
{ _ : unit & { _ : unit & unit }}
```

If we want to project out the final field, we must write `projT2 (projT2 x)` which, when implicit arguments are included, expands to [TODO: figure out why spacing is strange above this code example]

```
@projT2 unit (λ _ : unit, unit) (@projT2 unit (λ _ : unit, { _ : unit & unit })) x
```

This term grows quadratically in the number of projections because the type of the n^{th} field is repeated approximately $2n$ times. This is even more of a problem when we need to `destruct` x to prove something about the projections, as we need to `destruct` it as many times as their are fields, which adds another factor of n to the performance cost of building the proof from scratch; in Coq, this cost is either avoided due to sharing or else is hidden by a quadratic factor which a much larger constant factor. Note that this is a sort-of dual to the problem of Subsection 2.2.1; there, we encountered quadratic overhead in applying the constructors (which is also a problem here), whereas right now we are discussing quadratic overhead in applying the eliminators. See Figure 3-1 for the performance details.

We can avoid much of the cost of building the projection term by using *primitive projections* (see Subsection 8.1.6 for more explanation of this feature). Note that this feature is a sort-of dual to the proposed feature of dropping constructor parameters described in Section 2.2.1. This does drastically reduce the overhead of building the projection term, but only cuts in half the constant factor in destructing the variable so as to prove something about the projection. See Figure 3-2 for performance details.

There are two solutions to this issue:

1. use built-in *record* types
2. carefully define intermediate abstraction barriers to avoid the quadratic overhead

Both of these essentially solve the issue of quadratic overhead in projecting out the fields. This is the benefit of good abstraction barriers.

In Coq 8.11, **destruct** is unfortunately still quadratic due to issues with name generation, but the constant factor is much smaller; see ?? and #12271.

We now come to the question: how much do we pay for using this abstraction barrier? That is, how much is the one-time cost of defining the abstraction barrier. Obviously, we can just make definitions for each of the projections and for the eliminator, and pay the cubic (or perhaps even quartic; see the leading term in Figure 3-1) overhead once. There's an interesting question, though, of if we can avoid this overhead all-together.

As seen in ??, using records partially avoids the overhead. Defining the record type, though, still incurs a quadratic factor due to hashconsing the projections; see #12270.

[TODO: remove use of “you”] If your proof assistant does not support records out-of-the-box, or you want to avoid using them for whatever reason⁶, you can instead define intermediate abstraction barriers by hand. Here is what code that almost works looks like for four fields:

Local Set Implicit Arguments.

```
Record sigT {A} (P : A -> Type) := existT { projT1 : A ; projT2 : P projT1 }.
Definition sigT_eta {A P} (x : @sigT A P) : x = existT P (projT1 x) (projT2 x).
Proof. destruct x; reflexivity. Defined.
Definition _T0 := unit.
Definition _T1 := @sigT unit (fun _ : unit => _T0).
Definition _T2 := @sigT unit (fun _ : unit => _T1).
Definition _T3 := @sigT unit (fun _ : unit => _T2).
Definition T := _T3.
Definition Build_T0 (x0 : unit) : _T0 := x0.
Definition Build_T1 (x0 : unit) (rest : _T0) : _T1
  := @existT unit (fun _ : unit => _T0) x0 rest.
Definition Build_T2 (x0 : unit) (rest : _T1) : _T2
  := @existT unit (fun _ : unit => _T1) x0 rest.
Definition Build_T3 (x0 : unit) (rest : _T2) : _T3
  := @existT unit (fun _ : unit => _T2) x0 rest.
Definition Build_T (x0 : unit) (x1 : unit) (x2 : unit) (x3 : unit) : T
```

⁶Note that the UniMath library [TODO: cite unimath] does this. [TODO: explain reasoning]

```

:= Build_T3 x0 (Build_T2 x1 (Build_T1 x2 (Build_T0 x3))).
```

```

Definition _T0_proj (x : _T0) : unit := x.
Definition _T1_proj1 (x : _T1) : unit := projT1 x.
Definition _T1_proj2 (x : _T1) : _T0 := projT2 x.
Definition _T2_proj1 (x : _T2) : unit := projT1 x.
Definition _T2_proj2 (x : _T2) : _T1 := projT2 x.
Definition _T3_proj1 (x : _T3) : unit := projT1 x.
Definition _T3_proj2 (x : _T3) : _T2 := projT2 x.
```

```

Definition proj_T_1 (x : T) : unit := _T3_proj1 x.
Definition proj_T_1_rest (x : T) : _T2 := _T3_proj2 x.
Definition proj_T_2 (x : T) : unit := _T2_proj1 (proj_T_1_rest x).
Definition proj_T_2_rest (x : T) : _T1 := _T2_proj2 (proj_T_1_rest x).
Definition proj_T_3 (x : T) : unit := _T1_proj1 (proj_T_2_rest x).
Definition proj_T_3_rest (x : T) : _T0 := _T1_proj2 (proj_T_2_rest x).
Definition proj_T_4 (x : T) : unit := _T0_proj (proj_T_3_rest x).
```

```

Definition _T0_eta (x : _T0) : x = Build_T0 (_T0_proj x) := @eq_refl _T0 x.
Definition _T1_eta (x : _T1) : x = Build_T1 (_T1_proj1 x) (_T1_proj2 x)
  := @sigT_eta unit (fun _ : unit => _T0) x.
Definition _T2_eta (x : _T2) : x = Build_T2 (_T2_proj1 x) (_T2_proj2 x)
  := @sigT_eta unit (fun _ : unit => _T1) x.
Definition _T3_eta (x : _T3) : x = Build_T3 (_T3_proj1 x) (_T3_proj2 x)
  := @sigT_eta unit (fun _ : unit => _T2) x.
```

```

Definition T_eta (x : T)
  : x = Build_T (proj_T_1 x) (proj_T_2 x) (proj_T_3 x) (proj_T_4 x)
  := let lhs3 := x in
    let lhs2 := _T3_proj2 lhs3 in
    let lhs1 := _T2_proj2 lhs2 in
    let lhs0 := _T1_proj2 lhs1 in
    let final := _T0_proj lhs0 in
    let rhs0 := Build_T0 final in
    let rhs1 := Build_T1 (_T1_proj1 lhs1) rhs0 in
    let rhs2 := Build_T2 (_T2_proj1 lhs2) rhs1 in
    let rhs3 := Build_T3 (_T3_proj1 lhs3) rhs2 in
    (((@eq_trans _T3)
      lhs3 (Build_T3 (_T3_proj1 lhs3) lhs2) rhs3
      (_T3_eta lhs3)
      ((@f_equal _T2 _T3 (Build_T3 (_T3_proj1 lhs3)))
        lhs2 rhs2
        ((@eq_trans _T2)
          lhs2 (Build_T2 (_T2_proj1 lhs2) lhs1) rhs2
          (_T2_eta lhs2)
```

```

((@f_equal _T1 _T2 (Build_T2 (_T2_proj1 lhs2)))
  lhs1 rhs1
  ((@eq_trans _T1)
    lhs1 (Build_T1 (_T1_proj1 lhs1) lhs0) rhs1
    (_T1_eta lhs1)
    ((@f_equal _T0 _T1 (Build_T1 (_T1_proj1 lhs1)))
      lhs0 rhs0
      (_T0_eta lhs0))))))
: x = Build_T (proj_T_1 x) (proj_T_2 x) (proj_T_3 x) (proj_T_4 x).

```

Import EqNotations.

```

Definition T_rect (P : T -> Type)
  (f : forall (x0 : unit) (x1 : unit) (x2 : unit) (x3 : unit),
   P (Build_T x0 x1 x2 x3))
  (x : T)
  : P x
  := rew <- [P] T_eta x in
  f (proj_T_1 x) (proj_T_2 x) (proj_T_3 x) (proj_T_4 x).

```

It only almost works because, although the overall size of the terms, even accounting for implicits, is linear in the number of fields, we still incur a quadratic number of unfoldings in the final cast node in the proof of `T_eta`. Note that this cast node is only present to make explicit the conversion problem that must happen; removing it does not break anything, but then the quadratic cost is hidden in non-trivial substitutions of the `let`-binders into the types. It might be possible to avoid this quadratic factor by being even more careful, but I was unable to find a way to do it.⁷ Worse, though, due to the issue with nested `let`-binders described in Section 2.2.1, we would still incur a quadratic typechecking cost.

We can, however, avoid this cost by turning on primitive projections via `Set Primitive Projections` at the top of this block of code: this enables judgmental η -conversion for primitive records, whence we can prove `T_eta` with the proof term `@eq_refl T x`. See ?? for performance details.

⁷Note that even reflective automation (see Chapter 4) is not sufficient to solve this issue. Essentially, the bottleneck is that at the bottom of the chain of `let`-binders in the η proof, we have two different types for the η -principle. One of them uses the globally-defined projections out of `T`, while the other uses the projections of `x` defined in the local context. We need to convert between these two types in linear time. Converting between two differently defined projections takes time linear in the number of under-the-hood projections, i.e., linear in the number of fields. Doing this once for each projection thus takes quadratic time. Using a reflective representation of nested Σ types, and thus being able to prove the η principle once and for all in constant time, would not help here, because it takes quadratic time to convert between the type of the η principle in reflective-land and the type that we want. One thing that might help would be to have a version of conversion checking that was both memoized and could perform in-place reduction; see [TODO: cite <https://github.com/coq/coq/issues/12269>].

[**TODO:** make this timing graph] [**TODO:** move figures up to be closer to references]

[**TODO:** Find more examples to talk about here?]

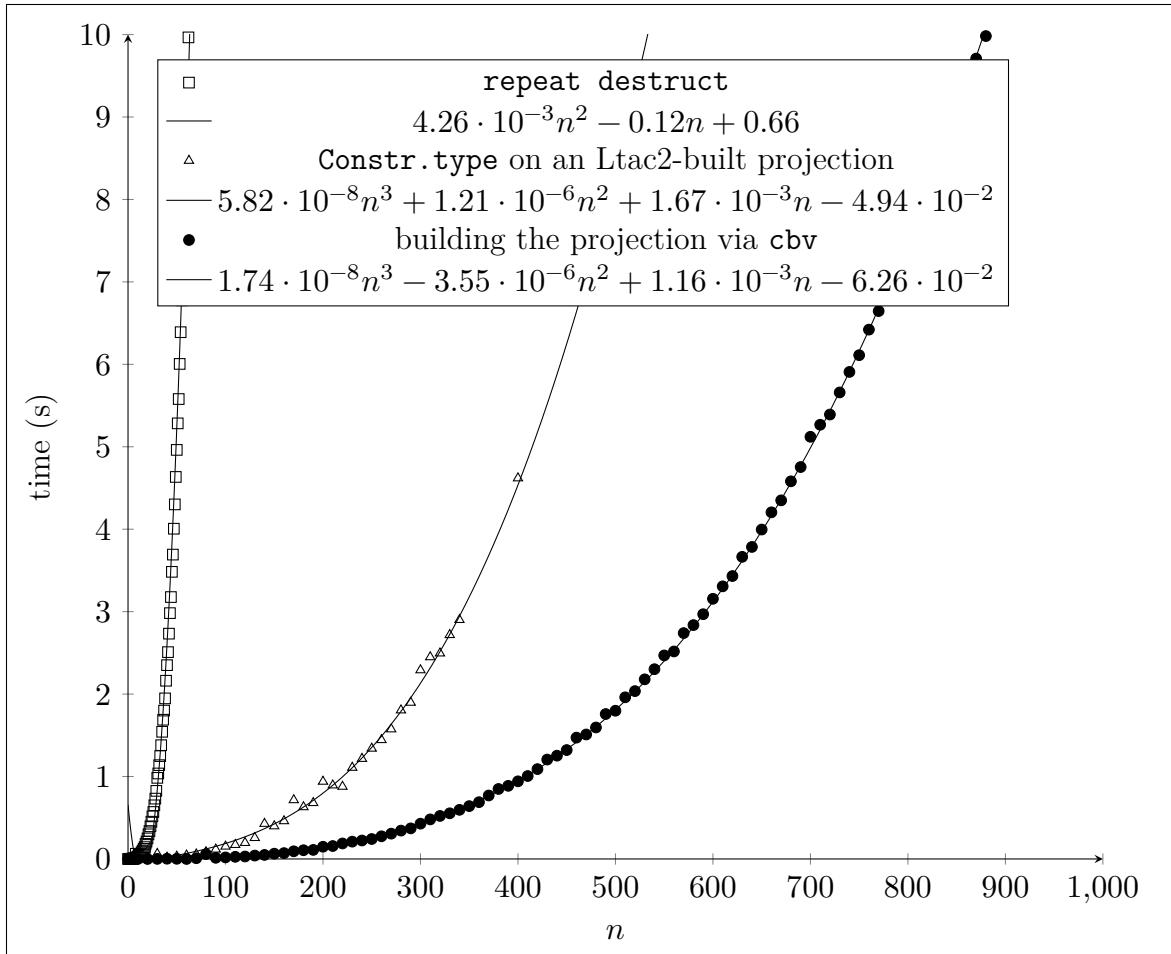


Figure 3-1: There are two ways we look at the performance of building a term like `projT1 (projT2 ... (projT2 x))` with n `projT2`s: we can define a recursive function that computes this term and then use `cbv` to reduce away the recursion, and time how long this takes; or we can build the term using Ltac2 and then typecheck it. This plot displays both of these methods, and in addition displays the time it takes to run `destruct` to break x into its component fields, as lower bound for how long it takes to prove anything about a nested Σ type with n fields.

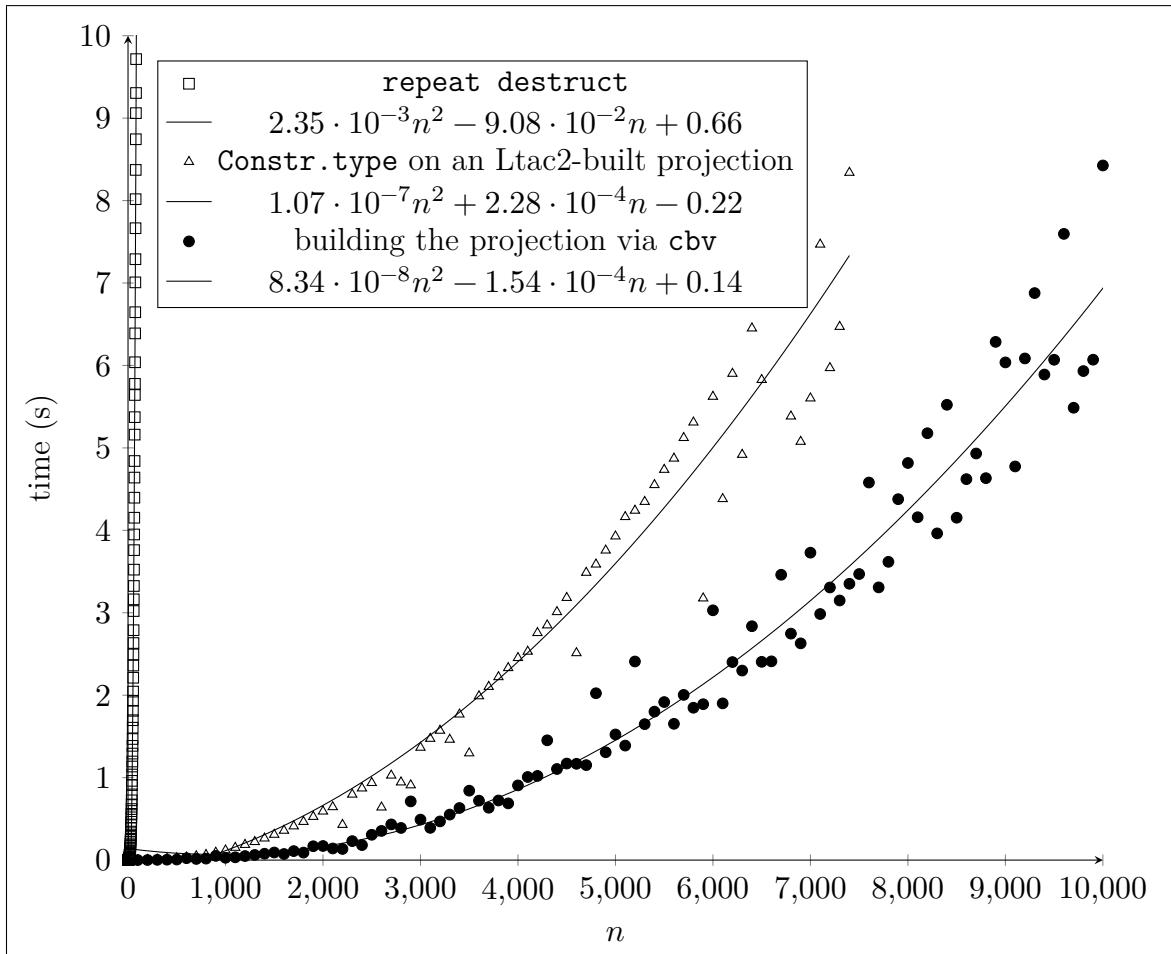


Figure 3-2: The same graph as Figure 3-1, but with primitive projections turned on. Note that the x -axis is $10\times$ larger on this plot.

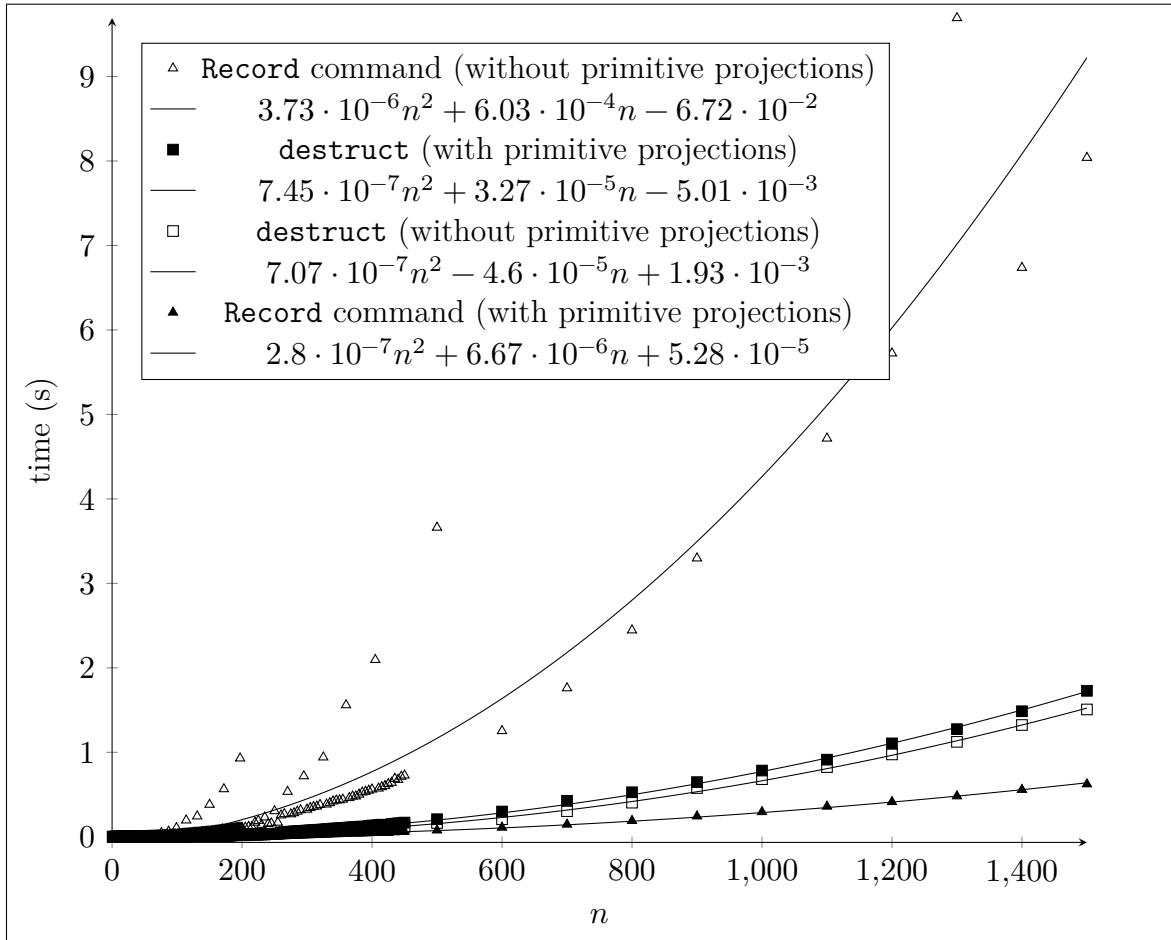


Figure 3-3: Timing of running a **Record** command to define a record with n fields, and the time to **destruct** such a record. Note that building the goal involving projecting out the last field takes less than 0.001s for all numbers of fields that we tested. (Presumably for large enough numbers of fields, we'd start getting a logarithmic overhead from parsing the name of the final field, which, when represented as x followed by the field number in base 10, does grow in size as $\log_{10} n$.) Note that the non-monotonic timing is *reproducible*, and we have asked the Coq developers about it at #12270.

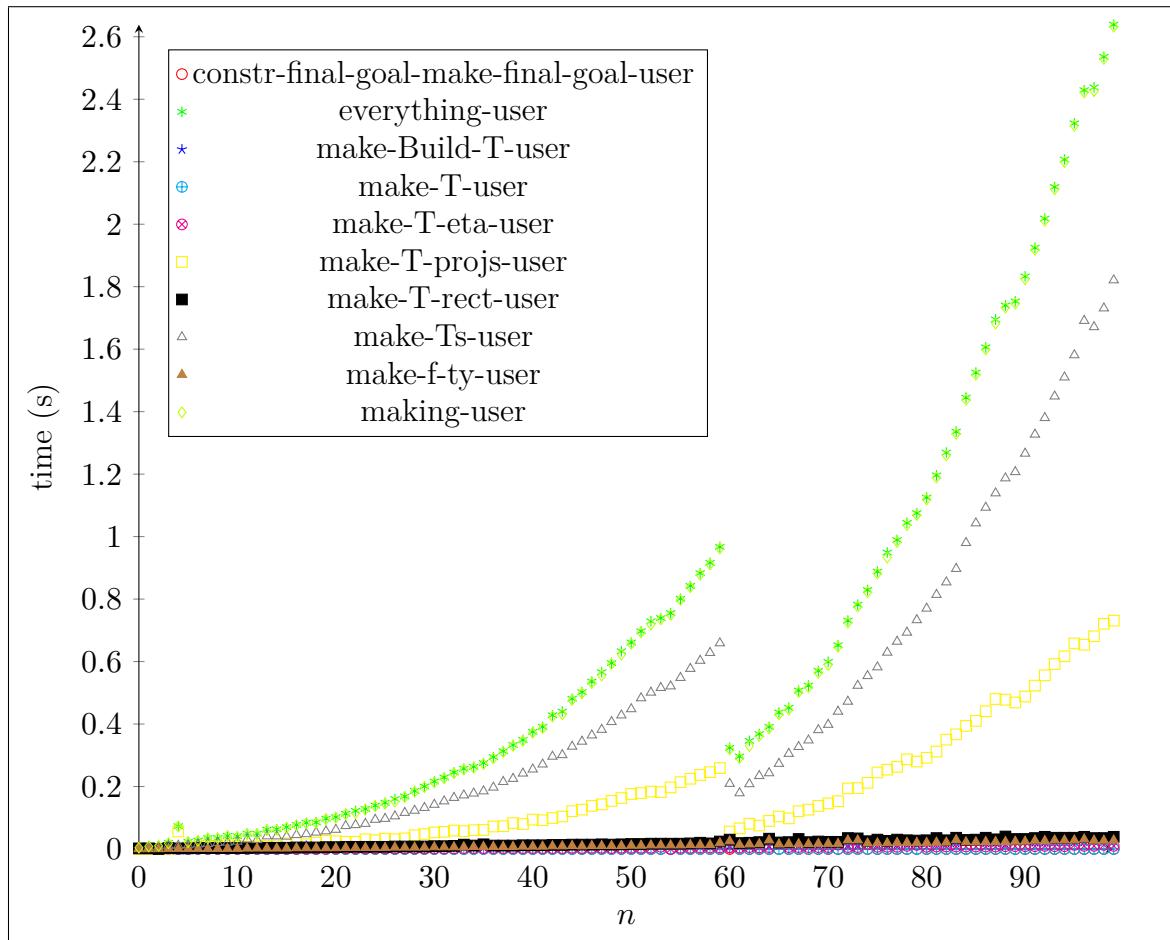


Figure 3-4: timing-performance-experiments/make-nested-prim-prod-abstraction.txt

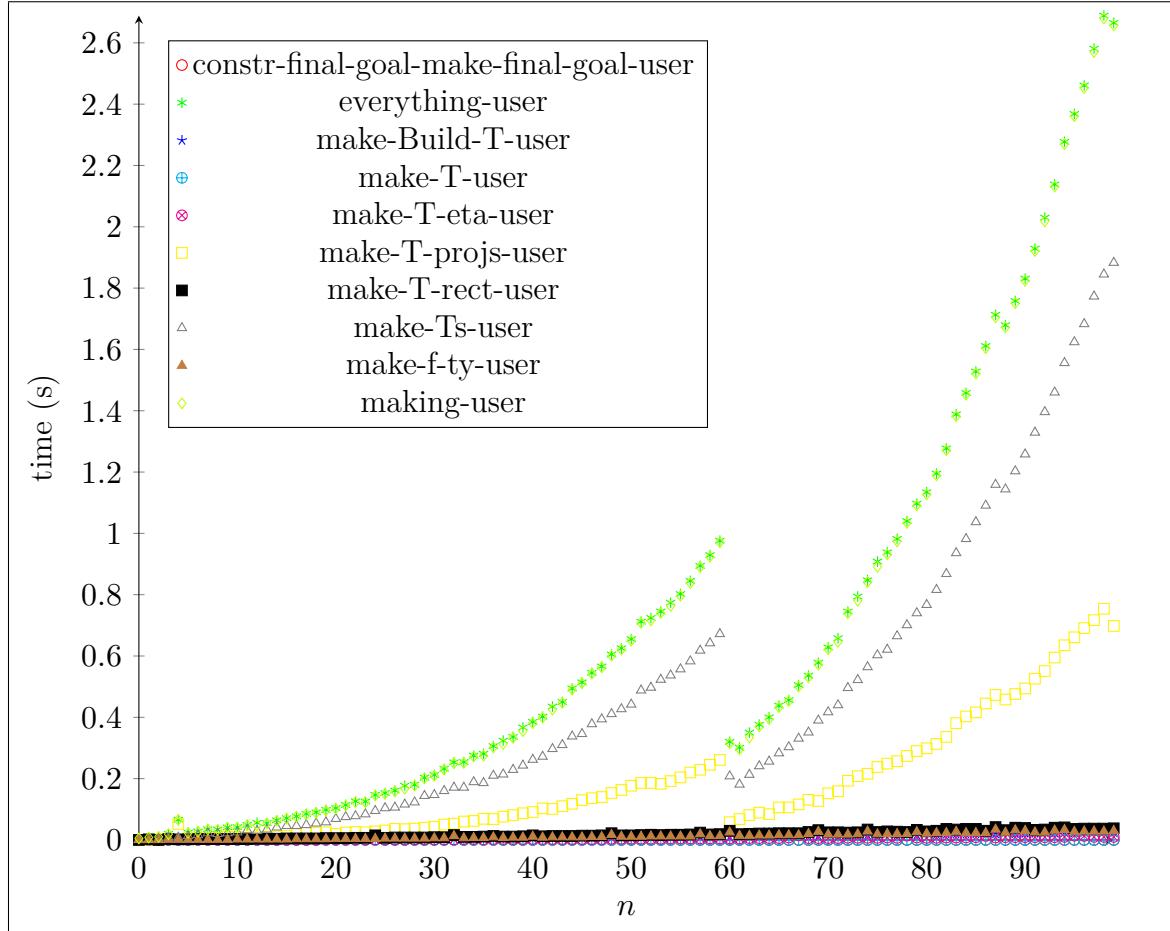


Figure 3-5: timing-performance-experiments/make-nested-prim-sig-abstraction.txt

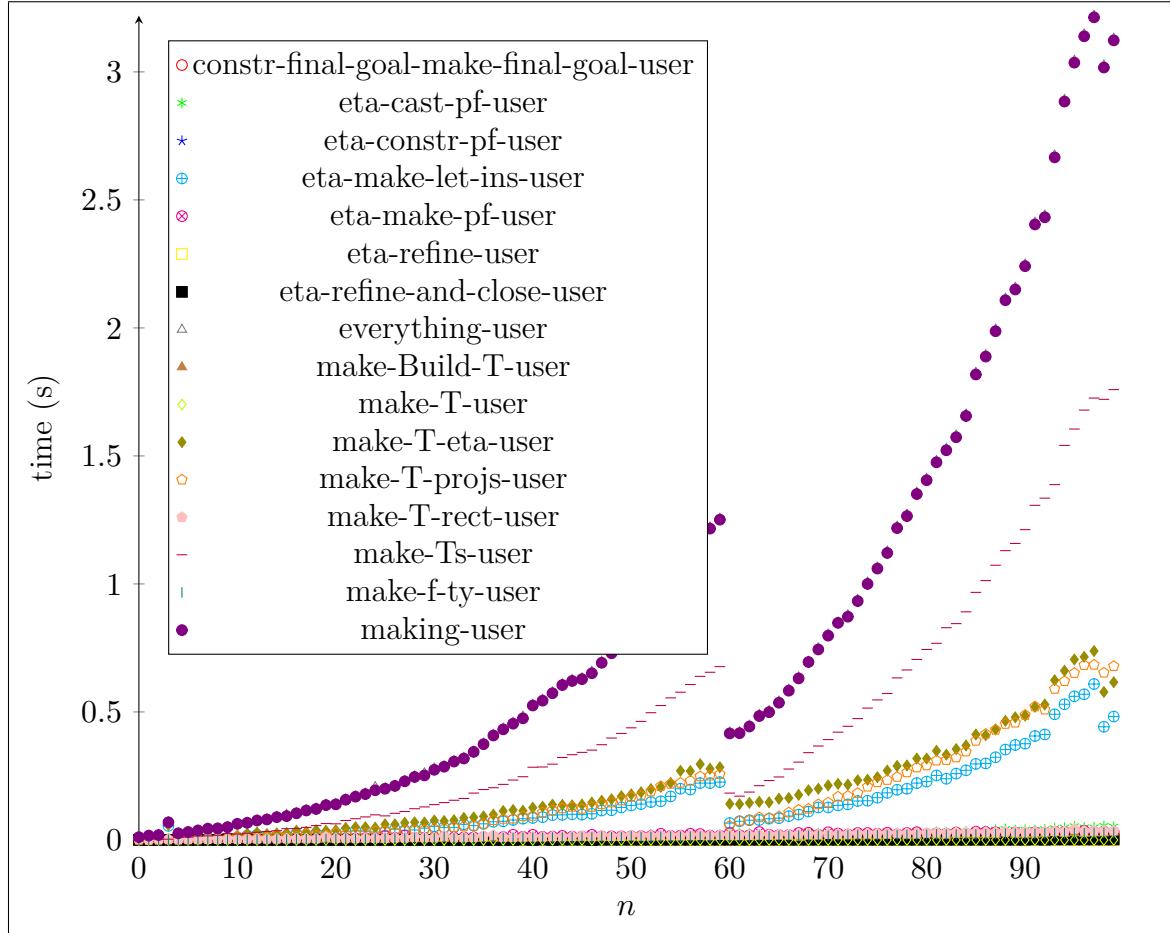


Figure 3-6: timing-performance-experiments/make-nested-prod-abstraction.txt

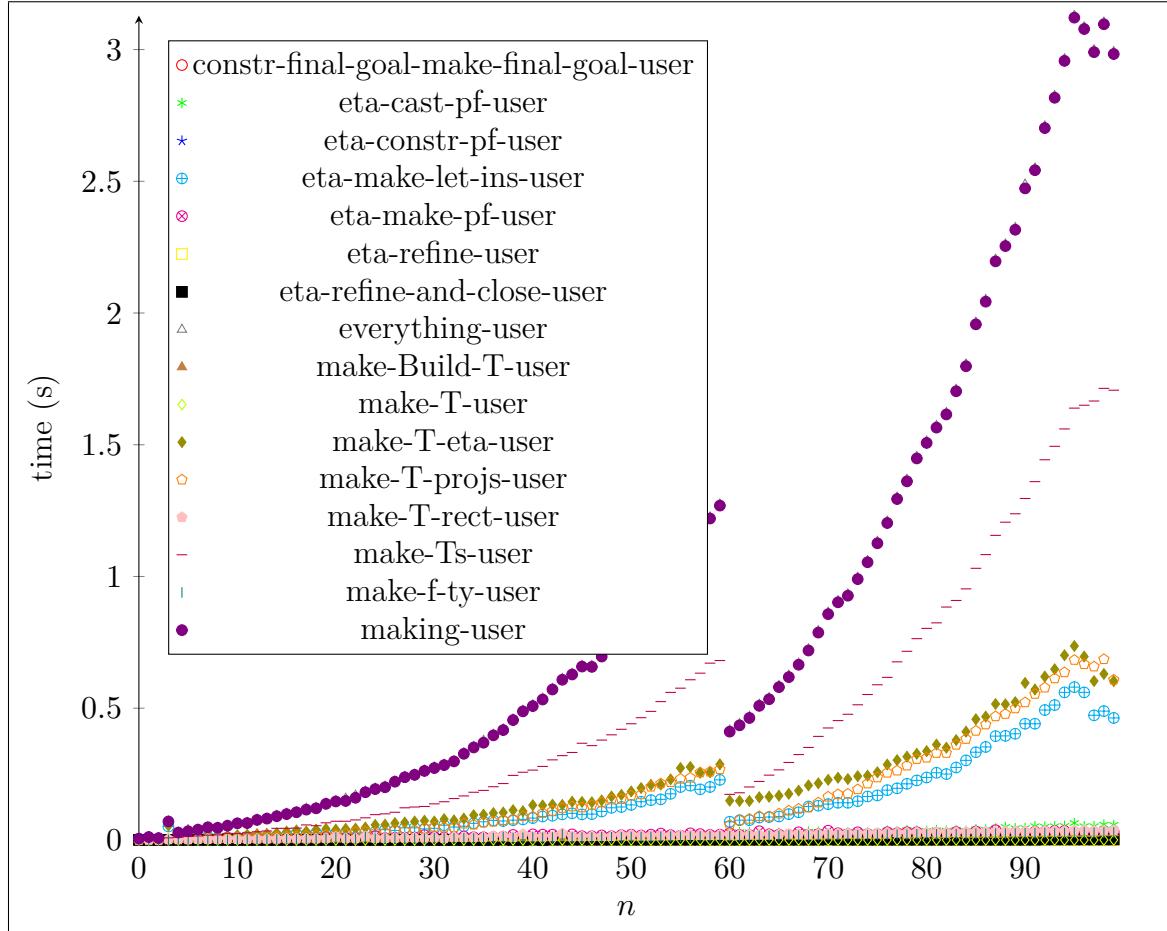


Figure 3-7: timing-performance-experiments/make-nested-sig-abstraction.txt

Part III

Program Transformation and Rewriting

Chapter 4

Reflective Program Transformation

4.1 Introduction

[**TODO:** reformat this for flow, right now it's from reification-by-parametricity]

Proof by reflection [37] is an established method for employing verified proof procedures, within larger proofs. [129, 130, 125, 80] [**TODO:** should we cite this many things here, or remove some of them?] There are a number of benefits to using verified functional programs written in the proof assistant’s logic, instead of tactic scripts. We can often prove that procedures always terminate without attempting fallacious proof steps, and perhaps we can even prove that a procedure gives logically complete answers, for instance telling us definitively whether a proposition is true or false. In contrast, tactic-based procedures may encounter runtime errors or loop forever. As a consequence, if we want to keep the trusted codebase small, as discussed in Subsection 1.2.2, these tactic procedures must output proof terms, justifying their decisions, and these terms can grow large, making for slower proving and requiring transmission of large proof terms to be checked slowly by others. A verified procedure need not generate a certificate for each invocation.

The starting point for proof by reflection is *reification*: translating a “native” term of the logic into an explicit abstract syntax tree. We may then feed that tree to verified procedures or any other functional programs in the logic. The benefits listed above are particularly appealing in domains where goals are very large. For instance, consider verification of large software systems, where we might want to reify thousands of lines of source code. Popular methods turn out to be surprisingly slow, often to the point where, counter-intuitively, the majority of proof-execution time is spent in reification – unless the proof engineer invests in writing a plugin directly in the proof assistant’s metalanguage (e.g., OCaml for Coq).

[**TODO:** move this paragraph elsewhere] In this paper, we show that reification

can be both simpler and faster than with standard methods. Perhaps surprisingly, we demonstrate how to reify terms almost entirely through reduction in the logic, with a small amount of tactic code for setup and no ML programming. Though our techniques should be broadly applicable, especially in proof assistants based on type theory, our experience is with Coq, and we review the requisite background in the remainder of this introduction. In Section 7.2, we summarize our survey into prior approaches to reification and provide high-quality implementations and documentation for them, serving a tutorial function independent of our new contributions. Experts on the subject might want to skip directly to Section 7.3, which explains our alternative technique. We benchmark our approach against 18 competitors in Section 7.4.

4.1.1 Proof-Script Primer

[TODO: rename this subsection] Basic Coq proofs are often written as lists of steps such as `induction` on some structure, `rewrite` using a known equivalence, or `unfold` of a definition. Very quickly, proofs can become long and tedious, both to write and to read, and hence Coq provides \mathcal{L}_{tac} , a scripting language for proofs. As theorems and proofs grow in complexity, users frequently run into performance and maintainability issues with \mathcal{L}_{tac} . Consider the case where we want to prove that a large algebraic expression, involving many `let ... in ...` expressions, is even:

```
Inductive is_even : nat -> Prop :=
| even_0 : is_even 0
| even_SS : forall x, is_even x -> is_even (S (S x)).
Goal is_even (let x := 100 * 100 * 100 * 100 in
              let y := x * x * x * x in
              y * y * y * y).
```

Coq stack-overflows if we try to reduce this goal. As a workaround, we might write a lemma that talks about evenness of `let ... in ...`, plus one about evenness of multiplication, and we might then write a tactic that composes such lemmas.

[TODO: factor some of this with chapter 2] Even on smaller terms, though, proof size can quickly become an issue. If we give a naive proof that 7000 is even, the proof term will contain all of the even numbers between 0 and 7000, giving a proof-term-size blow-up at least quadratic in size (recalling that natural numbers are represented in unary; the challenges remain for more efficient base encodings). Clever readers will notice that Coq could share subterms in the proof tree, recovering a term that is linear in the size of the goal. However, such sharing would have to be preserved very carefully, to prevent size blow-up from unexpected loss of sharing, and today's Coq version does not do that sharing. Even if it did, tactics that rely on assumptions about Coq's sharing strategy become harder to debug, rather than easier.

4.1.2 Reflective-Automation Primer

Enter reflective automation, which simultaneously solves both the problem of performance and the problem of debuggability. Proof terms, in a sense, are traces of a proof script. They provide Coq's kernel with a term that it can check to verify that no illegal steps were taken. Listing every step results in large traces.

The idea of reflective automation is that, if we can get a formal encoding of our goal, plus an algorithm to *check* the property we care about, then we can do much better than storing the entire trace of the program. We can prove that our checker is correct once and for all, removing the need to trace its steps.

A simple evenness checker can just operate on the unary encoding of natural numbers (Figure 7-1). We can use its correctness theorem to prove goals much more quickly:

```
Fixpoint check_is_even (n : nat) : bool
:= match n with
  | 0 => true
  | 1 => false
  | S (S n) => check_is_even n
end.
```

Figure 4-1: Evenness Checking

```
Theorem soundness : forall n, check_is_even n = true -> is_even n.
Goal is_even 2000.
Time repeat (apply even_SS || apply even_0). (* 1.8 s *)
Undo.
Time apply soundness; vm_compute; reflexivity. (* 0.004 s *)
```

The tactic **vm_compute** tells Coq to use its virtual machine for reduction, to compute the value of `check_is_even 2000`, after which **reflexivity** proves that `true = true`. Note how much faster this method is. In fact, even the asymptotic complexity is better; this new algorithm is linear rather than quadratic in `n`.

However, even this procedure takes a bit over three minutes to prove the goal `is_even (10 * 10 * 10 * 10 * 10 * 10 * 10 * 10 * 10)`. To do better, we need a formal representation of terms or expressions.

4.1.3 Reflective-Syntax Primer

[**TODO:** should the footnote be uncommented here?] Sometimes, to achieve faster proofs, we must be able to tell, for example, whether we got a term by multiplication or by addition, and not merely whether its normal form is 0 or a successor.

A reflective automation procedure generally has two steps. The first step is to *reify* the goal into some abstract syntactic representation, which we call the *term language* or an *expression language*.

The second step is to run the algorithm on the reified syntax. The following figure shows some examples of simple expressions in this language.

```
Inductive expr :=
| Nat0 : expr
| NatS (x : expr) : expr
| NatMul (x y : expr) : expr.
```

What should our expression language include? At a bare minimum, we must have multiplication nodes, and we must have `nat` literals. If we encode `S` and `0` separately, a decision that will become important later in Section 7.3, we get the inductive type of Figure 7-2.

Before diving into methods of reification, let us write the evenness checker.

```
Fixpoint check_is_even_expr (t : expr) : bool
:= match t with
| Nat0 => true
| NatS x => negb (check_is_even_expr x)
| NatMul x y => orb (check_is_even_expr x) (check_is_even_expr y)
end.
```

Before we can state the soundness theorem (whenever this checker returns `true`, the represented number is even), we must write the function that tells us what number our expression represents, called *denotation* or *interpretation*:

```
Fixpoint denote (t : expr) : nat
:= match t with
| Nat0 => 0
| NatS x => S (denote x)
| NatMul x y => denote x * denote y
end.
```

```
Theorem check_is_even_expr_sound (e : expr)
: check_is_even_expr e = true -> is_even (denote e).
```

Given a tactic `Reify` to produce a reified term from a `nat`, we can time `check_is_even_expr`. It is instant on the last example.

Before we proceed to reification, we will introduce one more complexity. If we want to support our initial example with `let ... in ...` efficiently, we must also have `let`-expressions. Our current procedure that inlines `let`-expressions takes 19 seconds, for example, on `let x0 := 10 * 10 in let x1 := x0 * x0 in ... let x24 := x23 * x23 in x24`. The choices of representation of binders, which are essential to encoding `let`-expressions, include higher-order abstract syntax (HOAS) [156], parametric higher-order abstract syntax (PHOAS) [50] also known as weak HOAS [51], de Bruijn indices [40], nominal representations [159], locally nameless representations [47, 121], named representations, and nested abstract syntax [95, 32]. A survey of a number of options for binding can be found in [24]. **[QUESTION FOR ADAM: is this a place to cite POPLMark? What should be said about it? Should I mention more sorts of representations?]**

Although we will eventually choose the PHOAS representation for the tools presented in ??, we will also briefly survey some of the options for encoding binders, with an eye towards performance implications.

PHOAS

The PHOAS representation [50, 51] is particularly convenient. In PHOAS, expression binders are represented by binders in Gallina, the functional language of Coq, and the expression language is parameterized over the type of the binder. Let us define a constant and notation for `let` expressions as definitions (a common choice in real Coq developments, to block Coq’s default behavior of inlining `let` binders silently; the same choice will also turn out to be useful for reification later). We thus have:

```

Inductive expr {var : Type} :=
| Nat0 : expr
| NatS : expr -> expr
| NatMul : expr -> expr -> expr
| Var : var -> expr
| LetIn : expr -> (var -> expr) -> expr.
Definition Let_In {A B} (v : A) (f : A -> B) := let x := v in f x.
Notation "'dlet' x := v 'in' f" := (Let_In v (fun x => f)).
Notation "'elet' x := v 'in' f" := (LetIn v (fun x => f)).
Fixpoint denote (t : @expr nat) : nat
:= match t with
  | Nat0 => 0
  | NatS x => S (denote x)
  | NatMul x y => denote x * denote y
  | Var v => v
  | LetIn v f => dlet x := denote v in denote (f x)
end.
Fixpoint check_is_even_expr (t : @expr bool) : bool
:= match t with
  | Nat0 => true
  | NatS x => negb (check_is_even_expr x)
  | NatMul x y => orb (check_is_even_expr x) (check_is_even_expr y)
  | Var v_even => v_even
  | LetIn v f => let v_even := check_is_even_expr v in
    check_is_even_expr (f v_even)
end.

```

Note, importantly, that `check_is_even_expr` and `denote` take `exprs` with *different* instantiations of the `var` parameter. This is necessary so that we can store the information about whether or not a particular `let`-bound expression is even (or what its denotation is) in the variable node itself. However, this means that we cannot

reuse the same expression as arguments to both functions to formulate the soundness condition. Instead, we must introduce a notion of *relatedness* of expressions with different instantiations of the `var` parameter.

Such a relatedness predicate has one constructor for each constructor of `expr`, essentially encoding that the two expressions have the same structure. For the `Var` case, we defer to membership in a list of “related” variables, which we extend each time we go underneath a binder.

```
Inductive related {var1 var2 : Type}
  : list (var1 * var2) -> @expr var1 -> @expr var2 -> Prop :=
| RelatedNat0 {Γ}
  : related Γ Nat0 Nat0
| RelatedNatS {Γ e1 e2}
  : related Γ e1 e2 -> related Γ (NatS e1) (NatS e2)
| RelatedNatMul {Γ x1 x2 y1 y2}
  : related Γ x1 x2 -> related Γ y1 y2 -> related Γ (NatMul x1 y1) (NatMul x2 y2)
| RelatedVar {Γ v1 v2}
  : (v1, v2) Γ -> related Γ (Var v1) (Var v2)
| RelatedLetIn {Γ e1 e2 f1 f2}
  : related Γ e1 e2 -> (∀ v1 v2, related ((v1, v2) :: Γ) (f1 v1) (f2 v2))
  -> related Γ (LetIn e1 f1) (LetIn e2 f2).
```

Conventionally, syntax trees are parametric over the value of the `var` parameter, and we require that all instantiations give related ASTs (in the empty context), whence we call the parametric AST *well-formed*:

```
Definition Expr := ∀ var, @expr var.
Definition Wf (e : Expr) := ∀ var1 var2, related [] (e var1) (e var2)
```

We could then prove a modified form of our soundness theorem:

```
Theorem check_is_even_expr_sound (e : Expr) (H : Wf e)
  : check_is_even_expr (e bool) = true -> is_even (denote (e nat)).
```

To complete the picture, we would need a tactic `Reify` which took in a term of type `nat` and gave back a term of type `forall var, @expr var`, plus a tactic `prove_wf` which solved a goal of the form `Wf e` by repeated application of constructors. Given these, we could solve an evenness goal by writing¹

¹Note that for the `refine` to be fast, we must issue something like `Strategy -10 [denote]` to tell Coq to unfold `denote` before `Let_In`.

```

match goal with
| [ |- is_even ?v ]
  => let e := Reify v in
    refine (check_is_even_expr_sound e _ _);
    [ prove_wf | vm_compute; reflexivity ]
end.

```

Multiple Types

One important point, not yet mentioned, is that sometimes we want our reflective language to handle multiple types of terms. For example, we might want to enrich our language of expressions with lists. Since expressions like “take the successor of this list” don’t make sense, the natural choice is to index the inductive over codes for types.

We might write:

```

Inductive type := Nat | List (_ : type).
Inductive expr {var : type -> Type} : type -> Type :=
| Nat0 : expr Nat
| NatS : expr Nat -> expr Nat
| NatMul : expr Nat -> expr Nat -> expr Nat
| Var {t} : var t -> expr t
| LetIn {t1 t2} : expr t1 -> (var t1 -> expr t2) -> expr t2
| Nil {t} : expr (List t)
| Cons {t} : expr t -> expr (List t) -> expr (List t)
| Length {t} : expr (List t) -> expr Nat.

```

We would then have to adjust the definitions of the other functions accordingly. The type signatures of the might functions become

```

Fixpoint denote_type (t : type) : Type
:= match t with
  | Nat => nat
  | List t => list (denote_type t)
end.

Fixpoint even_data_of_type (t : type) : Type
:= match t with
  | Nat => bool (* is the nat even or not? *)
  | List t => list (even_data_of_type t)
end.

Fixpoint denote {t} (e : @expr denote_type t) : denote_type t.
Fixpoint check_is_even_expr {t} (e : @expr even_data_of_type t)

```

```

: even_data_of_type t.
Inductive related {var1 var2 : type -> Type}
  : list { t : type & var1 t * var2 t} ->  $\forall \{t\}, \text{@expr var1 } t \rightarrow \text{@expr var2 } t \rightarrow \text{Prop}$ 
Definition Expr (t : type) :=  $\forall \text{ var, } \text{@expr var } t.$ 
Definition Wf {t} (e : Expr t)
  :=  $\forall \text{ var1 var2, related } [] \text{ (e var1) (e var2).}$ 

```

See, e.g., [50] for a fuller treatment.

de Bruijn Indices

The idea behind *de Bruijn indices* is that variables are encoded by numbers which count up starting from the nearest enclosing binder. We might write

```

Inductive expr :=
| Nat0 : expr
| NatS : expr -> expr
| NatMul : expr -> expr -> expr
| Var : nat -> expr
| LetIn : expr -> expr -> expr.
Fixpoint denote (default : nat) ( $\Gamma$  : list nat) (t : @expr nat) : nat
  := match t with
    | Nat0 => 0
    | NatS x => S (denote default  $\Gamma$  x)
    | NatMul x y => denote default  $\Gamma$  x * denote default  $\Gamma$  y
    | Var idx => nth_default default  $\Gamma$  idx
    | LetIn v f => dlet x := denote default  $\Gamma$  v in
      denote default (x ::  $\Gamma$ ) f
  end.

```

If we wanted a more efficient representation, we could choose better data-structures for the context Γ and variable indices than linked lists and unary-encoded natural numbers. One particularly convenient choice, in Coq, would be using the efficient `PositiveMap.t` data-structure which encodes a finite map of binary-encoded `positives` to any type.

One unfortunate result is that the natural denotation function is no longer total. Here we have chosen to give a denotation function which returns a default element when a variable reference is too large, but we could instead choose to return an `option nat`. In general, however, returning an optional result significantly complicates the denotation function when binders are involved, because the types `A -> option B` and `option (A -> B)` are not isomorphic. On the other hand, requiring a default denotation prevents syntax trees from being able to represent possibly empty types.

[**QUESTION FOR ADAM:** Is there a good reference for these sorts of issues? Are they well-known? Well-studied?]

This causes further problems when dealing with an AST type which can represent terms of multiple types. In that case, we might annotate each variable node with a type code, mandate decidable equality of type codes, and then during denotation, we'd check the type of the variable node with the type of the corresponding variable in the context.

Nested Abstract Syntax

If we want a variant of de Bruijn indices which guarantees well-typed syntax trees, we can use nested abstract syntax. [95, 32] On mono-typed ASTs, this looks like encoding the size of the context in the type of the expressions. For example, we could use `option` types: [95]

```
Notation " $\hat{V}$ " := (option V).
Inductive expr : Type -> Type :=
| Nat0 {V} : expr V
| NatS {V} : expr V -> expr V
| NatMul {V} : expr V -> expr V -> expr V
| Var {V} : V -> expr V
| LetIn {V} : expr V -> expr ( $\hat{V}$ ) -> expr V.
```

This may seem a bit strange to those accustomed to encodings of terms in proof assistants, but it generalizes to a quite familiar intrinsic encoding of dependent type theory using types, contexts, and terms. [29] Namely, when the expressions are multi-typed, we end up with something like

```
Inductive context :=
| emp : context
| push : type -> context -> context.
Inductive var : context -> type -> Type :=
| Var0 {t Γ} : var (push t Γ) t
| VarS {t t' Γ} : var Γ t -> var (push t' Γ) t.
Inductive expr : context -> type -> Type :=
| Nat0 {Γ} : expr Γ Nat
| NatS {Γ} : expr Γ Nat -> expr Γ Nat
| NatMul {Γ} : expr Γ Nat -> expr Γ Nat -> expr Γ Nat
| Var {t Γ} : var Γ t -> expr Γ t
| LetIn {Γ t1 t2} : expr Γ t1 -> expr (push t1 Γ) t2 -> expr Γ t2.
```

Note that this generalizes nicely to codes for dependent types if the proof assistant supports induction-induction.

Although this representation enjoys both decidable equality of binders (like de Bruijn indices), as well as being well-typed-by-construction (like PHOAS), it's unfortunately unfit for coding algorithms that need to scale without massive assistance from the proof assistant. In particular, the naïve encoding of this inductive datatype incurs a quadratic overhead in representing terms involving binders, because each node stores the entire context. It is possible in theory to avoid this blowup by dropping the indices of the inductive type from the runtime representation. [39] One way to simulate this in Coq would be to put `context` in `Prop` and then extract the code to OCaml, which erases the `Props`. Alternatively, if Coq is extended with support for dropping irrelevant subterms [76] from the term representation, then this speedup could be accomplished even inside Coq.

Nominal

Nominal representations [159] use names rather than indices for binders. These representations have the benefit of being more human-readable, but require reasoning about freshness of names and capture-avoiding substitution. Additionally, if the representation of names is not sufficiently compact, the overhead of storing names at every binder node can become significant.

Locally Nameless

We mention the locally nameless representation [47, 121] because it is the term representation used by Coq itself. This representation uses de Bruijn indices for closed terms, and names for variables which are not bound in the current term.

Much like nominal representations, locally nameless representations also incur the overhead of generating and storing names. Naïve algorithms for generating fresh names, such as the algorithm used in Coq, can easily incur overhead that's linear in the size of the context. Generating n fresh names then incurs $\Theta(n^2)$ overhead. [TODO: explain what evar substitutions are somewhere? reference it here?] Additionally, using a locally nameless representation requires that evar substitutions be named. [TODO: elaborate more on the performance bottlenecks of named evar substitutions] See also ??.

4.1.4 Performance of Proving Reflective Well-Formedness of PHOAS

We saw in Section 4.1.3 that in order to prove the soundness theorem, we needed a way to relate two PHOASTs, which generalized to a notion of well-formedness for the `Expr` type.

Unfortunately, the proof that two `exprs` are `related` is quadratic in the size of the expression, for much the same reason that proving conjunctions in Subsection 2.2.1

resulted in a proof term which was quadratic in the number of conjuncts. We present two ways to encode linearly-sized proofs of well-formedness in PHOAS.

Iterating Reflection

The first method of encoding linearly-sized proofs of `related` is itself a good study in how using proof by reflection can compress proof terms. Rather than constructing the inductive `related` proof, we can instead write a fixed point:

```
Fixpoint is_related {var1 var2 : Type} (Γ : list (var1 * var2))
  (e1 : @expr var1) (e2 : @expr var2) : Prop :=
  match e1, e2 with
  | Nat0, Nat0 => True
  | NatS e1, NatS e2 => is_related Γ e1 e2
  | NatMul x1 y1, NatMul x2 y2
    => is_related Γ x1 x2 /\ is_related Γ y1 y2
  | Var v1, Var v2 => List.In (v1, v2) Γ
  | LetIn e1 f1, LetIn e2 f2
    => is_related Γ e1 e2 /\ forall v1 v2, is_related ((v1, v2) :: Γ) (f1 v1) (f2 v2)
  | _, _ => False
end.
```

This unfortunately isn't quite linear in the size of the syntax tree, though it is significantly smaller. One way to achieve truly linear² proofs is to pick a more optimized representation for list membership and to convert the proposition to be an eliminator. This consists of replacing $A \wedge B$ with $\forall P, A \rightarrow B \rightarrow P$, and similar.

```
Fixpoint is_related_elim {var1 var2 : Type} (Γ : list (var1 * var2))
  (e1 : @expr var1) (e2 : @expr var2) : Prop :=
  match e1, e2 with
  | Nat0, Nat0 => True
  | NatS e1, NatS e2 => is_related_elim Γ e1 e2
  | NatMul x1 y1, NatMul x2 y2 => forall P : Prop,
    (is_related_elim Γ x1 x2 -> is_related_elim Γ y1 y2 -> P) -> P
  | Var v1, Var v2 => forall (P : Prop),
    (forall n, List.nth_error Γ (N.to_nat n) = Some (v1, v2) -> P) -> P
  | LetIn e1 f1, LetIn e2 f2 => forall P : Prop,
    (is_related_elim Γ e1 e2
      -> (forall v1 v2, is_related_elim ((v1, v2) :: Γ) (f1 v1) (f2 v2))
      -> P)
    -> P
```

²Actually, the size of the proof term will still have an extra logarithmic factor in the size of the syntax tree, due to the way we represent list membership proofs.

```

| _, _ => False
end.
```

We can now prove that `is_related_elim Γ e1 e2 \rightarrow is_related Γ e1 e2`.

Note that making use of the fixpoint is significantly more inconvenient than making use of the inductive; the proof of `check_is_even_expr_sound`, for example, proceeds most naturally by induction on the relatedness hypothesis. We could instead induct on one of the ASTs and destruct the other one, but this becomes quite hairy when the ASTs are indexed over their types.

[**TODO:** Should I benchmark this and insert graphs?]

Via de Bruijn

An alternative, ultimately superior, method of constructing compact proofs of relatedness involves a translation to a de Bruijn representation. We can define a boolean predicate on de Bruijn syntax representing well-formedness.

```

Fixpoint is_closed_under (max_idx : nat) (e : expr) : bool :=
  match expr with
  | Nat0 => true
  | NatS e => is_closed_under max_idx e
  | NatMul x y => is_closed_under max_idx x && is_closed_under max_idx y
  | Var n => n <? max_idx
  | LetIn v f => is_closed_under max_idx v && is_closed_under (S max_idx) f
  end.
Definition is_closed := is_closed_under 0.
```

Note that this check generalizes quite nicely to expressions indexed over their types—so long as type codes have decidable equality—where we can pass around a list (or more efficient map structure) of types for each variable, and just check that the types are equal.

Now we can prove that whenever a de Bruijn `expr` is closed, any two PHOAS `exprs` created from that AST will be related in the empty context. Therefore, if the PHOAS `expr` we start off with is the result of converting some de Bruijn `expr` to PHOAS, we can easily prove that it's well-formed simply by running `vm_compute` on the `is_closed` procedure. How might we get such a de Bruijn `expr`? The easiest way is to write a converter from PHOAS to de Bruijn.

Hence we can prove the theorem $\forall e, \text{is_closed}(\text{PHOAS_to_deBruijn } e) = \text{true} \wedge e = \text{deBruijn_to_PHOAS}(\text{PHOAS_to_deBruijn } e) \rightarrow \text{Wf } e$. The hypothesis of

this theorem is quite easy to check; we simply run `vm_compute` and then instantiate it with the proof term `conj (eq_refl true) (eq_refl e)`, which is linear in the size of `e`.

4.2 Reification

The one part of proof by reflection that we've neglected up to this point is reification. There are many ways of performing reification; in Chapter 7, we discuss 18 different ways of implementing reification, using 6 different metaprogramming facilities in the Coq ecosystem: \mathcal{L}_{tac} , Ltac2, Mtac [82, 102], type classes [181], canonical structures [80], and reification-specific OCaml plugins (quote [58], template-coq [16], ours). Figure 7-3 displays the simplest case: an Ltac script to reify a tree of function applications and constants. Unfortunately, all methods we surveyed become drastically more complicated or slower (and usually both) when adapted to reify terms with variable bindings such as `let-in` or λ nodes.

We have made detailed walkthroughs and source code of these implementations available³ in hope that they will be useful for others considering implementing reification using one of these metaprogramming mechanisms, instructive as nontrivial examples of multiple metaprogramming facilities, or helpful as a case study in Coq performance engineering.

[TODO: Maybe include detailed walkthroughs in the appendix?] However, we do *not* recommend reading these out of general interest: most complexity in the described implementations strikes us as needless, with significant aspects of the design being driven by surprising behaviors, misfeatures, bugs, and performance bottlenecks of the underlying machinery as opposed to the task of reification.

```
Ltac f v x := (* reify var term *)
lazymatch x with
| 0 => constr:(@Nat0 v)
| S ?x => let X := f v x in
            constr:(@NatS v X)
| ?x*x?y => let X := f v x in
              let Y := f v y in
              constr:(@NatMul v X Y)
```

Figure 4.3: Reification Without Binders in \mathcal{L}_{tac}

[TODO: should this paragraph be here, or above the last one?] There are a couple of complications that arise when reifying binders, which broadly fall into two categories. One category is the metaprogramming language's treatment of binders. In \mathcal{L}_{tac} , for example, the body of a function is not a well-typed term, because the variable binder refers to a non-existent name; getting the name to actually refer to something, so that we can inspect the term, is responsible for a great deal of the complexity in reification code in \mathcal{L}_{tac} . The other category is any mismatch between the representation of binders in the metaprogramming language, and the representation of binders in the reified syntax. If the metaprogramming language represents variables as de Bruijn indices, and we are reifying to a de Bruijn representation, then we can reuse the indices. If the metaprogramming language represents variables as names, and we are reifying to a named representation, then we can reuse the names. If the rep-

³<https://github.com/mit-plv/reification-by-parametricity>

resentations mismatch, then we need to do extra work to align the representations, such as keeping some sort of finite map structure from binders in the metalanguage to binders in the AST.

[TODO: what's a good way to end this chapter?]

4.A Reflective Evenness Checking in Coq

[TODO: fix input of this file]

Chapter 5

A Framework for Building Verified Partial Evaluators

[TODO: update this chapter with camera-ready file before editing]

Abstract

Partial evaluation is a classic technique for generating lean, customized code from libraries that start with more bells and whistles. It is also an attractive approach to creation of *formally verified* systems, where theorems can be proved about libraries, yielding correctness of all specializations “for free.” However, it can be challenging to make library specialization both performant and trustworthy. We present a new approach, prototyped in the Coq proof assistant, which supports specialization at the speed of native-code execution, without adding to the trusted code base. Our extensible engine, which combines the traditional concepts of tailored term reduction and automatic rewriting from hint databases, is also of interest to replace these ingredients in proof assistants’ proof checkers and tactic engines, at the same time as it supports extraction to standalone compilers from library parameters to specialized code.

5.1 Introduction

Mechanized proof is gaining in importance for development of critical software infrastructure. Oft-cited examples include the CompCert verified C compiler [120] and the seL4 verified operating-system microkernel [105]. Here we have very flexible systems that are ready to adapt to varieties of workloads, be they C source programs for CompCert or application binaries for seL4. For a verified operating system, such adaptation takes place at *runtime*, when we launch the application. However, some important bits of software infrastructure commonly do adaptation at *compile time*,

such that the fully general infrastructure software is not even installed in a deployed system.

Of course, compilers are a natural example of that pattern, as we would not expect CompCert itself to be installed on an embedded system whose application code was compiled with it. The problem is that writing a compiler is rather labor-intensive, with its crafting of syntax-tree types for source, target, and intermediate languages, its fine-tuning of code for transformation passes that manipulate syntax trees explicitly, and so on. An appealing alternative is *partial evaluation* [101], which relies on reusable compiler facilities to specialize library code to parameters, with no need to write that library code in terms of syntax-tree manipulations. Cutting-edge tools in this tradition even make it possible to use high-level functional languages to generate performance-competitive low-level code, as in Scala’s Lightweight Modular Staging [167].

It is natural to try to port this approach to construction of systems with mechanized proofs. On one hand, the typed functional languages in popular proof assistants’ logics make excellent hosts for flexible libraries, which can often be specialized through means as simple as partial application of curried functions. Term-reduction systems built into the proof assistants can then generate the lean residual programs. On the other hand, it is surprisingly difficult to realize the last sentence with good performance. The challenge is that we are not just implementing algorithms; we also want a proof to be checked by a small proof checker, and there is tension in designing such a checker, as fancier reduction strategies grow the trusted code base. It would seem like an abandonment of the spirit of proof assistants to bake in a reduction strategy per library, yet effective partial evaluation tends to be rather fine-tuned in this way. Performance tuning matters when generated code is thousands of lines long.

In this paper, we present an approach to verified partial evaluation in proof assistants, which requires no changes to proof checkers. To make the relevance concrete, we use the example of Fiat Cryptography [70], a Coq library that generates code for big-integer modular arithmetic at the heart of elliptic-curve cryptography algorithms. This domain-specific compiler has been adopted, for instance, in the Chrome Web browser, such that about half of all HTTPS connections from browsers are now initiated using code generated (with proof) by Fiat Cryptography. However, Fiat Cryptography was only used successfully to build C code for the two most widely used curves (P-256 and Curve25519). Their method of partial evaluation timed out trying to compile code for the third most widely used curve (P-384). Additionally, to achieve acceptable reduction performance, the library code had to be written manually in continuation-passing style. We will demonstrate a new Coq library that corrects both weaknesses, while maintaining the generality afforded by allowing rewrite rules to be mixed with partial evaluation.

5.1.1 A Motivating Example

We are interested in partial-evaluation examples that mix higher-order functions, inductive datatypes, and arithmetic simplification. For instance, consider the following Coq code.

```
Definition prefixSums (ls:list nat) : list nat :=
  let ls' := combine ls (seq 0 (length ls)) in
  let ls'' := map ( $\lambda$  p, fst p * snd p) ls' in
  let '(_, ls''') := fold_left ( $\lambda$  acc_ls''' n,
    let '(acc, ls''') := acc_ls''' in
    let acc' := acc + n in
    (acc', acc' :: ls''')) ls'' (0, []) in
  ls'''.
```

This function first computes list `ls'` that pairs each element of input list `ls` with its position, so, for instance, list $[a; b; c]$ becomes $[(a, 0); (b, 1); (c, 2)]$. Then we map over the list of pairs, multiplying the components at each position. Finally, we traverse that list, building up a list of all prefix sums.

We would like to specialize this function to particular list lengths. That is, we know in advance how many list elements we will pass in, but we do not know the values of those elements. For a given length, we can construct a schematic list with one free variable per element. For example, to specialize to length four, we can apply the function to list $[a; b; c; d]$, and we expect this output:

```
let acc := b + c * 2 in
let acc' := acc + d * 3 in
[acc'; acc; b; 0]
```

Notice how subterm sharing via `lets` is important. As list length grows, we avoid quadratic blowup in term size through sharing. Also notice how we simplified the first two multiplications with $a \cdot 0 = 0$ and $b \cdot 1 = b$ (each of which requires explicit proof in Coq), using other arithmetic identities to avoid introducing new variables for the first two prefix sums of `ls''`, as they are themselves constants or variables, after simplification.

To set up our partial evaluator, we prove the algebraic laws that it should use for simplification, starting with basic arithmetic identities.

```
Lemma zero_plus : forall n, 0 + n = n.
Lemma plus_zero : forall n, n + 0 = n.
Lemma times_zero : forall n, n * 0 = 0.
Lemma times_one : forall n, n * 1 = n.
```

Next, we prove a law for each list-related function, connecting it to the primitive-recursion combinator for some inductive type (natural numbers or lists, as appropriate). We use a special apostrophe marker to indicate a quantified variable that may only match with *compile-time constants*. We also use a further marker `ident.eagerly` to ask the reducer to simplify a case of primitive recursion by complete traversal of the designated argument's constructor tree.

```

Lemma eval_map A B (f : A -> B) l
: map f l = ident.eagerly list_rect _ _ []
  (λ x _ l', f x :: l') l.
Lemma eval_fold_left A B (f : A -> B -> A) l a
: fold_left f l a = ident.eagerly list_rect
  _ _ (λ a, a)
  (λ x _ r a, r (f a x)) l a.
Lemma eval_combine A B (la : list A) (lb : list B)
: combine la lb = list_rect _ (λ _, [])
  (λ x _ r lb, list_case (λ _, _) [] []
    (λ y ys, (x, y) :: r ys) lb) la lb.
Lemma eval_length A (ls : list A)
: length ls = list_rect _ 0 (λ _ _ n, S n) ls.

```

With all the lemmas available, we can package them up into a rewriter, which triggers generation of a specialized rewrite procedure and its soundness proof. Our Coq plugin introduces a new command `Make` for building rewriters

```

Make rewriter := Rewriter For (zero_plus, plus_zero,
  times_zero, times_one, eval_map, eval_fold_left,
  do_again eval_length, do_again eval_combine,
  eval_rect nat, eval_rect list, eval_rect prod)
  (with delta) (with extra idents (seq)).

```

Most inputs to `Rewriter For` list quantified equalities to use for left-to-right rewriting. However, we also use options `do_again`, to request that some rules trigger an extra bottom-up pass after being used for rewriting; `eval_rect`, to queue up eager evaluation of a call to a primitive-recursion combinator on a known recursive argument; `with delta`, to request evaluation of all monomorphic operations on concrete inputs; and `with extra idents`, to inform the engine of further permitted identifiers that do not appear directly in any of the rewrite rules.

Our plugin also provides new tactics like `Rewrite_rhs_for`, which applies a rewriter to the righthand side of an equality goal. That last tactic is just what we need to synthesize a specialized `prefixSums` for list length four, along with a proof of its equivalence to the original function.

```

Definition prefixSums4 :
{f : nat -> nat -> nat -> nat -> list nat
| forall a b c d, f a b c d = prefixSums [a;b;c;d]} :=
ltac:(eexists; Rewrite_rhs_for rewriter; reflexivity).

```

5.1.2 Concerns of Trusted-Code-Base Size

Crafting a reduction strategy is challenging enough in a standalone tool. A large part of the difficulty in a proof assistant is reducing in a way that leaves a proof trail that can be checked efficiently by a small kernel. Most proof assistants present user-friendly surface tactic languages that generate proof traces in terms of more elementary tactic steps. The trusted proof checker only needs to know about the elementary steps, and there is pressure to be sure that these steps are indeed elementary, not requiring excessive amounts of kernel code. However, hardcoding a new reduction strategy in the kernel can bring dramatic performance improvements. Generating thousands of lines of code with partial evaluation would be intractable if we were outputting sequences of primitive rewrite steps justifying every little term manipulation, so we must take advantage of the time-honored feature of type-theoretic proof assistants that reductions included in the definitional equality need not be requested explicitly.

Which kernel-level reductions *does* Coq support today? Currently, the trusted code base knows about four different kinds of reduction: left-to-right conversion, right-to-left conversion, a virtual machine (VM) written in C based on the OCaml compiler, and a compiler to native code. Furthermore, the first two are parameterized on an arbitrary user-specified ordering of which constants to unfold when, in addition to internal heuristics about what to do when the user has not specified an unfolding order for given constants. Recently, native support for 63-bit integers has been added to the VM and native machines. A recent pull request proposes adding support for native IEEE 754-2008 binary64 floats [132], and support for native arrays is in the works [65].

To summarize, there has been quite a lot of “complexity creep” in the Coq trusted base, to support efficient reduction, and yet realistic partial evaluation has *still* been rather challenging. Even the additional three reduction mechanisms outside Coq’s kernel (`cbn`, `simpl`, `cgv`) are not at first glance sufficient for verified partial evaluation.

5.1.3 Our Solution

Aehlig, Haftmann, and Nipkow [5] presented a very relevant solution to a related problem, using *normalization by evaluation (NbE)* [30] to bootstrap reduction of open terms on top of full reduction, as built into a proof assistant. However, it was simultaneously true that they expanded the proof-assistant trusted code base in ways specific to their technique, and that they did not report any experiments actually using the tool for partial evaluation (just traditional full reduction), potentially hiding performance-scaling challenges or other practical issues. We have adapted their

approach in a new Coq library embodying **the first partial-evaluation approach to satisfy the following criteria**.

- It integrates with a general-purpose, foundational proof assistant, **without growing the trusted base**.
- For a wide variety of initial functional programs, it provides **fast** partial evaluation with reasonable memory use.
- It allows reduction that **mixes rules of the definitional equality with equalities proven explicitly as theorems**.
- It **preserves sharing** of common subterms.
- It also allows **extraction of standalone partial evaluators**.

Our contributions include answers to a number of challenges that arise in scaling NbE-based partial evaluation in a proof assistant. First, we rework the approach of Aehlig, Haftmann, and Nipkow [5] to function *without extending a proof assistant’s trusted code base*, which, among other challenges, requires us to prove termination of reduction and encode pattern matching explicitly (leading us to adopt the performance-tuned approach of Maranget [131]).

Second, using partial evaluation to generate residual terms thousands of lines long raises *new scaling challenges*:

- Output terms may contain so *many nested variable binders* that we expect it to be performance-prohibitive to perform bookkeeping operations on first-order-encoded terms (e.g., with de Bruijn indices, as is done in \mathcal{R}_{tac} by Malecha and Bengtson [128]). For instance, while the reported performance experiments of Aehlig, Haftmann, and Nipkow [5] generate only closed terms with no binders, Fiat Cryptography may generate a single routine (e.g., multiplication for curve P-384) with nearly a thousand nested binders.
- Naive representation of terms without proper *sharing of common subterms* can lead to fatal term-size blow-up. Fiat Cryptography’s arithmetic routines rely on significant sharing of this kind.
- Unconditional rewrite rules are in general insufficient, and we need *rules with side conditions*. For instance, in Fiat Cryptography, some rules for simplifying modular arithmetic depend on proofs that operations in subterms do not overflow.
- However, it is also not reasonable to expect a general engine to discharge all side conditions on the spot. We need integration with *abstract interpretation* that can analyze whole programs to support reduction.

Briefly, our respective solutions to these problems are the *parametric higher-order abstract syntax (PHOAS)* [50] term encoding, a *let-lifting* transformation threaded throughout reduction, extension of rewrite rules with executable Boolean side conditions, and a design pattern that uses decorator function calls to include analysis results in a program.

Finally, we carry out the *first large-scale performance-scaling evaluation* of partial evaluation in a proof assistant, covering all elliptic curves from the published Fiat Cryptography experiments, along with microbenchmarks.

This paper proceeds through explanations of the trust stories behind our approach and earlier ones (Section 5.2), the core structure of our engine (Section 5.3), the additional scaling challenges we faced (Section 5.4), performance experiments (Section 5.5), and related work (Section 5.6) and conclusions. Our implementation is included as an anonymous supplement.

5.2 Trust, Reduction, and Rewriting

Since much of the narrative behind our design process depends on tradeoffs between performance and trustworthiness, we start by reviewing the general situation in proof assistants.

Across a variety of proof assistants, simplification of functional programs is a workhorse operation. Proof assistants like Coq that are based on type theory typically build in *definitional equality* relations, identifying terms up to reductions like β -reduction and unfolding of named identifiers. What looks like a single “obvious” step in an on-paper equational proof may require many of these reductions, so it is handy to have built-in support for checking a claimed reduction. 5-1a diagrams how such steps work in a system like Coq, where the system implementation is divided between a trusted *kernel*, for checking *proof terms* in a minimal language, and additional untrusted support, like a *tactic* engine evaluating a language of higher-level proof steps, in the process generating proof terms out of simpler building blocks. It is standard to include a primitive proof step that validates any reduction compatible with the definitional equality, as the latter is decidable. The figure shows a tactic that simplifies a goal using that facility.

In proof goals containing free variables, executing subterms can get stuck before reaching normal forms. However, we can often achieve further simplification by using equational rules that we prove explicitly, rather than just relying on the rules built into the definitional equality and its decidable equivalence checker. Coq’s `autorewrite` tactic, as diagrammed in 5-1b, is a good example: it takes in a database of quantified equalities and applies them repeatedly to rewrite in a goal. It is important that Coq’s kernel does not trust the `autorewrite` tactic. Instead, the tactic must output a proof term that, in some sense, is the moral equivalent of a line-by-line equational proof.

It can be challenging to keep these proof terms small enough, as naive rewrite-by-rewrite versions repeatedly copy large parts of proof goals, justifying a rewrite like $C[e_1] = C[e_2]$ for some context C given a proof of $e_1 = e_2$, with the full value of C replicated in the proof term for that single rewrite. Overcoming these challenges while retaining decidability of proof checking is tricky, since we may use `autorewrite` with rule sets that do not always lead to terminating reduction. Coq includes more experimental alternatives like `rewrite_strat`, which use bottom-up construction of multi-rewrite proofs, with sharing of common contexts. Still, as Section 5.5 will show, these methods that generate substantial proof terms are at significant performance disadvantages.

Now we summarize how Aehlig, Haftmann, and Nipkow [5] provide flexible and fast interleaving of standard λ -calculus reduction and use of proved equalities (the next section will go into more detail). 5-1c demonstrates a workflow based on *a deep embedding of a core ML-like language*. That is, within the logic of the proof assistant (Isabelle/HOL, in their case), a type of syntax trees for ML programs is defined, with an associated operational semantics. The basic strategy is, for a particular set of rewrite rules and a particular term to simplify, to *generate a (deeply embedded) ML program that, if it terminates, produces a syntax tree for the simplified term*. Their tactic uses *reification* to create ML versions of rule sets and terms. They also wrote a reduction function in ML and proved it sound once and for all, against the ML operational semantics. Combining that proof with proofs generated by reification, we conclude that an application of the reduction function to the reified rules and term is indeed an ML term that generates correct answers. The tactic then “throws the ML term over the wall,” using a general code-generation framework for Isabelle/HOL [92]. Trusted code compiles the ML code into the concrete syntax of a mainstream ML language, Standard ML in their case, and compiles it with an off-the-shelf compiler. The output of that compiled program is then passed back over to the tactic, in terms of an axiomatic assertion that the ML semantics really yields that answer.

As Aehlig, Haftmann, and Nipkow [5] argue, their use of external compilation and evaluation of ML code adds no real complexity on top of that required by the proof assistant – after all, the proof assistant itself must be compiled and executed somehow. However, the perceived increase of trusted code base is not spurious: it is one thing to trust that the toolchain and execution environment used by the proof assistant and the partial evaluator are well-behaved, and another to rely on two descriptions of ML (one deeply embedded in the proof assistant and another implied by the compiler) to agree on every detail of the semantics. Furthermore, there still is new trusted code to translate from the deeply embedded ML subset into the concrete syntax of the full-scale ML language. The vast majority of proof-assistant developments today rely on no such embeddings with associated mechanized semantics, so need we really add one to a proof-checking kernel to support efficient partial evaluation?

Our answer, diagrammed in 5-1d, shows a different way. We still reify terms and rules into a deeply embedded language. However, *the reduction engine is implemented*

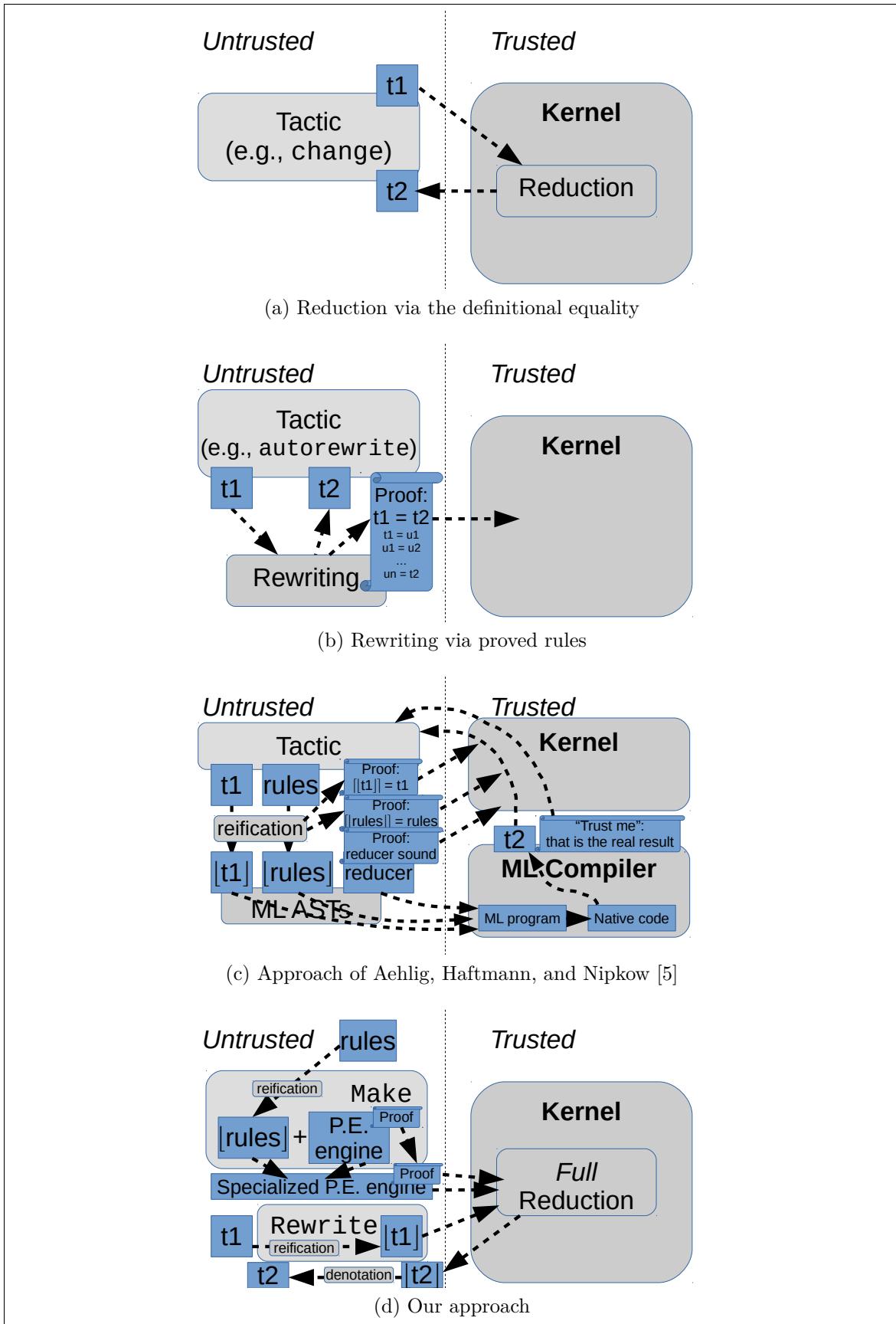


Figure 5-1: Different approaches to reduction and rewriting

directly in the logic, rather than as a deeply embedded syntax tree of an ML program. As a result, the kernel’s own reduction engine is prepared to execute our reduction engine for us – using an operation that would be included in a type-theoretic proof assistant in any case, with no special support for a language deep embedding. We also stage the process for performance reasons. First, the `Make` command creates a rewriter out of a list of rewrite rules, by specializing a generic partial-evaluation engine, which has a generic proof that applies to any set of proved rewrite rules. We perform partial evaluation on the specialized partial evaluator, using Coq’s normal reduction mechanisms, under the theory that we can afford to pay performance costs at this stage because we only need to create new rewriters relatively infrequently. Then individual rewritings involve reifying terms, asking the kernel to execute the specialized evaluator on them, and simplifying an application of an interpretation function to the result (this last step must be done using Coq’s normal reduction, and it is the bottleneck for outputs with enormous numbers of nested binders as discussed in section 5.5.1).

5.2.1 Our Approach in Nine Steps

Here is a bit more detail on the steps that go into applying our Coq plugin, many of which we expand on in the following sections. In order to build a precomputed rewriter with the `Make` command, the following actions are performed:

1. The given lemma statements are scraped for which named functions and types the rewriter package will support.
2. Inductive types enumerating all available primitive types and functions are emitted.
3. Tactics generate all of the necessary definitions and prove all of the necessary lemmas for dealing with this particular set of inductive codes. Definitions include operations like Boolean equality on type codes and lemmas like “all representable primitive types have decidable equality.”
4. The statements of rewrite rules are reified, and we prove soundness and syntactic-well-formedness lemmas about each of them. Each instance of the former involves wrapping the user-provided proof with the right adapter to apply to the reified version.
5. The definitions needed to perform reification and rewriting and the lemmas needed to prove correctness are assembled into a single package that can be passed by name to the rewriting tactic.

When we want to rewrite with a rewriter package in a goal, the following steps are performed:

1. We rearrange the goal into a single logical formula: all free-variable quantification in the proof context is replaced by changing the equality goal into an equality between two functions (taking the free variables as inputs).
2. We reify the side of the goal we want to simplify, using the inductive codes in the specified package. That side of the goal is then replaced with a call to a denotation function on the reified version.
3. We use a theorem stating that rewriting preserves denotations of well-formed terms to replace the denotation subterm with the denotation of the rewriter applied to the same reified term. We use Coq’s built-in full reduction (`vm_compute`) to reduce the application of the rewriter to the reified term.
4. Finally, we run `cbv` (a standard call-by-value reducer) to simplify away the invocation of the denotation function on the concrete syntax tree from rewriting.

5.3 The Structure of a Rewriter

We now simultaneously review the approach of Aehlig, Haftmann, and Nipkow [5] and introduce some notable differences in our own approach, noting similarities to the reflective rewriter of Malecha and Bengtson [128] where applicable.

First, let us describe the language of terms we support rewriting in. Note that, while we support rewriting in full-scale Coq proofs, where the metalanguage is dependently typed, the object language of our rewriter is nearly simply typed, with limited support for calling polymorphic functions. However, we still support identifiers whose definitions use dependent types, since our reducer does not need to look into definitions.

$$\begin{aligned} e ::= & \text{ App } e_1 \ e_2 \mid \text{Let } v := e_1 \ \text{In } e_2 \\ & \mid \text{Abs } (\lambda v. e) \mid \text{Var } v \mid \text{Ident } i \end{aligned}$$

The `Ident` case is for identifiers, which are described by an enumeration specific to a use of our library. For example, the identifiers might be codes for `+`, `:`, and literal constants. We write $\llbracket e \rrbracket$ for a standard denotational semantics.

5.3.1 Pattern-Matching Compilation and Evaluation

Aehlig, Haftmann, and Nipkow [5] feed a specific set of user-provided rewrite rules to their engine by generating code for an ML function, which takes in deeply embedded term syntax (actually *doubly* deeply embedded, within the syntax of the deeply embedded ML!) and uses ML pattern matching to decide which rule to apply at the top level. Thus, they delegate efficient implementation of pattern matching to the underlying ML implementation. As we instead build our rewriter in Coq’s logic, we have no such option to defer to ML. Indeed, Coq’s logic only includes primitive pattern-matching constructs to match one constructor at a time.

We could follow a naive strategy of repeatedly matching each subterm against a pattern for every rewrite rule, as in the rewriter of Malecha and Bengtson [128], but in that case we do a lot of duplicate work when rewrite rules use overlapping function symbols. Instead, we adopted the approach of Maranget [131], who describes compilation of pattern matches in OCaml to decision trees that eliminate needless repeated work (for example, decomposing an expression into $x + y + z$ only once even if two different rules match on that pattern). We have not yet implemented any of the optimizations described therein for finding *minimal* decision trees.

There are three steps to turn a set of rewrite rules into a functional program that takes in an expression and reduces according to the rules. The first step is pattern-matching compilation: we must compile the lefthand sides of the rewrite rules to a decision tree that describes how and in what order to decompose the expression, as well as describing which rewrite rules to try at which steps of decomposition. Because the decision tree is merely a decomposition hint, we require no proofs about it to ensure soundness of our rewriter. The second step is decision-tree evaluation, during which we decompose the expression as per the decision tree, selecting which rewrite rules to attempt. The only correctness lemma needed for this stage is that any result it returns is equivalent to picking some rewrite rule and rewriting with it. The third and final step is to actually rewrite with the chosen rule. Here the correctness condition is that we must not change the semantics of the expression. Said another way, any rewrite-rule replacement expression must match the semantics of the rewrite-rule pattern.

While pattern matching begins with comparing one pattern against one expression, Maranget’s approach works with intermediate goals that check multiple patterns against multiple expressions. A decision tree describes how to match a vector (or list) of patterns against a vector of expressions. It is built from these constructors:

- **TryLeaf k onfailure:** Try the k^{th} rewrite rule; if it fails, keep going with `onfailure`.
- **Failure:** Abort; nothing left to try.
- **Switch `icases` `app_case` `default`:** With the first element of the vector, match on its kind; if it is an identifier matching something in `icases`, remove the first element of the vector and run that decision tree; if it is an application and `app_case` is not `None`, try the `app_case` decision tree, replacing the first element of each vector with the two elements of the function and the argument it is applied to; otherwise, do not modify the vectors and use the `default` decision tree.
- **Swap i `cont`:** Swap the first element of the vector with the i^{th} element (0-indexed) and keep going with `cont`.

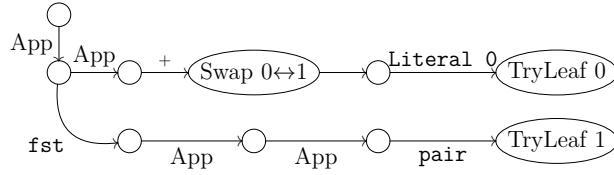
Consider the encoding of two simple example rewrite rules, where we follow Coq’s \mathcal{L}_{tac} language in prefacing pattern variables with question marks.

$$\begin{aligned} ?n + 0 &\rightarrow n \\ \text{fst}_{\mathbb{Z}, \mathbb{Z}}(?x, ?y) &\rightarrow x \end{aligned}$$

We embed them in an AST type for patterns, which largely follows our ASTs for expressions.

0. App (App (Ident +) Wildcard) (Ident (Literal 0))
1. App (Ident fst) (App (App (Ident pair) Wildcard) Wildcard)

The decision tree produced is



where every non-swap node implicitly has a “default” case arrow to `Failure`.

We implement, in Coq’s logic, an evaluator for these trees against terms. Note that we use Coq’s normal partial evaluation to turn our general decision-tree evaluator into a specialized matcher to get reasonable efficiency. Although this partial evaluation of our partial evaluator is subject to the same performance challenges we highlighted in the introduction, it only has to be done once for each set of rewrite rules, and we are targeting cases where the time of per-goal reduction dominates this time of meta-compilation.

For our running example of two rules, specializing gives us this match expression.

```

match e with
| App f y => match f with
| Ident fst => match y with
| App (App (Ident pair) x) y => x
| _ => e end
| App (Ident +) x => match y with
| Ident (Literal 0) => x | _ => e end
| _ => e end | _ => e end.
  
```

5.3.2 Adding Higher-Order Features

Fast rewriting at the top level of a term is the key ingredient for supporting customized algebraic simplification. However, not only do we want to rewrite throughout the

structure of a term, but we also want to integrate with simplification of higher-order terms, in a way where we can prove to Coq that our syntax-simplification function always terminates. Normalization by evaluation (NbE) [30] is an elegant technique for adding the latter aspect, in a way where we avoid needing to implement our own λ -term reducer or prove it terminating.

To orient expectations: we would like to enable the following reduction

$$(\lambda f x y. f x y) (+) z 0 \rightsquigarrow z$$

using the rewrite rule

$$?n + 0 \rightarrow n$$

Aehlig, Haftmann, and Nipkow [5] also use NbE, and we begin by reviewing its most classic variant, for performing full β -reduction in a simply typed term in a guaranteed-terminating way. The simply typed λ -calculus syntax we use is:

$$t ::= t \rightarrow t \mid b \quad e ::= \lambda v. e \mid e e \mid v \mid c$$

with v for variables, c for constants, and b for base types.

We can now define normalization by evaluation. First, we choose a “semantic” representation for each syntactic type, which serves as the result type of an intermediate interpreter.

$$\begin{aligned} \text{NbE}_t(t_1 \rightarrow t_2) &:= \text{NbE}_t(t_1) \rightarrow \text{NbE}_t(t_2) \\ \text{NbE}_t(b) &:= \text{expr}(b) \end{aligned}$$

Function types are handled as in a simple denotational semantics, while base types receive the perhaps-counterintuitive treatment that the result of “executing” one is a syntactic expression of the same type. We write $\text{expr}(b)$ for the metalanguage type of object-language syntax trees of type b , relying on a dependent type family expr .

Now the core of NbE, shown in Figure 5-2, is a pair of dual functions `reify` and `reflect`, for converting back and forth between syntax and semantics of the object language, defined by primitive recursion on type syntax. We split out analysis of term syntax in a separate function `reduce`, defined by primitive recursion on term syntax, when usually this functionality would be mixed in with `reflect`. The reason for this choice will become clear when we extend NbE to handle our full problem domain.

We write v for object-language variables and x for metalanguage (Coq) variables, and we overload λ notation using the metavariable kind to signal whether we are building a host λ or a λ syntax tree for the embedded language. The crucial first clause for `reduce` replaces object-language variable v with fresh metalanguage variable x , and

```

reifyt : NbEt(t) → expr(t)
reifyt1→t2(f) := λv. reifyt2(f(reflectt1(v)))
reifyb(f) := f

reflectt : expr(t) → NbEt(t)
reflectt1→t2(e) := λx. reflectt2(e(reifyt1(x)))
reflectb(e) := e

reduce : expr(t) → NbEt(t)
reduce(λv. e) := λx. reduce([x/v]e)
reduce(e1 e2) := (reduce(e1)) (reduce(e2))
reduce(x) := x
reduce(c) := reflect(c)

NbE : expr(t) → expr(t)
NbE(e) := reify(reduce(e))

```

Figure 5-2: Implementation of normalization by evaluation

then we are somehow tracking that all free variables in an argument to `reduce` must have been replaced with metalanguage variables by the time we reach them. We reveal in Subsection 5.4.1 the encoding decisions that make all the above legitimate, but first let us see how to integrate use of the rewriting operation from the previous section. To fuse NbE with rewriting, we only modify the constant case of `reduce`. First, we bind our specialized decision-tree engine under the name `rewrite-head`. Recall that this function only tries to apply rewrite rules at the top level of its input.

In the constant case, we still reflect the constant, but underneath the binders introduced by full η -expansion, we perform one instance of rewriting. In other words, we change this one function-definition clause:

$$\text{reflect}_b(e) := \text{rewrite-head}(e)$$

It is important to note that a constant of function type will be η -expanded only once for each syntactic occurrence in the starting term, though the expanded function is effectively a thunk, waiting to perform rewriting again each time it is called. From first principles, it is not clear why such a strategy terminates on all possible input terms, though we work up to convincing Coq of that fact.

The details so far are essentially the same as in the approach of Aehlig, Haftmann, and

Nipkow [5]. Recall that their rewriter was implemented in a deeply embedded ML, while ours is implemented in Coq’s logic, which enforces termination of all functions. Aehlig et al. did not prove termination, which indeed does not hold for their rewriter in general, which works with untyped terms, not to mention the possibility of rule-specific ML functions that diverge themselves. In contrast, we need to convince Coq up-front that our interleaved λ -term normalization and algebraic simplification always terminate. Additionally, we need to prove that our rewriter preserves denotations of terms, which can easily devolve into tedious binder bookkeeping, depending on encoding.

The next section introduces the techniques we use to avoid explicit termination proof or binder bookkeeping, in the context of a more general analysis of scaling challenges.

5.4 Scaling Challenges

Aehlig, Haftmann, and Nipkow [5] only evaluated their implementation against closed programs. What happens when we try to apply the approach to partial-evaluation problems that should generate thousands of lines of low-level code?

5.4.1 Variable Environments Will Be Large

We should think carefully about representation of ASTs, since many primitive operations on variables will run in the course of a single partial evaluation. For instance, Aehlig, Haftmann, and Nipkow [5] reported a significant performance improvement changing variable nodes from using strings to using de Bruijn indices [40]. However, de Bruijn indices and other first-order representations remain painful to work with. We often need to fix up indices in a term being substituted in a new context. Even looking up a variable in an environment tends to incur linear time overhead, thanks to traversal of a list. Perhaps we can do better with some kind of balanced-tree data structure, but there is a fundamental performance gap versus the arrays that can be used in imperative implementations. Unfortunately, it is difficult to integrate arrays soundly in a logic. Also, even ignoring performance overheads, tedious binder bookkeeping complicates proofs.

Our strategy is to use a variable encoding that pushes all first-order bookkeeping off on Coq’s kernel, which is itself performance-tuned with some crucial pieces of imperative code. Parametric higher-order abstract syntax (PHOAS) [50] is a dependently typed encoding of syntax where binders are managed by the enclosing type system. It allows for relatively easy implementation and proof for NbE, so we adopted it for our framework.

Here is the actual inductive definition of term syntax for our object language, PHOAS-style. The characteristic oddity is that the core syntax type `expr` is parameterized on a dependent type family for representing variables. However, the final representation

type `Expr` uses first-class polymorphism over choices of variable type, bootstrapping on the metalanguage's parametricity to ensure that a syntax tree is agnostic to variable type.

```

Inductive type := arrow (s d : type)
| base (b : base_type).
Infix "->" := arrow.
Inductive expr (var : type -> Type)
  : type -> Type :=
| Var {t} (v : var t) : expr var t
| Abs {s d} (f : var s -> expr var d)
  : expr var (s -> d)
| App {s d} (f : expr var (s -> d))
  (x : expr var s) : expr var d
| Const {t} (c : const t) : expr var t
Definition Expr (t : type) : Type :=
  forall var, expr var t.

```

A good example of encoding adequacy is assigning a simple denotational semantics. First, a simple recursive function assigns meanings to types.

```

Fixpoint denoteT (t : type) : Type
:= match t with
  | arrow s d => denoteT s -> denoteT d
  | base b      => denote_base_type b
end.

```

Next we see the convenience of being able to *use* an expression by choosing how it should represent variables. Specifically, it is natural to choose *the type-denotation function itself* as the variable representation. Especially note how this choice makes rigorous the convention we followed in the prior section, where a recursive function enforces that values have always been substituted for variables early enough.

```

Fixpoint denoteE {t} (e : expr denoteT t) : denoteT t
:= match e with
  | Var v      => v
  | Abs f      => λ x, denoteE (f x)
  | App f x   => (denoteE f) (denoteE x)
  | Ident c    => denoteI c
end.
Definition DenoteE {t} (E : Expr t) : denoteT t
:= denoteE (E denoteT).

```

It is now easy to follow the same script in making our rewriting-enabled NbE fully formal. Note especially the first clause of `reduce`, where we avoid variable substitution precisely because we have chosen to represent variables with normalized semantic values. The subtlety there is that base-type semantic values are themselves expression syntax trees, which depend on a nested choice of variable representation, which we retain as a parameter throughout these recursive functions. The final definition λ -quantifies over that choice.

```

Fixpoint nbeT var (t : type) : Type
:= match t with
| arrow s d => nbeT var s -> nbeT var d
| base b      => expr var b
end.

Fixpoint reify {var t} : nbeT var t -> expr var t
:= match t with
| arrow s d => λ f,
  Abs (λ x, reify (f (reflect (Var x))))
| base b      => λ e, e
end

with reflect {var t} : expr var t -> nbeT var t
:= match t with
| arrow s d => λ e,
  λ x, reflect (App e (reify x))
| base b      => rewrite_head
end.

Fixpoint reduce {var t}
(e : expr (nbeT var) t) : nbeT var t
:= match e with
| Abs e      => λ x, reduce (e (Var x))
| App e1 e2 => (reduce e1) (reduce e2)
| Var x     => x
| Ident c    => reflect (Ident c)
end.

Definition Rewrite {t} (E : Expr t) : Expr t
:= λ var, reify (reduce (E (nbeT var t))).
```

One subtlety hidden above in implicit arguments is in the final clause of `reduce`, where the two applications of the `Ident` constructor use different variable representations. With all those details hashed out, we can prove a pleasingly simple correctness theorem, with a lemma for each main definition, with inductive structure mirroring recursive structure of the definition, also appealing to correctness of last section's pattern-compilation operations.

$$\forall t, E : \text{Expr } t. \llbracket \text{Rewrite}(E) \rrbracket = \llbracket E \rrbracket$$

Even before getting to the correctness theorem, we needed to convince Coq that the function terminates. While for Aehlig, Haftmann, and Nipkow [5], a termination proof would have been a whole separate enterprise, it turns out that PHOAS and NbE line up so well that Coq accepts the above code with no additional termination proof. As a result, the Coq kernel is ready to run our `Rewrite` procedure during checking.

To understand how we now apply the soundness theorem in a tactic, it is important to note that the Coq kernel’s built-in reduction strategies have, to an extent, been tuned to work well to show equivalence between a simple denotational-semantics application and the semantic value it produces, while it is rather difficult to code up one reduction strategy that works well for all partial-evaluation tasks. Therefore, we should restrict ourselves to (1) running full reduction in the style of functional-language interpreters and (2) running normal reduction on “known-good” goals like correctness of evaluation of a denotational semantics on a concrete input.

Operationally, then, we apply our tactic in a goal containing a term e that we want to partially evaluate. In standard proof-by-reflection style, we *reify* e into some E where $\llbracket E \rrbracket = e$, replacing e accordingly, asking Coq’s kernel to validate the equivalence via standard reduction. Now we use the `Rewrite` correctness theorem to replace $\llbracket E \rrbracket$ with $\llbracket \text{Rewrite}(E) \rrbracket$. Next we may ask the Coq kernel to simplify $\text{Rewrite}(E)$ by *full reduction via compilation to native code*, since we carefully designed $\text{Rewrite}(E)$ and its dependencies to produce closed syntax trees. Finally, where E' is the result of that reduction, we simplify $\llbracket E' \rrbracket$ with standard reduction, producing a normal-looking Coq term.

5.4.2 Subterm Sharing is Crucial

For some large-scale partial-evaluation problems, it is important to represent output programs with sharing of common subterms. Redundantly inlining shared subterms can lead to exponential increase in space requirements. Consider the Fiat Cryptography [70] example of generating a 64-bit implementation of field arithmetic for the P-256 elliptic curve. The library has been converted manually to continuation-passing style, allowing proper generation of `let` binders, whose variables are often mentioned multiple times. We ran their code generator (actually just a subset of its functionality, but optimized by us a bit further, as explained in Subsection 5.5.2) on the P-256 example and found it took about 15 seconds to finish. Then we modified reduction to inline `let` binders instead of preserving them, at which point the reduction job terminated with an out-of-memory error, on a machine with 64 GB of RAM. (The successful run uses under 2 GB.)

We see a tension here between performance and niceness of library implementation. The Fiat Cryptography authors found it necessary to CPS-convert their code to coax Coq into adequate reduction performance. Then all of their correctness theorems were complicated by reasoning about continuations. It feels like a slippery slope on

the path to implementing a domain-specific compiler, rather than taking advantage of the pleasing simplicity of partial evaluation on natural functional programs. Our reduction engine takes shared-subterm preservation seriously while applying to libraries in direct style.

Our approach is **let**-lifting: we lift **lets** to top level, so that applications of functions to **lets** are available for rewriting. For example, we can perform the rewriting

$$\begin{aligned} & \text{map } (\lambda x. y + x) (\text{let } z := e \text{ in } [0; 1; 2; z; z + 1]) \\ \rightsquigarrow & \text{let } z := e \text{ in } [y; y + 1; y + 2; y + z; y + (z + 1)] \end{aligned}$$

using the rules

$$\begin{array}{ll} \text{map } ?f [] \rightarrow [] & ?n + 0 \rightarrow n \\ \text{map } ?f (?x :: ?xs) \rightarrow f x :: \text{map } f xs & \end{array}$$

Our approach is to define a telescope-style type family called **UnderLets**:

```
Inductive UnderLets {var} (T : Type) :=
| Base (v : T)
| UnderLet {A}(e : @expr var A)(f : var A -> UnderLets T).
```

A value of type **UnderLets** *T* is a series of **let** binders (where each expression *e* may mention earlier-bound variables) ending in a value of type *T*. It is easy to build various “smart constructors” working with this type, for instance to construct a function application by lifting the **lets** of both function and argument to a common top level.

Such constructors are used to implement an NbE strategy that outputs **UnderLets** telescopes. Recall that the NbE type interpretation mapped base types to expression syntax trees. We now parameterize that type interpretation by a Boolean declaring whether we want to introduce telescopes.

```
Fixpoint nbeT' {var} (with_lets : bool) (t : type)
:= match t with
| base t => if with_lets
            then @UnderLets var (@expr var t)
            else @expr var t
| arrow s d => nbeT' false s -> nbeT' true d
end.
Definition nbeT := nbeT' false.
Definition nbeT_with_lets := nbeT' true.
```

There are cases where naive preservation of **let** binders leads to suboptimal perfor-

mance, so we include some heuristics. For instance, when the expression being bound is a constant, we always inline. When the expression being bound is a series of list “cons” operations, we introduce a name for each individual list element, since such a list might be traversed multiple times in different ways.

5.4.3 Rules Need Side Conditions

Many useful algebraic simplifications require side conditions. One simple case is supporting *nonlinear* patterns, where a pattern variable appears multiple times. We can encode nonlinearity on top of linear patterns via side conditions.

$$?n_1 + ?m - ?n_2 \rightarrow m \text{ if } n_1 = n_2$$

The trouble is how to support predictable solving of side conditions during partial evaluation, where we may be rewriting in open terms. We decided to sidestep this problem by allowing side conditions only as executable Boolean functions, to be applied only to variables that are confirmed as *compile-time constants*, unlike Malecha and Bengtson [128] who support general unification variables. We added a variant of pattern variable that only matches constants. Semantically, this variable style has no additional meaning, and in fact we implement it as a special identity function that should be called in the right places within Coq lemma statements. Rather, use of this identity function triggers the right behavior in our tactic code that reifies lemma statements. We introduce a notation where a prefixed apostrophe signals a call to the “constants only” function.

Our reification inspects the hypotheses of lemma statements, using type classes to find decidable realizations of the predicates that are used, synthesizing one Boolean expression of our deeply embedded term language, standing for a decision procedure for the hypotheses. The `Make` command fails if any such expression contains pattern variables not marked as constants. Therefore, matching of rules can safely run side conditions, knowing that Coq’s full-reduction engine can determine their truth efficiently.

5.4.4 Side Conditions Need Abstract Interpretation

With our limitation that side conditions are decided by executable Boolean procedures, we cannot yet handle directly some of the rewrites needed for realistic partial evaluation. For instance, Fiat Cryptography reduces high-level functional to low-level code that only uses integer types available on the target hardware. The starting library code works with infinite-precision integers, while the generated low-level code should be careful to avoid unintended integer overflow. As a result, the setup may be too naive for our running example rule $?n + 0 \rightarrow n$. When we get to reducing

fixed-precision-integer terms, we must be legalistic:

$$\text{add_with_carry}_{64}(\text{?}n, 0) \rightarrow (0, n) \text{ if } 0 \leq n < 2^{64}$$

We developed a design pattern to handle this kind of rule.

First, we introduce a family of functions $\text{clip}_{l,u}$, each of which forces its integer argument to respect lower bound l and upper bound u . Partial evaluation is proved with respect to unknown realizations of these functions, only requiring that $\text{clip}_{l,u}(n) = n$ when $l \leq n < u$. Now, before we begin partial evaluation, we can run a verified abstract interpreter to find conservative bounds for each program variable. When bounds l and u are found for variable x , it is sound to replace x with $\text{clip}_{l,u}(x)$. Therefore, at the end of this phase, we assume all variable occurrences have been rewritten in this manner to record their proved bounds.

Second, we proceed with our example rule refactored:

$$\begin{aligned} \text{add_with_carry}_{64}(\text{clip}_{\text{?}l, \text{?}u}(\text{?}n), 0) &\rightarrow (0, \text{clip}_{l,u}(n)) \\ &\text{if } u < 2^{64} \end{aligned}$$

If the abstract interpreter did its job, then all lower and upper bounds are constants, and we can execute side conditions straightforwardly during pattern matching.

5.5 Evaluation

Our implementation, attached to this submission as an anonymized supplement with a roadmap in Section 5.D, includes a mix of Coq code for the proved core of rewriting, tactic code for setting up proper use of that core, and OCaml plugin code for the manipulations beyond the current capabilities of the tactic language. We report here on experiments to isolate performance benefits for rewriting under binders and reducing higher-order structure.

5.5.1 Microbenchmarks

We start with microbenchmarks focusing attention on particular aspects of reduction and rewriting, with Section 5.A going into more detail.

Rewriting Under Binders

Consider

```
let v1 := v0 + v0 + 0 in
  :
let vn := vn-1 + vn-1 + 0 in
  vn + vn + 0
```

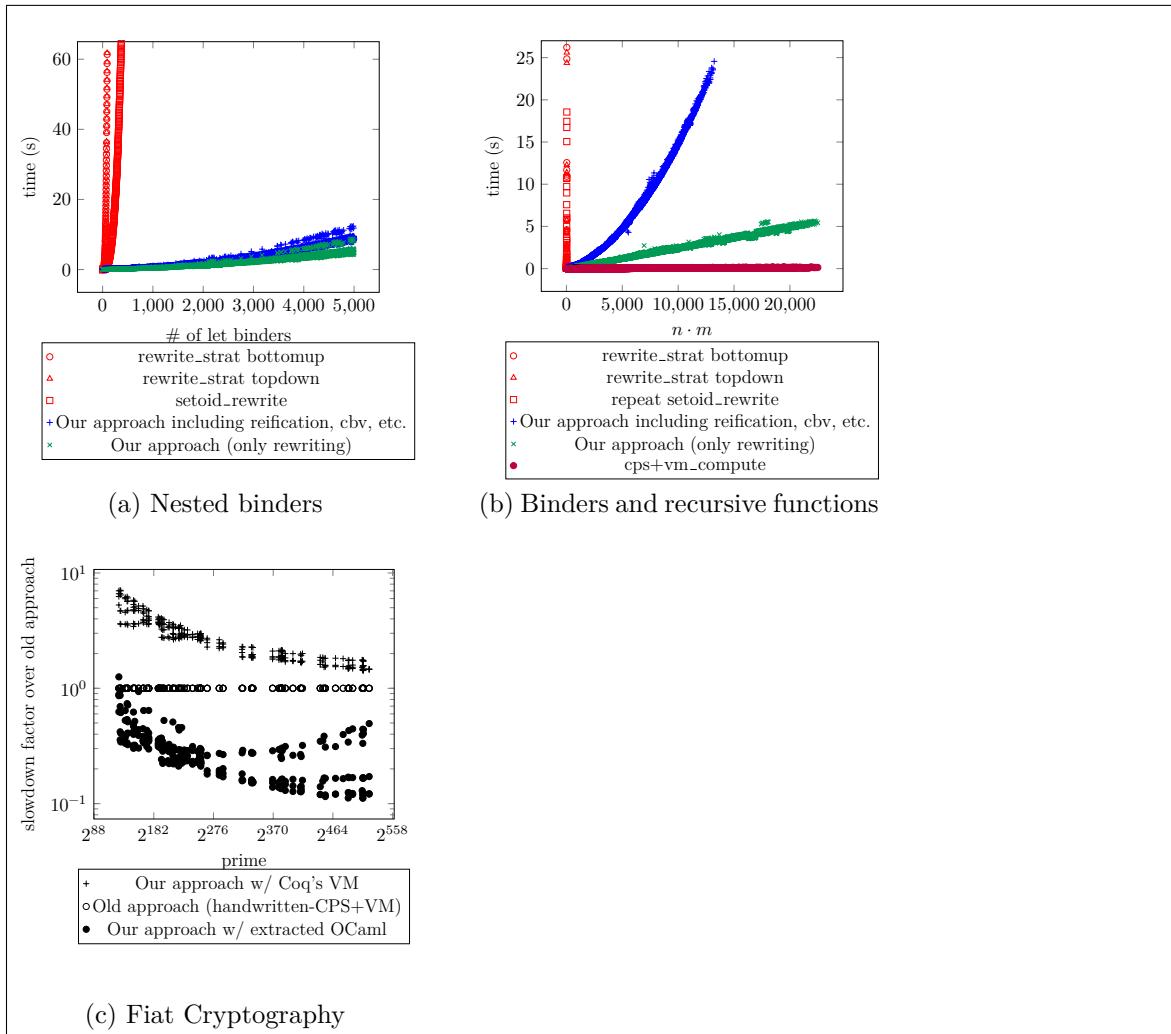


Figure 5-3: Timing of different partial-evaluation implementations

We want to remove all of the `+ 0s`. We can start from this expression directly, in which case reification alone takes as much time as `setoid_rewrite`. As the reification method was not especially optimized, and there exist fast reification methods [90], we instead start from a call to a recursive function that generates such a sequence of `let` bindings.

5-3a shows the results. The comparison points are Coq’s `setoid_rewrite` and `rewrite_strat`. The former performs one rewrite at a time, taking minimal advantage of commonalities across them and thus generating quite large, redundant proof terms. The latter makes top-down or bottom-up passes with combined generation of proof terms. For our own approach, we list both the total time and the time taken for core execution of a verified rewrite engine, without counting reification (converting goals to ASTs) or its inverse (interpreting results back to normal-looking goals).

The comparison here is very favorable for our approach. The competing tactics spike upward toward timeouts at just a few hundred generated binders, while our engine is only taking about 10 seconds for examples with 5,000 nested binders.

As detailed in Subsection 5.A.2, we ran a variant of this experiment with inlining of `lets`, forcing terms to grow quite large. Specifically, we generate n nested `lets`, each repeatedly adding a designated free variable into a sum, m times. Holding m fixed at a small value and letting n scale, we continue dominating the methods described above, though Coq’s `rewrite!` tactic (to rewrite with one lemma many times) does better for $m < 2$. Holding n fixed and letting m scale, all other approaches quickly spike upward to timeouts, while ours holds steady even for $m = 1000$.

Binders and Recursive Functions

The next experiment uses the following example.

$$\begin{aligned} \text{map-dbl}(\ell) &:= \begin{cases} [] & \text{if } \ell = [] \\ \text{let } y := h + h \text{ in } y :: \text{map-dbl}(t) & \text{if } \ell = h :: t \end{cases} \\ \text{make}(n, m, v) &:= \begin{cases} [v, \dots, v] & \text{if } m = 0 \\ \text{map-dbl}(\text{make}(n, m - 1, v)) & \text{if } m > 0 \end{cases} \\ \text{example}_{n,m} &:= \forall v, \text{make}(n, m, v) = [] \end{aligned}$$

Note that the `let ... in ...` binding blocks further reduction of `map-dbl`, which we iterate m times, and so we need to take care to preserve sharing when reducing here.

5-3b compares performance between our approach, `repeat setoid_rewrite`, and two variants of `rewrite_strat`. Additionally, we consider another option, which was adopted by Fiat Cryptography at a larger scale: rewrite our functions to improve reduction behavior. Specifically, both functions are rewritten in continuation-passing style, which makes them harder to read and reason about but allows standard VM-style reduction.

based reduction to achieve good performance. The figure shows that `rewrite_strat` variants are essentially unusable for this example, with `setoid_rewrite` performing only marginally better, while our approach applied to the original, more readable definitions loses ground steadily to VM-based reduction on CPSed code. On the largest terms ($n \cdot m > 20,000$), the gap is 6s vs. 0.1s of compilation time, which should often be acceptable in return for simplified coding and proofs, plus the ability to mix proved rewrite rules with built-in reductions. See Subsection 5.A.3 for more on this microbenchmark and Subsection 5.A.4 for an even more extreme example of full reduction with a Sieve of Eratosthenes as in the experiments of Aehlig, Haftmann, and Nipkow [5] (ours 10s, VM 0.3s).

5.5.2 Macrobenchmark: Fiat Cryptography

Finally, we consider an experiment (described in more detail in Section 5.B) replicating the generation of performance-competitive finite-field-arithmetic code for all popular elliptic curves by Erbsen et al. [70]. In all cases, we generate essentially the same code as they did, so we only measure performance of the code-generation process. We stage partial evaluation with three different reduction engines (i.e., three `Make` invocations), respectively applying 85, 56, and 44 rewrite rules (with only 2 rules shared across engines), taking total time of about 5 minutes to generate all three engines. These engines support 95 distinct function symbols.

5-3c graphs running time of three different partial-evaluation methods for Fiat Cryptography, as the prime modulus of arithmetic scales up. Times are normalized to the performance of the original method, which relied entirely on standard Coq reduction. Actually, in the course of running this experiment, we found a way to improve the old approach for a fairer comparison. It had relied on Coq’s configurable `cbv` tactic to perform reduction with selected rules of the definitional equality, which the Fiat Cryptography developers had applied to blacklist identifiers that should be left for compile-time execution. By instead hiding those identifiers behind opaque module-signature ascription, we were able to run Coq’s more-optimized virtual-machine-based reducer.

As the figure shows, our approach running partial evaluation inside Coq’s kernel begins with about a $10\times$ performance disadvantage vs. the original method. With log scale on both axes, we see that this disadvantage narrows to become nearly negligible for the largest primes, of around 500 bits. (We used the same set of prime moduli as in the experiments run by Erbsen et al. [70], which were chosen based on searching the archives of an elliptic-curves mailing list for all prime numbers.) It makes sense that execution inside Coq leaves our new approach at a disadvantage, as we are essentially running an interpreter (our normalizer) within an interpreter (Coq’s kernel), while the old approach ran just the latter directly. Also recall that the old approach required rewriting Fiat Cryptography’s library of arithmetic functions in continuation-passing style, enduring this complexity in library correctness proofs, while our new approach applies to a direct-style library. Finally, the old approach included a custom reflection-

based arithmetic simplifier for term syntax, run after traditional reduction, whereas now we are able to apply a generic engine that combines both, without requiring more than proving traditional rewrite rules.

The figure also confirms clear performance advantage of running reduction in code extracted to OCaml, which is possible because our plugin produces verified code in Coq’s functional language. By the time we reach middle-of-the-pack prime size around 300 bits, the extracted version is running about 10× as quickly as the baseline.

5.6 Related Work

We have already discussed the work of Aehlig, Haftmann, and Nipkow [5], which introduced the basic structure that our engine shares, but which required a substantially larger trusted code base, did not tackle certain challenges in scaling to large partial-evaluation problems, and did not report any performance experiments in partial evaluation.

We have also mentioned \mathcal{R}_{tac} [128], which implements an experimental reflective version of `rewrite_strat` supporting arbitrary setoid relations, unification variables, and arbitrary semi-decidable side conditions solvable by other reflective tactics, using de Bruijn indexing to manage binders. We were unfortunately unable to get the rewriter to work with Coq 8.10 and were also not able to determine from the paper how to repurpose the rewriter to handle our benchmarks.

Our implementation builds on fast full reduction in Coq’s kernel, via a virtual machine [**vmcompute**] or compilation to native code [34]. Especially the latter is similar in adopting an NbE style for full reduction, simplifying even under λ s, on top of a more traditional implementation of OCaml that never executes preemptively under λ s. Neither approach unifies support for rewriting with proved rules, and partial evaluation only applies in very limited cases, where functions that should not be evaluated at compile time must have properly opaque definitions that the evaluator will not consult. Neither implementation involved a machine-checked proof suitable to bootstrap on top of reduction support in a kernel providing simpler reduction.

A variety of forms of pragmatic partial evaluation have been demonstrated, with Lightweight Modular Staging [167] in Scala as one of the best-known current examples. A kind of type-based overloading for staging annotations is used to smooth the rough edges in writing code that manipulates syntax trees. The LMS-Verify system [15] can be used for formal verification of generated code after-the-fact. Typically LMS-Verify has been used with relatively shallow properties (though potentially applied to larger and more sophisticated code bases than we tackle), not scaling to the kinds of functional-correctness properties that concern us here, justifying investment in verified partial evaluators.

5.7 Future Work

There are a number of natural extensions to our engine. For instance, we do not yet allow pattern variables marked as “constants only” to apply to container datatypes; we limit the mixing of higher-order and polymorphic types, as well as limiting use of first-class polymorphism; we do not support proving equalities on functions; we only support decidable predicates as rule side conditions, and the predicates may only mention pattern variables restricted to matching constants; we have hardcoded support for a small set of container types and their eliminators; we support rewriting with equality and no other relations (e.g., subset inclusion); and we require decidable equality for all types mentioned in rules. It may be helpful to design an engine that lifts some or all of these limitations, building on the basic structure that we present here.

5.A Additional Information on Microbenchmarks

We performed all benchmarks on a 3.5 GHz Core i7 running Linux and Coq 8.10.0. We name the subsections here with the names that show up in the code supplement.

5.A.1 UnderLetsPlus0

We provide more detail on the “nested binders” microbenchmark of Section 5.5.1 displayed in 5-3a.

Recall that we are removing all of the + 0s from

```
let v1 := v0 + v0 + 0 in
  :
let vn := vn-1 + vn-1 + 0 in
  vn + vn + 0
```

The code used to define this microbenchmark is

```
Definition make_lets_def (n:nat) (v acc : Z) :=
@nat_rect
  (fun _ => Z * Z -> Z)
  (fun '(v, acc) => acc + acc + v)
  (fun _ rec '(v, acc) =>
    dlet acc := acc + acc + v in rec (v, acc))
n
(v, acc).
```

We note some details of the rewriting framework that were glossed over in the main body of the paper, which are useful for using the code: Although the rewriting framework does not support dependently typed constants, we can automatically preprocess uses of eliminators like `nat_rect` and `list_rect` into non-dependent versions. The tactic that does this preprocessing is extensible via \mathcal{L}_{tac} ’s reassignment feature. Since pattern-matching compilation mixed with NbE requires knowing how many arguments a constant can be applied to, we must internally use a version of the recursion principle whose type arguments do not contain arrows; current preprocessing can handle recursion principles with either no arrows or one arrow in the motive. Even though we will eventually plug in 0 for v , we jump through some extra hoops to ensure that our rewriter cannot cheat by rewriting away the $+ 0$ before reducing the recursion on n .

We can reduce this expression in three ways.

Our Rewriter

One lemma is required for rewriting with our rewriter:

```
Lemma Z.add_0_r : forall z, z + 0 = z.
```

Creating the rewriter takes about 12 seconds on the machine we used for running the performance experiments:

```
Make myrew := Rewriter For
  (Z.add_0_r, eval_rect nat, eval_rect prod).
```

Recall from Subsection 5.1.1 that `eval_rect` is a definition provided by our framework for eagerly evaluating recursion associated with certain types. It functions by triggering typeclass resolution for the lemmas reducing the recursion principle associated to the given type. We provide instances for `nat`, `prod`, `list`, `option`, and `bool`. Users may add more instances if they desire.

setoid_rewrite and rewrite_strat

To give as many advantages as we can to the preexisting work on rewriting, we pre-reduce the recursion on `nats` using `cbv` before performing `setoid_rewrite`. (Note that `setoid_rewrite` cannot itself perform reduction without generating large proof terms, and `rewrite_strat` is not currently capable of sequencing reduction with rewriting internally due to bugs such as #10923.) Rewriting itself is easy; we may use any of `repeat setoid_rewrite Z.add_0_r`, `rewrite_strat topdown Z.add_0_r`, or `rewrite_strat bottomup Z.add_0_r`.

5.A.2 Plus0Tree

This is a version of Subsection 5.A.1 without any let binders, discussed in Section 5.5.1 but not displayed in Figure 5-3.

We use two definitions for this microbenchmark:

```
Definition iter (m : nat) (acc v : Z) :=
  @nat_rect
  (fun _ => Z -> Z)
  (fun acc => acc)
  (fun _ rec acc => rec (acc + v))
  m
  acc.
```

```

Definition make_tree (n m : nat) (v acc : Z) :=
Eval cbv [iter] in
@nat_rect
  (fun _ => Z * Z -> Z)
  (fun '(v, acc) => iter m (acc + acc) v)
  (fun _ rec '(v, acc) =>
    iter m (rec (v, acc) + rec (v, acc)) v)
n
(v, acc).

```

We can see from the graphs in Figure 5-4 and Figure 5-5 that (a) we incur constant overhead over most of the other methods which dominates on small examples; (b) when the term is quite large and there are few opportunities for rewriting relative to the term-size (i.e., $m \leq 2$), we are worse than **rewrite !Z.add_0_r**, but still better than the other methods; and (c) when there are many opportunities for rewriting relative to the term-size ($m > 2$), we thoroughly dominate the other methods.

5.A.3 LiftLetsMap

We now discuss in more detail the “binders and recursive functions” example from Section 5.5.1.

The expression we want to get out at the end looks like:

```

let v1,1 := v + v in
:
let v1,n := v + v in
let v2,1 := v1,1 + v1,1 in
:
let v2,n := v1,n + v1,n in
:
[vm,1, ..., vm,n]

```

Recall that we make this example with the code

```

Definition map_double (ls : list Z) :=
list_rect
  []
  (λ x xs rec, let y := x + x in y :: rec)
ls.

```

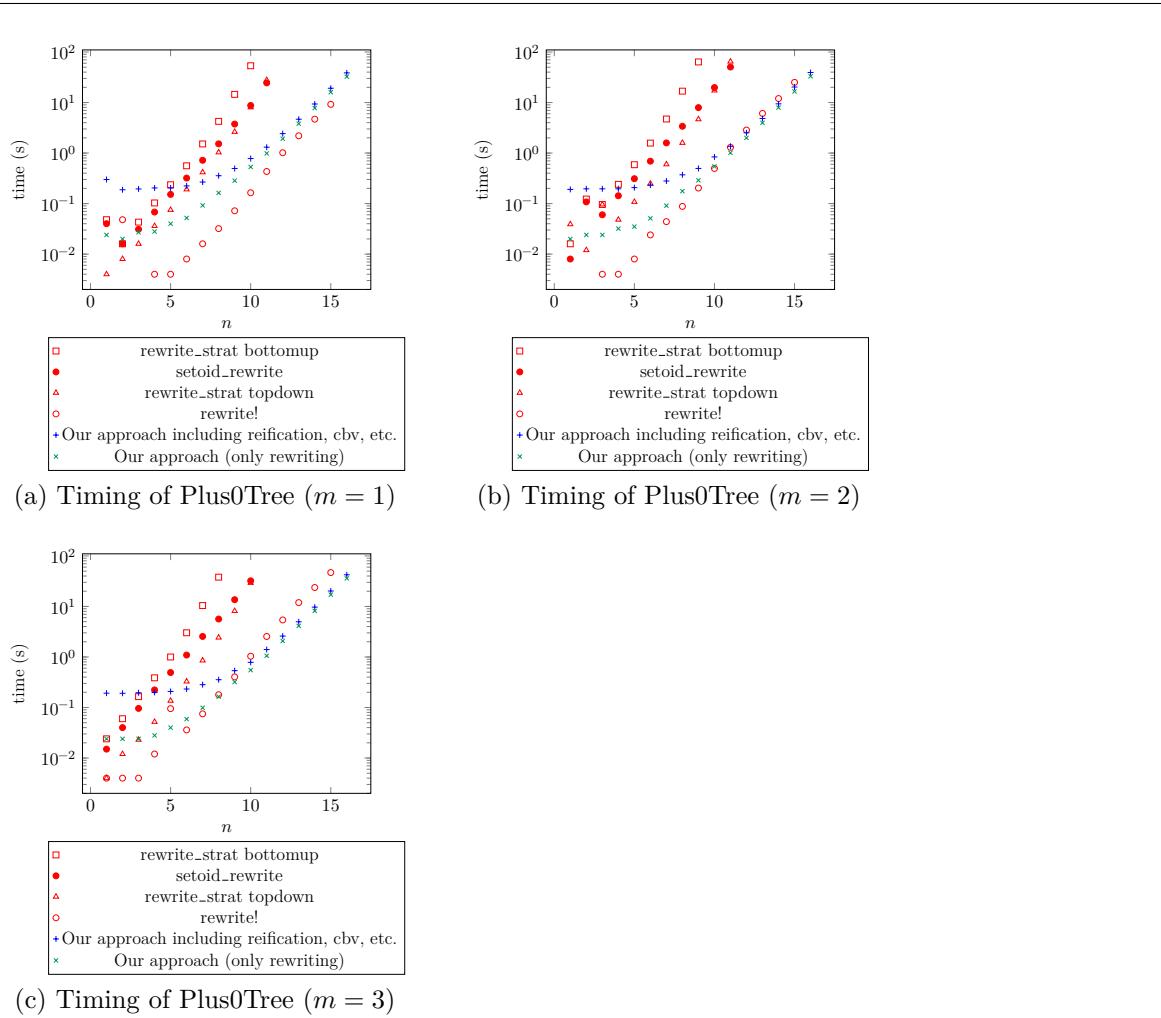


Figure 5-4: Timing of different partial-evaluation implementations for Plus0Tree for fixed m . Note that we have a logarithmic time scale, because term size is proportional to 2^n .

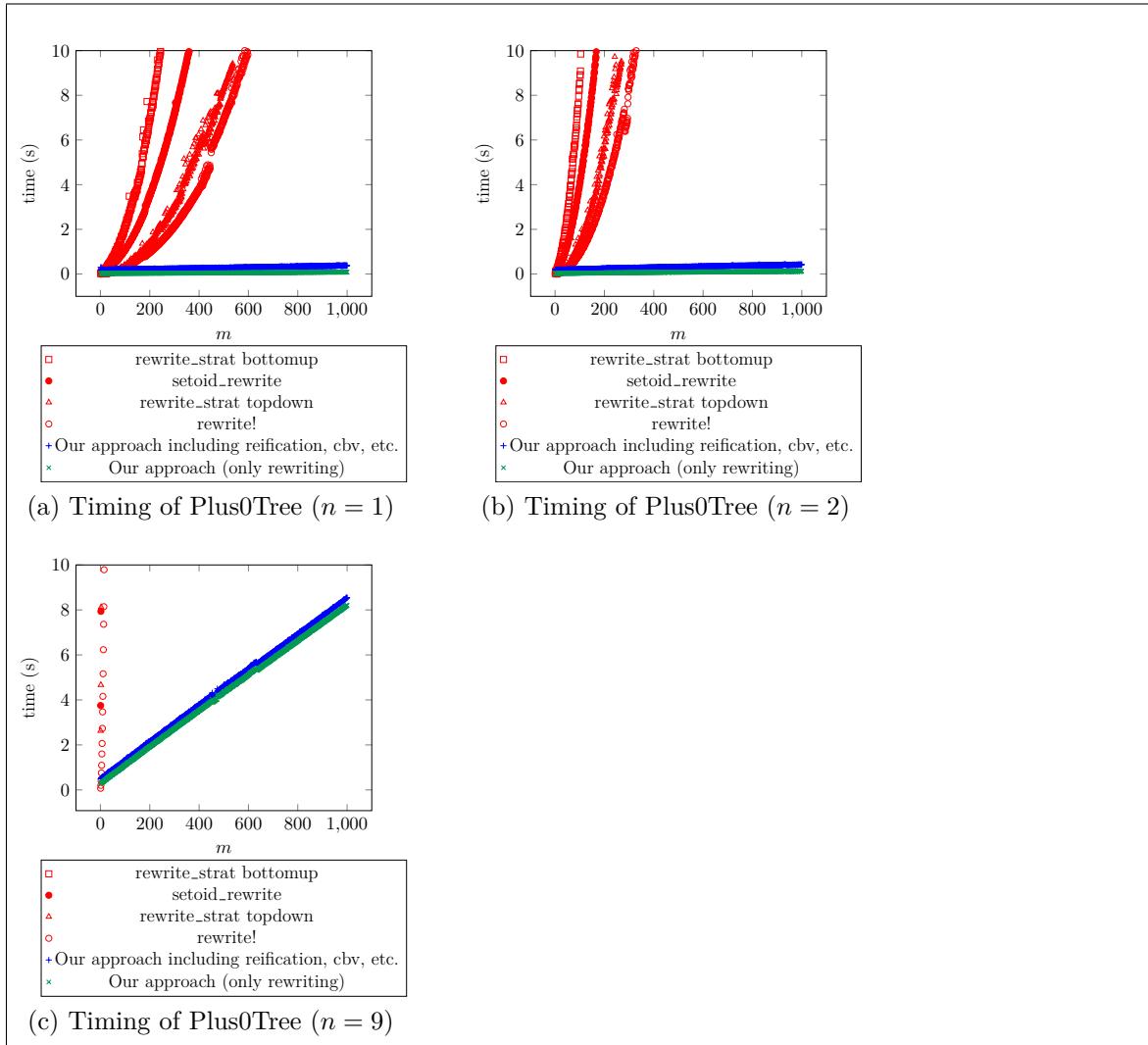


Figure 5-5: Timing of different partial-evaluation implementations for Plus0Tree for fixed n (1, 2, and then we jump to 9)

```

Definition make (n : nat) (m : nat) (v : Z) :=
  nat_rect
  _ (List.repeat v n)
  (λ _ rec, map_double rec)
  m.

```

We can perform this rewriting in four ways; see 5-3b.

Note that `rewrite_strat` grows quite quickly, hitting a minute when the total number of rewrites ($n \cdot m$) is in the mid-40s. Our method performs much better, but the fact that we have to perform `cbv` at the end costs us; about 99% of the difference between the full time of our method and just the rewriting is spent in the final `cbv` at the end. This is due to the unfortunate fact that reduction in Coq is quadratic in the number of nested binders present; see Coq bug #11151. Finally, and unsurprisingly, `vm_compute` outperforms us.

Our Rewriter

One lemma is required for rewriting with our rewriter:

```

Lemma eval_repeat A x n :
  @List.repeat A x ('n)
  = ident.eagerly nat_rect _
    []
    (λ k repeat_k, x :: repeat_k)
  ('n).

```

Recall that the apostrophe marker ('') is explained in Subsection 5.1.1. Recall again from Subsection 5.1.1 that we use `ident.eagerly` to ask the reducer to simplify a case of primitive recursion by complete traversal of the designated argument's constructor tree. Our current version only allows a limited, hard-coded set of eliminators with `ident.eagerly` (`nat_rect` on return types with either zero or one arrows, `list_rect` on return types with either zero or one arrows, and `List.nth_default`), but nothing in principle prevents automatic generation of the necessary code.

We construct our rewriter with

```

Make myrew := Rewriter For
  (eval_repeat, eval_rect list, eval_rect nat)
  (with extra idents (Z.add)).

```

On the machine we used for running all our performance experiments, this command takes about 13 seconds to run. Note that all identifiers which appear in any goal to

be rewritten must either appear in the type of one of the rewrite rules or in the tuple passed to `with extra idents`.

Rewriting is relatively simple, now. Simply invoke the tactic `Rewrite_for myrew`. We support rewriting on only the left-hand-side and on only the right-hand-side using either the tactic `Rewrite_lhs_for myrew` or else the tactic `Rewrite_rhs_for myrew`, respectively.

`rewrite_strat`

To reduce adequately using `rewrite_strat`, we need the following two lemmas:

```
Lemma lift_let_list_rect T A P N C (v : A) fls
: @list_rect T P N C (Let_In v fls)
= Let_In v (fun v => @list_rect T P N C (fls v)).
Lemma lift_let_cons T A x (v : A) f
: @cons T x (Let_In v f)
= Let_In v (fun v => @cons T x (f v)).
```

Note that `Let_In` is the constant we use for writing `let ... in ...` expressions that do not reduce under ζ . Throughout most of this paper, anywhere that `let ... in ...` appears, we have actually used `Let_In` in the code. It would alternatively be possible to extend the reification preprocessor to automatically convert `let ... in ...` to `Let_In`, but this may cause problems when converting the interpretation of the reified term with the pre-reified term, as Coq's conversion does not allow fine-tuning of when to inline or unfold `lets`.

To rewrite, we start with `cbv [example make map dbl]` to expose the underlying term to rewriting. One would hope that one could just add these two hints to a database `db` and then write `rewrite_strat (repeat (eval cbn [list_rect]; try bottomup hints db))`, but unfortunately this does not work due to a number of bugs in Coq: #10934, #10923, #4175, #10955, and the potential to hit #10972. Instead, we must put the two lemmas in separate databases, and then write `repeat (cbn [list_rect]; (rewrite_strat (try repeat bottomup hints db1)); (rewrite_strat (try repeat bottomup hints db2)))`. Note that the rewriting with `lift_let_cons` can be done either top-down or bottom-up, but `rewrite_strat` breaks if the rewriting with `lift_let_list_rect` is done top-down.

CPS and the VM

If we want to use Coq's built-in VM reduction without our rewriter, to achieve the prior state-of-the-art performance, we can do so on this example, because it only involves partial reduction and not equational rewriting. However, we must (a) module-opacify the constants which are not to be unfolded, and (b) rewrite all of our code in CPS.

Then we are looking at

$$\text{map_dbl_cps}(\ell, k) := \begin{cases} k([]) & \text{if } \ell = [] \\ \text{let } y := h +_{\text{ax}} h \text{ in } & \text{if } \ell = h :: t \\ \text{map_dbl_cps}(t, & \\ & (\lambda ys, k(y :: ys))) \end{cases}$$

$$\text{make_cps}(n, m, v, k) := \begin{cases} k([\underbrace{v, \dots, v}_n]) & \text{if } m = 0 \\ \text{make_cps}(n, m - 1, v, & \text{if } m > 0 \\ & (\lambda \ell, \text{map_dbl_cps}(\ell, k))) \end{cases}$$

$$\text{example_cps}_{n,m} := \forall v, \text{ make_cps}(n, m, v, \lambda x. x) = []$$

Then we can just run `vm_compute`. Note that this strategy, while quite fast, results in a stack overflow when $n \cdot m$ is larger than approximately $2.5 \cdot 10^4$. This is unsurprising, as we are generating quite large terms. Our framework can handle terms of this size but stack-overflows on only slightly larger terms.

Takeaway

From this example, we conclude that `rewrite_strat` is unsuitable for computations involving large terms with many binders, especially in cases where reduction and rewriting need to be interwoven, and that the many bugs in `rewrite_strat` result in confusing gymnastics required for success. The prior state of the art—writing code in CPS—suitably tweaked by using module pacity to allow `vm_compute`, remains the best performer here, though the cost of rewriting everything in CPS may be prohibitive. Our method soundly beats `rewrite_strat`. We are additionally bottlenecked on `cbv`, which is used to unfold the goal post-rewriting and costs about a minute on the largest of terms; see Coq bug #11151 for a discussion on what is wrong with Coq’s reduction here.

5.A.4 SieveOfEratosthenes

To benchmark how much overhead we add when we are reducing fully, we compute the Sieve of Eratosthenes, taking inspiration on benchmark choice from Aehlig, Haftmann, and Nipkow [5]. We find in Figure 5-6 that we are slower than `vm_compute`, `native_compute`, and `cbv`, but faster than `lazy`, and of course much faster than `simpl` and `cbn`, which are quite slow.

We define the sieve using `PositiveMap.t` and `list Z`:

```
Definition sieve' (fuel : nat) (max : Z) :=
List.rev
```

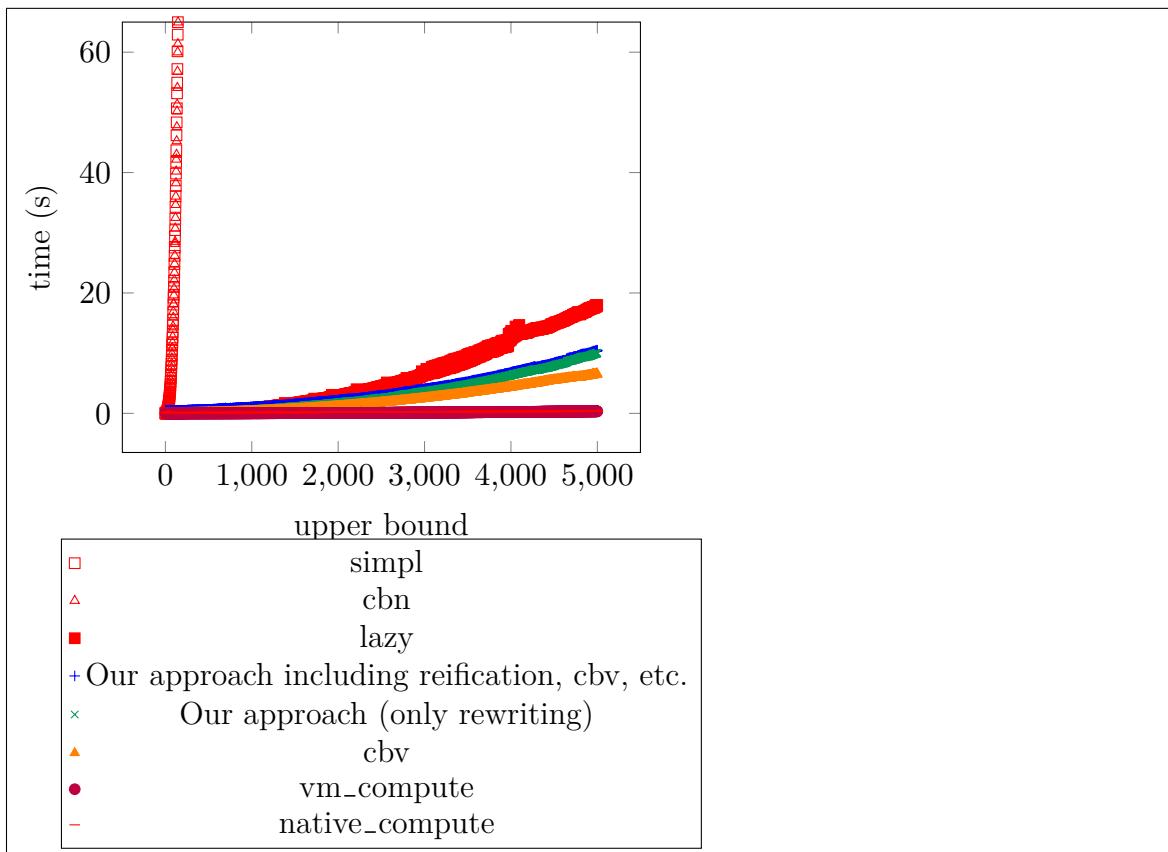


Figure 5-6: Timing of different full-evaluation implementations for SieveOfEratosthenes

```

(fst
  (@nat_rect
    (λ _, list Z (* primes *) *
      PositiveSet.t (* composites *) *
      positive (* np (next_prime) *) ->
      list Z (* primes *) *
      PositiveSet.t (* composites *))
    (λ '(primes, composites, next_prime),
      (primes, composites))
    (λ _ rec '(primes, composites, np),
      rec
        (if (PositiveSet.mem np composites ||
            (Z.pos np >? max))%bool%Z
          then
            (primes, composites, Pos.succ np)
          else
            (Z.pos np :: primes,
              List.fold_right
                PositiveSet.add
                composites
                (List.map
                  (λ n, Pos.mul (Pos.of_nat (S n)) np)
                  (List.seq 0 (Z.to_nat(max/Z.pos np)))), 
                  Pos.succ np)))
        fuel
        (nil, PositiveSet.empty, 2%positive))).

```

```

Definition sieve (n : Z)
:= Eval cbv [sieve'] in sieve' (Z.to_nat n) n.

```

We need four lemmas and an additional instance to create the rewriter:

```

Lemma eval_fold_right A B f x ls :
@List.fold_right A B f x ls
= ident.eagerly list_rect _ _
  x
  (λ l ls fold_right_ls, f l fold_right_ls)
  ls.

Lemma eval_app A xs ys :
xs ++ ys
= ident.eagerly list_rect A _
  ys
  (λ x xs app_xs_ys, x :: app_xs_ys)
  xs.

```

```

Lemma eval_map A B f ls :
@List.map A B f ls
= ident.eagerly list_rect _ _
[]
(λ l ls map_ls, f l :: map_ls)
ls.

Lemma eval_rev A xs :
@List.rev A xs
= (@list_rect _ (fun _ => _))
[]
(λ x xs rev_xs, rev_xs ++ [x])%list
xs.

Scheme Equality for PositiveSet.tree.

Definition PositiveSet_t_beq
: PositiveSet.t -> PositiveSet.t -> bool
:= tree_beq.

Global Instance PositiveSet_reflect_eqb
: reflect_rel (@eq PositiveSet.t) PositiveSet_t_beq
:= reflect_of_brel
internal_tree_dec_bl internal_tree_dec_lb.

```

We then create the rewriter with

```

Make myrew := Rewriter For
  (eval_rect nat, eval_rect prod, eval_fold_right,
  eval_map, do_again eval_rev, eval_rect bool,
  @fst_pair, eval_rect list, eval_app)
  (with extra idents (Z.eqb, orb, Z.gtb,
  PositiveSet.elements, @fst, @snd,
  PositiveSet.mem, Pos.succ, PositiveSet.add,
  List.fold_right, List.map, List.seq, Pos.mul,
  S, Pos.of_nat, Z.to_nat, Z.div, Z.pos, 0,
  PositiveSet.empty))
  (with delta).

```

To get `cbn` and `simpl` to unfold our term fully, we emit

```
Global Arguments Pos.to_nat !_ / .
```

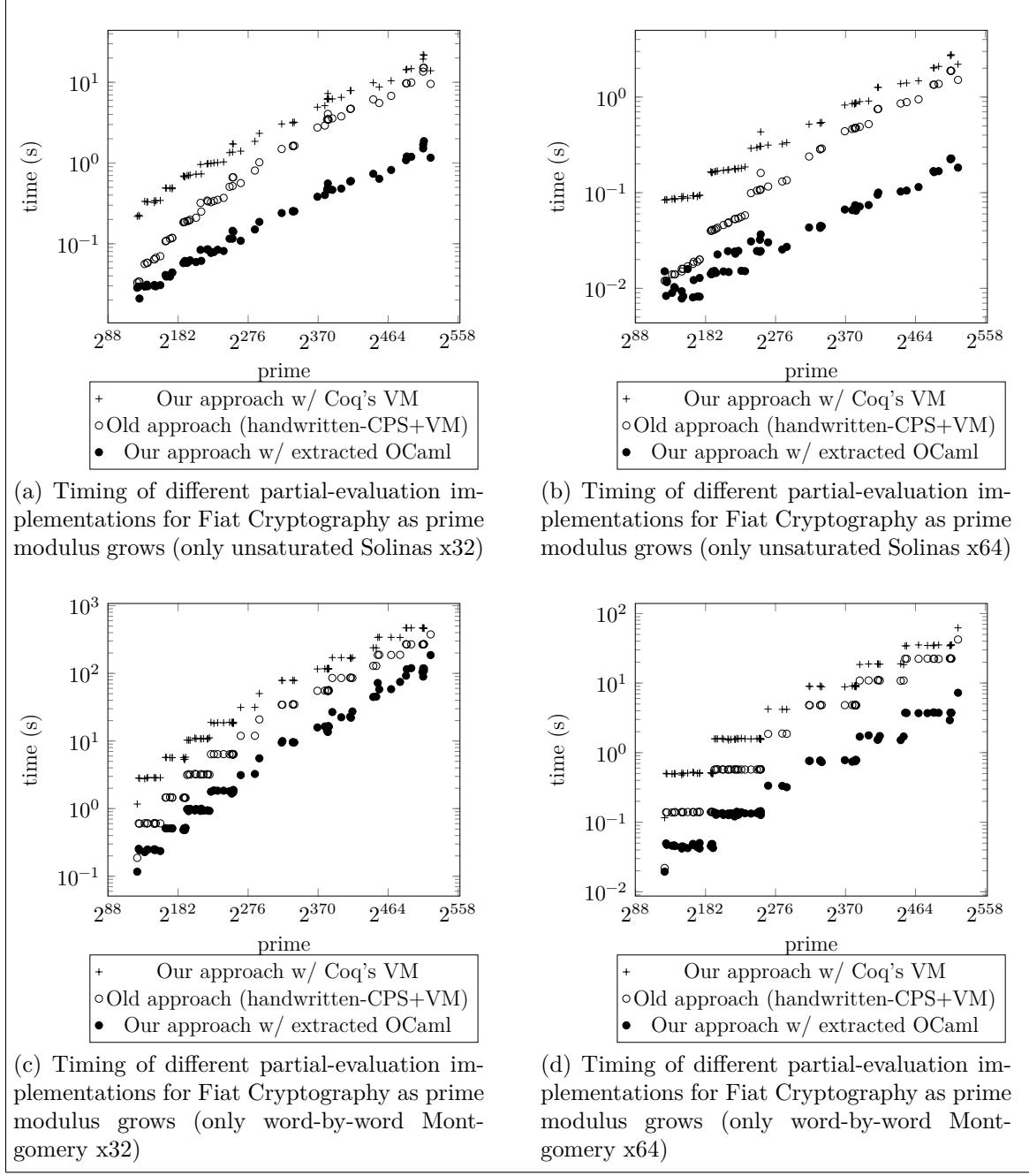


Figure 5-7: Timing of different partial-evaluation implementations for Fiat Cryptography as prime modulus grows

5.B Additional Information on Fiat Cryptography Benchmarks

It may also be useful to see performance results with absolute times, rather than normalized execution ratios vs. the original Fiat Cryptography implementation. Furthermore, the benchmarks fit into four quite different groupings: elements of the cross product of two algorithms (unsaturated Solinas and word-by-word Montgomery) and bitwidths of target architectures (32-bit or 64-bit). Here we provide absolute-time graphs by grouping in Figure 5-7.

5.C Experience vs. Lean and `setoid_rewrite`

Although all of our toy examples work with `setoid_rewrite` or `rewrite_strat` (until the terms get too big), even the smallest of examples in Fiat Cryptography fell over using these tactics. When attempting to use `rewrite_strat` for partial evaluation and rewriting on unsaturated Solinas with 1 limb on small primes (such as 29), we were able to get `rewrite_strat` to finish after about 90 seconds. The bugs in `rewrite_strat` made finding the right magic invocation quite painful, nonetheless; the invocation we settled on involved *sixteen* consecutive calls to `rewrite_strat` with varying arguments and strategies. Trying to synthesize code for two limbs on slightly larger primes (such as 113, which needs two limbs on a 64-bit machine) took about three hours. The widely used primes tend to have around five to ten limbs; we leave extrapolating this slowdown to the reader.

We have attached this experiment using `rewrite_strat` as `fiat_crypto_via_rewrite_strat.v`, which is meant to be run in emacs/PG from inside the `fiat-crypto` directory, or in `coqc` by setting `COQPATH` to the value emitted by `make printenv` in `fiat-crypto` and then invoking the command `coqc -q -R /path/to/fiat-crypto/src Crypto /path/to/fiat_crypto_via_rewrite_strat.v`. To test with the two-limb prime 113, change `of_string "2^5-3"` 8 in the definition of `p` to `of_string "2^7-15"` 64.

We also tried Lean, in the hopes that rewriting in Lean, specifically optimized for performance, would be up to the challenge. Although Lean performed about 30% better than Coq on the 1-limb example, taking a bit under a minute, it did not complete on the two-limb example even after four hours (after which we stopped trying), and a five-limb example was still going after 40 hours.

We have attached our experiments with running `rewrite` in Lean on the Fiat Cryptography code as a supplement as well. We used Lean version 3.4.2, commit cbd2b6686ddb, Release. Run `make` in `fiat-crypto-lean` to run the one-limb example; change `open ex` to `open ex2` to try the two-limb example, or to `open ex5` to try the five-limb example.

5.D Reading the Code Supplement

We have attached both the code for implementing the rewriter, as well as a copy of Fiat Cryptography adapted to use the rewriting framework. Both code supplements build with Coq 8.9 and Coq 8.10, and they require that whichever OCaml was used to build Coq be installed on the system to permit building plugins. (If Coq was installed via opam, then the correct version of OCaml will automatically be available.) Both code bases can be built by running `make` in the top-level directory.

The performance data for both repositories are included at the top level as `.txt` and `.csv` files.

The performance data for the microbenchmarks can be rebuilt using `make perf-SuperFast` `perf-Fast` `perf-Medium` followed by `make perf-csv` to get the `.txt` and `.csv` files. The microbenchmarks should run in about 24 hours when run with `-j5` on a 3.5 GHz machine. There also exist targets `perf-Slow` and `perf-VerySlow`, but these take significantly longer.

The performance data for the macrobenchmark can be rebuilt from the Fiat Cryptography copy included by running `make perf -k`. We ran this with `PERF_MAX_TIME=3600` to allow each benchmark to run for up to an hour; the default is 10 minutes per benchmark. Expect the benchmarks to take over a week of time with an hour timeout and five cores. Some tests are expected to fail, making `-k` a necessary flag. Again, the `perf-csv` target will aggregate the logs and turn them into `.txt` and `.csv` files.

The entry point for the rewriter is the Coq source file `rewriter/src/Rewriter/Util/plugins/RewriterBuild.v`.

The rewrite rules used in Fiat Cryptography are defined in `fiat-crypto/src/Rewriter/Rules.v` and proven in `fiat-crypto/src/Rewriter/RulesProofs.v`. Note that the Fiat Cryptography copy uses `COQPATH` for dependency management, and `.dir-locals.el` to set `COQPATH` in emacs/PG; you must accept the setting when opening a file in the directory for interactive compilation to work. Thus interactive editing either requires ProofGeneral or manual setting of `COQPATH`. The correct value of `COQPATH` can be found by running `make printenv`.

We will now go through this paper and describe where to find each reference in the code base.

5.D.1 Code from Section 5.1, Introduction Code from Subsection 5.1.1, A Motivating Example

The `prefixSums` example appears in the Coq source file `rewriter/src/Rewriter/Rewriter/Examples/PrefixSums.v`. Note that we use `dlet` rather than `let` in binding `acc'` so that we can preserve the `let` binder even under ι reduction, which much of

Coq’s infrastructure performs eagerly. Because we attempt to isolate the dependency on the axiom of functional extensionality as much as possible, we also in practice require `Proper` instances for each higher-order identifier saying that each constant respects function extensionality. We hope to remove the dependency on function extensionality altogether in the future. Although we glossed over this detail in the body of this paper, we also prove

```
Global Instance: forall A B,
  Proper ((eq ==> eq ==> eq) ==> eq ==> eq ==> eq)
    (@fold_left A B).
```

The `Make` command is exposed in the file `rewriter/src/Rewriter/Util/plugins/RewriterBuild.v` and defined in the OCaml file `rewriter/src/Rewriter/Util/plugins/rewriter_build_plugin.mlg`. Note that one must run `make` to create this latter file; it is copied over from a version-specific file at the beginning of the build.

The `do_again`, `eval_rect`, and `ident.eagerly` constants are defined at the bottom of module `RewriteRuleNotations` in `rewriter/src/Rewriter/Language/Pre.v`.

Code from Subsection 5.1.2, Concerns of Trusted-Code-Base Size

There is no code mentioned in this section.

Code from Subsection 5.1.3, Our Solution

We claimed that our solution meets five criteria. We briefly justify each criterion with a sentence or a pointer to code:

- We claimed that we **did not grow the trusted base** (excepting the axiom of functional extensionality). In any example file (of which a couple can be found in `rewriter/src/Rewriter/Rewriter/Examples/`), the `Make` command creates a rewriter package. Running `Print Assumptions` on this new constant (often named `rewriter` or `myrew`) should demonstrate a lack of axioms other than functional extensionality. `Print Assumptions` may also be run on the proof that results from using the rewriter.
- We claimed **fast** partial evaluation with reasonable memory use; we assume that the performance graphs stand on their own to support this claim. Note that memory usage can be observed by making the benchmarks while passing `TIMED=1` to `make`.
- We claimed to allow reduction that **mixes rules of the definitional equality with equalities proven explicitly as theorems**; the “rules of the definitional equality” are, for example, β reduction, and we assert that it should be self-evident that our rewriter supports this.

- We claimed common-subterm **sharing preservation**. This is implemented by supporting the use of the `dlet` notation which is defined in `rewriter/src/Rewriter/Util/LetIn.v` via the `Let_In` constant. We will come back to the infrastructure that supports this.
- We claimed **extraction of standalone partial evaluators**. The extraction is performed in the Coq source file `perf_unsaturated_solinis.v`, in the source file `perf_word_by_word_montgomery.v`, and in the source files `saturated_solinis.v`, `unsaturated_solinis.v`, and `word_by_word_montgomery.v`, all in the directory `fiat-crypto/src/ExtractionOCaml/`. The OCaml code can be extracted and built using the target `make standalone-ocaml` (or `make perf-standalone` for the `perf_` binaries). There may be some issues with building these binaries on Windows as some versions of `ocamlopt` on Windows seem not to support outputting binaries without the `.exe` extension.

The P-384 curve is mentioned. This is the curve with prime modulus $2^{384} - 2^{128} - 2^{96} + 2^{32} - 1$, and the benchmarks for this curve can be found in the files matching the glob `fiat-crypto/src/Rewriter/PerfTesting/Specific/generated/p2384m2128m296p232m1_*_word`. While the `.log` files are included in the tarball, the `.v` and `.sh` files are automatically generated in the course of running `make perf -k`.

We mention integration with abstract interpretation; the abstract-interpretation pass is implemented in `fiat-crypto/src/AbstractInterpretation/`.

5.D.2 Code from Section 5.2, Trust, Reduction, and Rewriting

The individual rewritings mentioned are implemented via the `Rewrite_*` tactics exported at the top of `rewriter/src/Rewriter/Util/plugins/RewriterBuild.v`. These tactics bottom out in tactics defined at the bottom of `rewriter/src/Rewriter/Rewriter/AllTactics.v`.

Code from Subsection 5.2.1, Our Approach in Nine Steps

We match the nine steps with functions from the source code:

1. The given lemma statements are scraped for which named functions and types the rewriter package will support. This is performed by `rewriter_scrape_data` in the file `rewriter/src/Rewriter/Util/plugins/rewriter_build.ml` which invokes the tactic named `make_scrape_data` in a submodule in `rewriter/src/Rewriter/Language/IdentifiersBasicGenerate.v` on a goal headed by the constant we provide under the name `Pre.ScrapedData.t_with_args` in `rewriter/src/Rewriter/Language/PreCommon.v`.

2. Inductive types enumerating all available primitive types and functions are emitted. This step is performed by `rewriter_emit_inductives` in file `rewriter/src/Rewriter/Util/plugins/rewriter_build.ml` invoking tactics, like `make_base_elim` in `rewriter/src/Rewriter/Language/IdentifiersBasicGenerate.v`, on goals headed by constants from `rewriter/src/Rewriter/Language/IdentifiersBasicLibrary.v`, including `base_elim_with_args` for example, to turn scraped data into eliminators for the inductives. The actual emitting of inductives is performed by code in the file `rewriter/src/Rewriter/Util/plugins/inductive_from_elim.ml`.
3. Tactics generate all of the necessary definitions and prove all of the necessary lemmas for dealing with this particular set of inductive codes. This step is performed by `make_rewriter_of_scraped_and_ind` in the source file `rewriter/src/Rewriter/Util/plugins/rewriter_build.ml` which invokes `make_rewriter_all` defined in the file `rewriter/src/Rewriter/Rewriter/AllTactics.v` on a goal headed by the provided constant `VerifiedRewriter_with_ind_args` defined in `rewriter/src/Rewriter/Rewriter/ProofsCommon.v`. The definitions emitted can be found by looking at the tactic `Build_Rewriter` in `rewriter/src/Rewriter/Rewriter/AllTactics.v`, the tactics `build_package` in the source file `rewriter/src/Rewriter/Language/IdentifiersBasicGenerate.v` and also in the Coq source file found in `rewriter/src/Rewriter/Language/IdentifiersGenerate.v` (there is a different tactic named `build_package` in each of these files), and the tactic `prove_package_proofs_via` which can be found in the Coq source file `rewriter/src/Rewriter/Language/IdentifiersGenerateProofs.v`.
4. The statements of rewrite rules are reified, and we prove soundness and syntactic-well-formedness lemmas about each of them. This step is performed as part of the previous step, when the tactic `make_rewriter_all` transitively calls `Build_Rewriter` from `rewriter/src/Rewriter/Rewriter/AllTactics.v`. Reification is handled by the tactic `Build_RewriterT` in `rewriter/src/Rewriter/Rewriter/Reify.v`, while soundness and syntactic-well-formedness are handled by the tactics `prove_interp_good` and `prove_good` respectively, both in the source file `rewriter/src/Rewriter/Rewriter/ProofsCommonTactics.v`.
5. The definitions needed to perform reification and rewriting and the lemmas needed to prove correctness are assembled into a single package that can be passed by name to the rewriting tactic. This step is also performed by `make_rewriter_of_scraped` in the source file `rewriter/src/Rewriter/Util/plugins/rewriter_build.ml`.

When we want to rewrite with a rewriter package in a goal, the following steps are performed, with code in the following places:

1. We rearrange the goal into a single logical formula: all free-variable quantification in the proof context is replaced by changing the equality goal into an equality between two functions (taking the free variables as inputs). Note that

it is not actually an equality between two functions but rather an `equiv` between two functions, where `equiv` is a custom relation we define indexed over type codes that is equality up to function extensionality. This step is performed by the tactic `generalize_hyps_for_rewriting` in `rewriter/src/Rewriter/Rewriter/AllTactics.v`.

2. We reify the side of the goal we want to simplify, using the inductive codes in the specified package. That side of the goal is then replaced with a call to a denotation function on the reified version. This step is performed by the tactic `do_reify_rhs_with` in `rewriter/src/Rewriter/Rewriter/AllTactics.v`.
3. We use a theorem stating that rewriting preserves denotations of well-formed terms to replace the denotation subterm with the denotation of the rewriter applied to the same reified term. We use Coq's built-in full reduction (`vm_compute`) to reduce the application of the rewriter to the reified term. This step is performed by the tactic `do_rewrite_with` in `rewriter/src/Rewriter/Rewriter/AllTactics.v`.
4. Finally, we run `cbv` (a standard call-by-value reducer) to simplify away the invocation of the denotation function on the concrete syntax tree from rewriting. This step is performed by the tactic `do_final_cbv` in `rewriter/src/Rewriter/Rewriter/AllTactics.v`.

These steps are put together in the tactic `Rewrite_for_gen` in `rewriter/src/Rewriter/Rewriter/AllTactics.v`.

Our Approach in More Than Nine Steps

As the nine steps of Subsection 5.2.1 do not exactly match the code, we describe here a more accurate version of what is going on. For ease of readability, we do not clutter this description with references to the code supplement, instead allowing the reader to match up the steps here with the more coarse-grained ones in Subsection 5.2.1 or Section 5.D.2.

In order to allow easy invocation of our rewriter, a great deal of code (about 6500 lines) needed to be written. Some of this code is about reifying rewrite rules into a form that the rewriter can deal with them in. Other code is about proving that the reified rewrite rules preserve interpretation and are well-formed. We wrote some plugin code to automatically generate the inductive type of base-type codes and identifier codes, as well as the two variants of the identifier-code inductive used internally in the rewriter. One interesting bit of code that resulted was a plugin that can emit an inductive declaration given the Church encoding (or eliminator) of the inductive type to be defined. We wrote a great deal of tactic code to prove basic properties about these inductive types, from the fact that one can unify two identifier codes and extract constraints on their type variables from this unification, to the fact that type codes

have decidable equality. Additional plugin code was written to invoke the tactics that construct these definitions and prove these properties, so that we could generate an entire rewriter from a single command, rather than having the user separately invoke multiple commands in sequence.

In order to build the precomputed rewriter, the following actions are performed:

1. The terms and types to be supported by the rewriter package are scraped from the given lemmas.
2. An inductive type of codes for the types is emitted, and then three different versions of inductive codes for the identifiers are emitted (one with type arguments, one with type arguments supporting pattern type variables, and one without any type arguments, to be used internally in pattern-matching compilation).
3. Tactics generate all of the necessary definitions and prove all of the necessary lemmas for dealing with this particular set of inductive codes. Definitions cover categories like “Boolean equality on type codes” and “how to extract the pattern type variables from a given identifier code,” and lemma categories include “type codes have decidable equality” and “the types being coded for have decidable equality” and “the identifiers all respect function extensionality.”
4. The rewrite rules are reified, and we prove interpretation-correctness and well-formedness lemmas about each of them.
5. The definitions needed to perform reification and rewriting and the lemmas needed to prove correctness are assembled into a single package that can be passed by name to the rewriting tactic.
6. The denotation functions for type and identifier codes are marked for early expansion in the kernel via the `Strategy` command; this is necessary for conversion at `Qed`-time to perform reasonably on enormous goals.

When we want to rewrite with a rewriter package in a goal, the following steps are performed:

1. We use `etransitivity` to allow rewriting separately on the left- and right-hand-sides of an equality. Note that we do not currently support rewriting in non-equality goals, but this is easily worked around using `let v := open_constr:(_) in replace <some term> with v` and then rewriting in the second goal.
2. We revert all hypotheses mentioned in the goal, and change the form of the goal from a universally quantified statement about equality into a statement that two functions are extensionally equal. Note that this step will fail if any hypotheses are functions not known to respect function extensionality via typeclass search.

3. We reify the side of the goal that is not an existential variable using the inductive codes in the specified package; the resulting goal equates the denotation of the newly reified term with the original evar.
4. We use a lemma stating that rewriting preserves denotations of well-formed terms to replace the goal with the rewriter applied to our reified term. We use `vm_compute` to prove the well-formedness side condition reflectively. We use `vm_compute` again to reduce the application of the rewriter to the reified term.
5. Finally, we run `cbv` to unfold the denotation function, and we instantiate the evar with the resulting rewritten term.

There are a couple of steps that contribute to the trusted base. We must trust that the rewriter package we generate from the rewrite rules in fact matches the rewrite rules we want to rewrite with. This involves partially trusting the scraper, the reifier, and the glue code. We must also trust the VM we use for reduction at various points in rewriting. Otherwise, everything is checked by Coq. We do, however, depend on the axiom of function extensionality in one place in the rewriter proof; after spending a couple of hours trying to remove this axiom, we temporarily gave up.

5.D.3 Code from Section 5.3, The Structure of a Rewriter

The expression language e corresponds to the inductive `expr` type defined in module `Compilers.expr` in `rewriter/src/Rewriter/Language/Language.v`.

Code from Subsection 5.3.1, Pattern-Matching Compilation and Evaluation

The pattern -atching compilation step is done by the tactic `CompileRewrites` in `rewriter/src/Rewriter/Rewriter/Rewriter.v`, which just invokes the Gallina definition named `compile_rewrites` with ever-increasing amounts of fuel until it succeeds. (It should never fail for reasons other than insufficient fuel, unless there is a bug in the code.) The workhorse function of this code is `compile_rewrites_step`.

The decision-tree evaluation step is done by the definition `eval_rewrite_rules`, also in the file `rewriter/src/Rewriter/Rewriter/Rewriter.v`. The correctness lemmas are `eval_rewrite_rules_correct` in the file `rewriter/src/Rewriter/Rewriter/InterpProofs.v` and the theorem `wf_eval_rewrite_rules` in `rewriter/src/Rewriter/Rewriter/Wf.v`. Note that the second of these lemmas, not mentioned in the paper, is effectively saying that for two related syntax trees, `eval_rewrite_rules` picks the same rewrite rule for both. (We actually prove a slightly weaker lemma, which is a bit harder to state in English.)

The third step of rewriting with a given rule is performed by the definition `rewrite_with_rule` in `rewriter/src/Rewriter/Rewriter/Rewriter.v`. The correctness proof is `interp_rewrite_with_r`

in `rewriter/src/Rewriter/Rewriter/InterpProofs.v`. Note that the well-formedness-preservation proof for this definition is inlined into the proof `wf_eval_rewrite_rules` mentioned above.

The inductive description of decision trees is `decision_tree` in `rewriter/src/Rewriter/Rewriter.v`.

The pattern language is defined as the inductive pattern in `rewriter/src/Rewriter/Rewriter.v`. Note that we have a `Raw` version and a typed version; the pattern-matching compilation and decision-tree evaluation of Aehlig, Haftmann, and Nipkow [5] is an algorithm on untyped patterns and untyped terms. We found that trying to maintain typing constraints led to headaches with dependent types. Therefore when doing the actual decision-tree evaluation, we wrap all of our expressions in the dynamically typed `rawexpr` type and all of our patterns in the dynamically typed `Raw.pattern` type. We also emit separate inductives of identifier codes for each of the `expr`, `pattern`, and `Raw.pattern` type families.

We partially evaluate the partial evaluator defined by `eval_rewrite_rules` in the tactic `make_rewrite_head` in `rewriter/src/Rewriter/Reify.v`.

Code from Subsection 5.3.2, Adding Higher-Order Features

The type NbE_t mentioned in this paper is not actually used in the code; the version we have is described in Subsection 5.4.2 as the definition `value'` in `rewriter/src/Rewriter/Rewriter.v`.

The functions `reify` and `reflect` are defined in `rewriter/src/Rewriter/Rewriter.Rewriter.v` and share names with the functions in the paper. The function `reduce` is named `rewrite_bottomup` in the code, and the closest match to `NbE` is `rewrite`.

5.D.4 Code from Section 5.4, Scaling Challenges

Code from Subsection 5.4.1, Variable Environments Will Be Large

The inductives `type`, `base_type` (actually the inductive type `base.type.type` in the supplemental code), and `expr`, as well as the definition `Expr`, are all defined in `rewriter/src/Rewriter/Language/Language.v`. The definition `denoteT` is the fixpoint `type.interp` (the fixpoint `interp` in the module `type`) in `rewriter/src/Rewriter/Language/Language.v`. The definition `denoteE` is `expr.interp`, and `DenoteE` is the fixpoint `expr.Interp`.

As mentioned above, `nbeT` does not actually exist as stated but is close to `value'` in `rewriter/src/Rewriter/Rewriter.Rewriter.v`. The functions `reify` and `reflect` are defined in `rewriter/src/Rewriter/Rewriter.Rewriter.v` and share names with the functions in the paper. The actual code is somewhat more complicated than the version presented in the paper, due to needing to deal with converting well-typed-by-construction expressions to dynamically typed expressions for use in decision-tree

evaluation and also due to the need to support early partial evaluation against a concrete decision tree. Thus the version of `reflect` that actually invokes rewriting at base types is a separate definition `assemble_identifier_rewriters`, while `reify` invokes a version of `reflect` (named `reflect`) that does not call rewriting. The function named `reduce` is what we call `rewrite_bottomup` in the code; the name `Rewrite` is shared between this paper and the code. Note that we eventually instantiate the argument `rewrite_head` of `rewrite_bottomup` with a partially evaluated version of the definition named `assemble_identifier_rewriters`. Note also that we use `fuel` to support `do_again`, and this is used in the definition `repeat_rewrite` that calls `rewrite_bottomup`.

The correctness theorems are `InterpRewrite` in `rewriter/src/Rewriter/Rewriter/InterpProofs.v` and `Wf_Rewrite` in `rewriter/src/Rewriter/Rewriter/Wf.v`.

Packages containing rewriters and their correctness theorems are in the record `VerifiedRewriter` in `rewriter/src/Rewriter/Rewriter/ProofsCommon.v`; a package of this type is then passed to the tactic `Rewrite_for_gen` from `rewriter/src/Rewriter/Rewriter/AllTactics.v` to perform the actual rewriting. The correspondence of the code to the various steps in rewriting is described in the second list of Section 5.D.2.

Code from Subsection 5.4.2, Subterm Sharing is Crucial

To run the P-256 example in the copy of Fiat Cryptography attached as a code supplement, after building the library, run the code

```
Require Import Crypto.Rewriter.PerfTesting.Core.
Require Import Crypto.Util.Option.

Import WordByWordMontgomery.
Import Core.RuntimeDefinitions.

Definition p : params
  := Eval compute in invert_Some
    (of_string "2^256-2^224+2^192+2^96-1" 64).

Goal True.
(* Successful run: *)
Time let v := (eval cbv
  -[Let_In
    runtime_nth_default
    runtime_add
    runtime_sub
    runtime_mul
    runtime_opp
    runtime_div
    runtime_modulo
```

```

RT_Z.add_get_carry_full
RT_Z.add_with_get_carry_full
RT_Z.mul_split]
in (GallinaDefOf p)) in
idtac.
(* Unsuccessful OOM run: *)
Time let v := (eval cbv
-[ (*Let_In*)
  runtime_nth_default
  runtime_add
  runtime_sub
  runtime_mul
  runtime_opp
  runtime_div
  runtime_modulo
  RT_Z.add_get_carry_full
  RT_Z.add_with_get_carry_full
  RT_Z.mul_split]
in (GallinaDefOf p)) in
idtac.
Abort.

```

The UnderLets monad is defined in the file `rewriter/src/Rewriter/Language/UnderLets.v`.

The definitions `nbeT'`, `nbeT`, and `nbeT_with_lets` are in `rewriter/src/Rewriter/Rewriter/Rewriter.v` and are named `value'`, `value`, and `value_with_lets`, respectively.

Code from Subsection 5.4.3, Rules Need Side Conditions

The “variant of pattern variable that only matches constants” is actually special support for the reification of `ident.literal` (defined in the module `RewriteRuleNotations` in `rewriter/src/Rewriter/Language/Pre.v`) threaded throughout the rewriter. The apostrophe notation `'` is also introduced in the module `RewriteRuleNotations` in `rewriter/src/Rewriter/Language/Pre.v`. The support for side conditions is handled by permitting rewrite-rule-replacement expressions to return `option expr` instead of `expr`, allowing the function `expr_to_pattern_and_replacement` in the file `rewriter/src/Rewriter/Rewriter/Reify.v` to fold the side conditions into a choice of whether to return `Some` or `None`.

Code from Subsection 5.4.4, Side Conditions Need Abstract Interpretation

The abstract-interpretation pass is defined in `fiat-crypto/src/AbstractInterpretation/`, and the rewrite rules handling abstract-interpretation results are the Gallina definitions `arith_with_casts_rewrite_rulesT`, in addition to `strip_literal_casts_rewrite_rulesT`,

in addition to `fancy_with_casts_rewrite_rulesT`, and finally in addition to `mul_split_rewrite_rule` all defined in `fiat-crypto/src/Rewriter/Rules.v`.

The `clip` function is the definition `ident.cast` in `fiat-crypto/src/Language/PreExtra.v`.

5.D.5 Code from Section 5.5, Evaluation Code from Subsection 5.5.1, Microbenchmarks

This code is found in the files in `rewriter/src/Rewriter/Rewriter/Examples/`. We ran the microbenchmarks using the code in `rewriter/src/Rewriter/Rewriter/Examples/PerfTesting/Harness.v` together with some Makefile cleverness. The file names correspond to the section titles in Section 5.A.

Code from Subsection 5.5.2, Macrobenchmark: Fiat Cryptography

The rewrite rules are defined in `fiat-crypto/src/Rewriter/Rules.v` and proven in the file `fiat-crypto/src/Rewriter/RulesProofs.v`. They are turned into rewriters in the various files in `fiat-crypto/src/Rewriter/Passes/`. The shared inductives and definitions are defined in the Coq source files `fiat-crypto/src/Language/IdentifiersBasicGENERATED.v`, `fiat-crypto/src/Language/IdentifiersGENERATED.v`, and `fiat-crypto/src/Language/IdentifiersGENERATEDProofs.v`. Note that we invoke the subtactics of the `Make` command manually to increase parallelism in the build and to allow a shared language across multiple rewriter packages.

Chapter 6

Engineering Challenges in the Rewriter

premature optimization is the root of all evil

— Donald Knuth

[**TODO:** Better chapter title?] [**TODO:** consider replacing “pain” with more formal words like “overhead”, “bottleneck”, etc (Adam says it’s possibly too informal)] Chapter 5 discussed in detail our framework for building verified partial evaluators, going into the context, motivation, and the techniques used to put the framework together. However, there was a great deal of engineering effort that went into building this tool which we glossed over. Much of the engineering effort was mundane, and we elide the details entirely. However, we believe some of the engineering effort serves as a good case-study for the difficulties of building proof-based systems at scale. This chapter is about exposing the details relevant to understanding how the bottlenecks and principles identified elsewhere in this thesis played out in designing and implementing this tool. Note that many of the examples and descriptions in this chapter are highly technical, and we expect the discussion will only be of interest to the motivated reader, familiar with Coq, who wants to see more concrete non-toy examples of the bottlenecks and principles we’ve been describing; other readers are encouraged to skip this chapter.

While the core rewriting engine of the framework is about 1 300 lines of code, and early simplified versions of the core engine were only about 150 lines of code¹, the correctness proofs take nearly another 8 000 lines of code! As such, this tool, developed to solve performance scaling issues in verified syntax transformation, itself serves as a

¹See <https://github.com/JasonGross/fiat-crypto/blob/3b3e926e4186caa1a4003c81c65dad0a1c04b43d/src/E> for the file `src/Experiments/RewriteRulesSimpleNat.v` from the branch `experiments-small-rewrite-rule-compilation` on JasonGross/fiat-crypto on GitHub.

good case study of some of the pain that arises when scaling proof-based engineering projects.

Our discussion in this section is organized by the conceptual structure of the normalization and pattern matching compilation engine; we hope that organizing the discussion in this way will make the examples more understandable, motivated, and incremental. We note, however, that many of the challenges fall into the same broad categories that we've identified earlier in this thesis: issues arising from the power and (mis)use of dependent types, as introduced in Subsection 1.2.1 (Dependent Types: What? Why? How?); and issues arising from API mismatches, as described in Chapter 3 (Design-based fixes).

6.1 Pre-Reduction

The two biggest underlying causes of engineering challenges are expression API mismatch, which we'll discuss in Section 6.2 (NbE vs. Pattern Matching Compilation: Mismatched Expression APIs and Leaky Abstraction Barriers), and our desire to reduce away known computations in the rewriting engine once and for all when compiling rewriting rules, rather than again and again every time we perform a rewrite. In practice, performing this early reduction nets us an approximately 2 \times speed-up.

6.1.1 What does this reduction consist of?

Recall from Subsection 5.3.1 that the core of our rewriting engine consists of three steps:

1. The first step is pattern-matching compilation: we must compile the lefthand sides of the rewrite rules to a decision tree that describes how and in what order to decompose the expression, as well as describing which rewrite rules to try at which steps of decomposition.
2. The second step is decision-tree evaluation, during which we decompose the expression as per the decision tree, selecting which rewrite rules to attempt.
3. The third and final step is to actually rewrite with the chosen rule.

The first step is performed once and for all; it depends only on the rewrite rules, and not on the expression we are rewriting in. The second and third steps do, in fact, depend on the expression being rewritten, and it is in these steps that we seek to eliminate needless work early.

The key insight, which allows us to perform this precompilation at all, is that the most of the decisions we seek to eliminate depend only on the *head identifier* of

any application.² We thus augment the $\text{reduce}(c)$ constant case of Figure 5-2 in Subsection 5.3.2 by first η -expanding the identifier, before proceeding to η -expand the identifier application and perform rewriting with rewrite-head once we have an η -long form.

Now that we know what the reduction consists of, we can now discuss what goes in to making the reduction possible, and the engineering challenges that arise.

6.1.2 CPS

Due to the pervasive use of Gallina **match** statements on terms which are not known during this compilation phase, we need to write essentially all of the decision-tree-evaluation code in continuation-passing style. This causes a moderate amount of pain, distributed over the entire rewriter.

[TODO: find a better transition] The way that CPS permits reduction under blocked **match** statements is essentially the same as the way it permits reduction of functions in the presence of unreduced **let** binders in Subterm Sharing is Crucial). Consider the expression

```
option_map List.length (option_map (λ x. List.repeat x 5) y)
```

where `option_map : (A → B) → option A → option B` maps a function over an option, and `List.repeat x n` creates a list consisting of `n` copies of `x`. If we fully reduce this term, we get the Gallina term

```
match
  match y with
  | Some x => Some [x; x; x; x; x]
  | None => None
end
with
| Some x =>
  Some
    ((fix Ffix (x0 : list _) : nat :=
      match x0 with
      | [] => 0
      | _ :: x2 => S (Ffix x2)
    end) x)
```

²In order to make this simplification, we need to restrict the rewrite rules we support a little bit. In particular, we only support rewrite rules operating on η -long applications of concrete identifiers to arguments. This means that we cannot support identifiers with variable arrow structure (e.g., a variadic `curry` function) nor do we support rewriting things like `List.map f` to `List.map g`—we only support rewriting `List.map f xs` to `List.map g ys`.

```

| None => None
end

```

Consider now a CPS'd version of `option_map`:

```

Definition option_map_cps {A B} (f : A → B) (x : option A)
  : ∀ {T}, (option B → T) → T
  := λ T cont.
    match x with
    | Some x => cont (Some (f x))
    | None => cont None
  end.

```

Then we could write the somewhat more confusing term

```
option_map_cps (λ x. List.repeat x 5) y (option_map List.length)
```

whence reduction gives us

```

match y with
| Some _ => Some 5
| None => None
end

```

So we see that rewriting terms in continuation-passing style allows reduction to proceed without getting blocked on unknown terms.

Note that if we wanted to pass this list length into a further continuation, we'd need to instead write a term like

```

λ cont.
  option_map_cps (λ x. List.repeat x 5) y
    (λ ls. option_map_cps List.length ls cont)

```

which reduces to

```

λ cont. match y with
| Some _ => cont (Some 5)
| None => cont None
end

```

6.1.3 Type Codes

The pattern-matching compilation algorithm of Aehlig, Haftmann, and Nipkow [5] does not deal with types. In general, unification of types is somewhat more complicated than unification of terms, because terms are indexed over types. We have two options, here:

1. We can treat terms and types as independent and untyped, simply collecting a map of unification variables to types, checking non-linear occurrences (such as the types in `@fst ?A ?B (@pair ?A ?B ?x ?y)`) for equality, and run a typechecking pass afterwards to reconstruct well-typedness. In this case, we would consider the rewriting to have failed if the replacement is not well-typed.
2. We can perform matching on types first, taking care to preserve typing information, and then perform matching on terms afterwards, taking care to preserve typing information.

The obvious tradeoff between these options is that the former option requires doing more work at runtime, because we end up doing needless comparisons that we could know in advance will always turn out a particular way. Importantly, note that Coq's reduction will not be able to reduce away these runtime comparisons; reduction alone is not enough to deduce that a boolean equality function defined by recursion will return true when passed identical arguments, if the arguments are not also concrete terms.

Following standard practice in dependently-typed languages, we chose the second option. We now believe that this was a mistake, as it's fiendishly hard to deconstruct the expressions in a way that preserves enough typing information to completely avoid the need to compare type codes for equality and cast across proofs. For example, to preserve typing information when matching for `@fst ?A ?B (@pair ?A ?B ?x ?y)`, we would have to end up with the following `match` statement. Note that the reader is not expected to understand this statement, and the author was only able to construct it with some help from Coq's typechecker.

```
| App f v =>
let f :=
  match f in expr t return option (ident t) with
  | Ident idc => Some idc
  | _ => None
end in
match f with
| Some maybe_fst =>
  match v in expr s return ident (s -> _) -> _ with
  | App f y =>
```

```

match f in expr _s
  return
    match _s with arrow b _ => expr b | _ => unit end
    -> match _s with arrow _ ab => ident (ab -> _) | _ => unit end
    -> _
  with
  | App f x =>
  let f :=
    match f in expr t return option (ident t) with
    | Ident idc => Some idc
    | _ => None
  end in
  match f with
  | Some maybe_pair =>
    match maybe_pair in ident t
    return
      match t with arrow a _ => expr a | _ => unit end
      -> match t with arrow a (arrow b _) => expr b | _ => unit end
      -> match t with arrow a (arrow b ab) => ident (ab -> _) | _ => unit end
      -> _
    with
    | @pair a b =>
      fun (x : expr a) (y : expr b) (maybe_fst : ident _) =>
        let is_fst := match maybe_fst with fst => true | _ => false end in
        if is_fst
        then ... (* now we can finally do something with a, b, x, and y *)
        else ...
        | _ => ...
      end x
    | None => ...
  end
  | _ => ...
  end y
  | _ => ...
end maybe_fst
| None => ...
end

```

This is quite the mouthful.

Furthermore, there are two additional complications. First, this sort of match expression must be generated *automatically*. Since pattern-matching evaluation happens on *lists* of expressions, we'd need to know exactly what each match reveals about the types of all other expressions in the list. Additionally, in order to allow reduction to

happen where it should, we need to make sure to match the head identifier *first*, without convoying it across matches on unknown variables. Note that in the code above, we did not follow this requirement, as it would complicate the **return** clauses even more (presuming we wanted to propagate typing information as we'd have to in the general case rather than cutting corners). The convoy pattern, for those unfamiliar with it, is explained in detail in Chapter 8 (“More Dependent Types”) of *Certified Programming with Dependent Types*. [48]

Second, trying to prove anything about functions written like this is an enormous pain. Because of the intricate dependencies in typing information involved in the convoy pattern, Coq's **destruct** tactic is useless. The **dependent destruction** tactic is sometimes able to handle such goals, but even when it can, it often introduces a dependency on the axiom `JMeq_eq`, which is equivalent to assuming *uniqueness of identity proofs* (UIP), that all proofs of equality are equal—note that this contradicts, for example, the popular univalence axiom of homotopy type theory [194]. In order to prove anything about such functions without assuming UIP, the proof effectively needs to replicate the complicated **return** clauses of the function definition. However, since they are not to be replicated exactly, but merely be generated from the same insights, such proof terms often have to be written almost entirely by hand. These proofs are furthermore quite hard to maintain, as even small changes in the structure of the function often require intricate changes in the proof script.

Due to a lack of foresight and an unfortunate reluctance to take the design back to the drawing board after we already had working code, we ended up mixing these two approaches, getting, not quite the worst of both worlds, but definitely a significant fraction of the pain of both worlds: We must deal with both the pain of indexing our term unification information over our type unification information, and we must still insert typecasts in places where we have lost the information that the types will line up.

6.1.4 How Do We Know What We Can Unfold?

Coq's built-in reduction is somewhat limited, especially when we want it to have reasonable performance. This is, after all, a large part of the problem this tool is intended to solve.

In practice, we make use of three reduction passes; that we cannot interleave them is a limitation of the built-in reduction:

1. First, we unfold everything except for a specific list of constants; these constants are the ones that contain computations on information not fully known at pre-evaluation time.
2. Next, we unfold all instances of a particular set of constants; these constants are the ones that we make sure to only use when we know that inlining them

won't incur extra overhead.

3. Finally, we use `cbn` to simplify a small set of constants in only the locations that these constants are applied to constructors.

Ideally, we'd either be able to do the entire simplification in the third step, or we'd be able to avoid the third step entirely. Unfortunately, Coq's reduction is not fast enough to do the former, and the latter requires a significant amount of effort. In particular, the strategy that we'd need to follow is to have two versions of every function which sometimes computes on known data and sometimes computes on unknown data, and we'd need to track in all locations which data is known and which data is unknown.

We already track known and unknown data to some extent (see, for example, the `known` argument to the `rIdent` constructor discussed below). Additionally, we have two versions of a couple of functions, such as the bind function of the option monad, where we decide which to use based on, e.g., whether or not the option value that we're binding will definitely be known at pre-reduction time.

Note that tracking this sort of information is non-trivial, as there's no help from the typechecker.

We'll come back to this in Subsection 6.4.1.

6.2 NbE vs. Pattern Matching Compilation: Mismatched Expression APIs and Leaky Abstraction Barriers

[**TODO:** should I be consistent about naming syntax that's well-typed by construction? I call it "intrinsically-typed syntax", "syntax that is well-typed by construction", "intrinsically-well-typed syntax", and "type-indexed syntax"...] We introduced normalization by evaluation (NbE) [30] in Subsection 5.1.3 and expanded on it in Subsection 5.3.2 as a way to support higher-order reduction of λ -terms. The termination argument for NbE proceeds by recursion on the type of the term we're reducing. In particular, the most natural way to define these functions in a proof assistant is to proceed by structural recursion on the type of the term being reduced. This feature suggests that using intrinsically-typed syntax is more natural for NbE, and we saw in Section 4.1.3 that denotation functions are also simpler on syntax that is well-typed by construction.

However, the pattern-matching compilation algorithm of Maranget [131] inherently operates on untyped syntax. We thus have four options:

- (1) use intrinsically-well-typed syntax everywhere, paying the cost in the pattern-

matching compilation and evaluation algorithm;

- (2) use untyped syntax in both NbE and rewriting, paying the associated costs in NbE, denotation, and in our proofs;
- (3) use intrinsically-well-typed syntax in most passes, and untyped syntax for pattern matching compilation;
- (4) invent a pattern-matching compilation algorithm that is well-suited to type-indexed syntax.

We ultimately chose option (3). I was not clever enough to follow through on option (4), and while options (1) and (2) are both interesting, option (3) seemed to follow the well-established convention of using whichever datatype is best-suited to the task at hand. As we'll shortly see, all of these options come with significant costs, and (3) is not as obviously a good choice as it might seem at first glance.

6.2.1 Pattern-Matching Evaluation on Type-Indexed Terms

While the cost of performing pattern-matching compilation on type-indexed terms is noticeable, it's relatively insignificant compared to the cost of evaluating decisions trees directly on type-indexed terms. In particular, pattern-matching compilation effectively throws away the type information whenever it encounters it; whether we do this early or late does not matter much, and we only perform this compilation once for any given set of rewrite rules.

By contrast, evaluation of the decision tree needs to produce *term ASTs* that are used in rewriting, and hence we need to preserve type information in the input. Recall from Subsection 5.3.1 that decision-tree evaluation operates on lists of terms. Here already we hit our first snag: if we want to operate on well-typed terms, we must index our lists over a list of types. This is not so bad, but recall also from Subsection 5.3.1 that decision trees contain four constructors:

- **TryLeaf k onfailure:** Try the k^{th} rewrite rule; if it fails, keep going with `onfailure`.
- **Failure:** Abort; nothing left to try.
- **Switch `icases` `app_case` `default`:** With the first element of the vector, match on its kind; if it is an identifier matching something in `icases`, which is a list of pairs of identifiers and decision trees, remove the first element of the vector and run that decision tree; if it is an application and `app_case` is not `None`, try the `app_case` decision tree, replacing the first element of each vector with the two elements of the function and the argument it is applied to; otherwise, do not modify the vectors and use the `default` decision tree.

- **Swap i cont:** Swap the first element of the vector with the i^{th} element (0-indexed) and keep going with **cont**.

The first two constructors are not very interesting, as far as overhead goes, but the third and fourth constructors are quite painful.

Note that the type of `eval_decision_tree` would be something like $\forall \{T : \text{Type}\} (d : \text{decision_tree}) (ts : \text{list type}) (es : \text{exprlist ts}) (K : \text{exprlist ts} \rightarrow \text{option T})$, `option T`.

We cover the **Swap** case first, because it is simpler. To perform a **Swap**, we must exchange two elements of the type-indexed list. Hence we need both two swap the elements of the list of types, and then to have a separate, dependently-typed swap function for the vector of expressions. Moreover, since we need to undo the swapping inside the continuation [**TODO: does this need more explanation or code?**], we must have an *separate* unswap function on expression vectors which goes from a swapped type list to the original one. We could instead elide the swap node, but then we could no longer use matching, `hd`, and `t1` to operate on the expressions, and would instead need special operations to do surgery in the middle of the list, in a way that preserves type-indexing.

To perform a **Switch**, we must break apart the first element of our type-indexed list, determining whether it is an application, and identifier, or other. Note that even with dependent types, we cannot avoid needing a failure case for when the type-indexed list is empty, even though such a case should never occur because good decision trees will never have a **Switch** node after consuming the entire vector of expressions. This failure case cannot be avoided because there is no type-level relation between the expression vector and the decision tree. This mismatch—the need to include failure cases that one might expect to be eliminated by dependent typing information—is a sign that the amount of dependency in the types is wrong. It may be too little, whence the developer should see if there is a way to incorporate the lack-of-error into the typing information (which in this case would require indexing the type of the decision tree over the length of the vector). It may alternatively be to much dependent typing, and the developer might be well-served by removing more dependency from the types and letting more things fall into the error case.

After breaking apart the first element, we must convoy the continuation across the **match** statement so that we can pass an expression vector of the correct type to the continuation `K`. In code, this branch might look something like [**TODO: Note that Adam found this code hard to understand; should more prose be added to clarify it?**]

```
...
| Switch icases app_case default
=> match es in exprlist ts
```

```

    return (exprlist ts → option T) → option T
  with
  | [] => λ _, None
  | e :: es
    => match e in expr t
      return (exprlist (t :: ts) → option T) → option T
      with
      | App s d f x => λ K,
        let K' : exprlist ((s → d) :: s :: ts)
          (* new continuation to pass on recursively *)
          := λ es', K (App (hd es') (hd (tl es'))) :: tl (tl es')) in
          ... (* do something with app_case *)
      | Ident t idc => λ K,
        let K' : exprlist ts
          (* new continuation to pass on recursively *)
          := λ es', K (Ident idc :: es') in
          ... (* do something with icases *)
      | _ => λ K, ... (* do something with default *)
    end
  end K
...

```

Note that `hd` and `tl` *must* be type-indexed, and we *cannot* simply match on `es'` in the `App` case; there is no way to preserve the connection between the types of the first two elements of `es'` inside such a `match` statement.

This may not look too bad, but it gets worse. Since the `match` on `e` will not be known until we are actually doing the rewriting on a concrete expression, and the continuation is convoyed across this `match`, there is no way to evaluate the continuation during compilation of rewrite rules. If we don't want to evaluate the continuation early, we'd have to be very careful not to duplicate it across all of the decision tree evaluation cases, as we might otherwise incur a super-linear runtime factor in the number of rewrite rules. As noted in Section 6.1, our early reduction nets us a 2× speedup in runtime of rewriting, and is therefore relatively important to be able to do. [TODO: redo performance experiments here, maybe insert a plot]

Here we see something interesting, which does not appear to be as much of a concern in other programming languages: the representation of our data forces our hand about how much efficiency can be gained from precomputation, even when the representation choices are relatively minor.

6.2.2 Untyped Syntax in NbE

There is no good way around the fact that NbE requires typing information to argue termination. Since NbE will be called on subterms of the overall term, even if we use syntax that is not guaranteed to be type-correct, we must still store the type information in the nodes of the AST.

Furthermore, as we say in Section 4.1.3 (de Bruijn Indices), converting from untyped syntax to intrinsically-typed syntax, as well as writing a denotation function, requires either that all types be non-empty, or that we carry around a proof of well-typedness to use during recursion. [QUESTION FOR ADAM: Is there a good reference for these sorts of issues? Are they well-known? Well-studied?] As discussed in Chapter 3 and specifically in Section 3.2 (When And How To Use Dependent Types Painlessly), needing to mix proofs with programs is often a big warning flag, unless the mixing can be hidden behind a well-designed API. However, if we are going to be hiding the syntax behind an API of being well-typed, it seems like we might as well just use intrinsically well-typed syntax, which naturally inhabits that API. Furthermore, unlike in many cases where the API is best treated as opaque everywhere, here the API mixing proofs and programs needs to have adequate behavior under reduction, and ought to have good behavior even under partial reduction. This severely complicates the task of building a good abstraction barrier, as we not only need to ensure that the abstraction barrier does not need to be broken in the course of term-building and typechecking, but we must also ensure that the abstraction barrier can be broken in a principled way via reduction without introducing significant overhead.

6.2.3 Mixing Typed and Untyped Syntax

The third option is to use whichever datatype is most naturally suited for each pass, and to convert between them as necessary. This is the option that we ultimately chose, and the one, we believe, that would be most natural to choose to engineers and developers coming from non-dependently-typed languages.

There are a number of considerations that arose when fleshing out this design, and a number of engineering-pain-points that we encountered. The theme to all of these, as in Chapter 3, is that imperfectly opaque abstraction barriers cause headaches in a non-local manner.

We got lucky, in some sense, that the rewriting pass *always* has a well-typed default option: do no rewriting. Hence we do not need to worry about carrying around proofs of well-typedness, and this avoids some of the biggest issues described in Untyped Syntax in NbE.

The biggest constraint driving our design decisions is that we need conversion between the two representations to be $\mathcal{O}(1)$; if we need to walk the entire syntax tree to convert between typed and untyped representations at every rewriting location, we'll incur

quadratic overhead in the size of the term being rewritten. We can actually relax this constraint a little bit: by designing the untyped representation to be completely evaluated away during the compilation of rewrite rules, we can allow conversion from the untyped syntax to the typed syntax to walk any part of the term that already needed to be revealed for rewriting, giving us amortized constant time rather than truly constant time. [TODO: is this last sentence understandable?] As such, we need to be able to embed well-typed syntax directly into the non-type-indexed representation at cost $\mathcal{O}(1)$.

As the entire purpose of the untyped syntax is to (a) allow us to perform matching on the AST to determine which rewrite rule to use, and furthermore (b) allow us to reuse the decomposition work so as to avoid needing to decompose the term multiple times, we need an inductive type which can embed PHOAS expressions, and has separate nodes for the structure that we need, namely application and identifiers:

```
Inductive rawexpr : Type :=
| rIdent (known : bool) {t} (idc : ident t) {t'} (alt : expr t')
| rApp (f x : rawexpr) {t} (alt : expr t)
| rExpr {t} (e : expr t)
| rValue {t} (e : NbEt t).
```

There are three perhaps-unexpected things to note about this inductive type, which we will discuss in later subsections:

1. The constructor `rValue` holds an `NbE`-value of the type `NbEt` introduced in Subsection 5.3.2. We will discuss this in Section 6.7 (Delayed Rewriting in Variable Nodes).
2. The constructors `rIdent` and `rExpr` hold “alternate” PHOAS expressions. We will discuss this in Subsection 6.4.2 (Revealing “Enough” Structure).
3. The constructor `rIdent` has an extra boolean `known`. We will discuss this in Section 6.4.1 (The `known` argument).

With this inductive type in hand, it’s easy to see how `rExpr` allows us $\mathcal{O}(1)$ embedding of intrinsically typed `exprs` into untyped `rawexprs`.

While it’s likely that sufficiently good abstraction barriers around this datatype would allow us to use it with relatively little pain, we did not succeed in designing good enough abstraction barriers. The bright side of this failure is that we now have a number of examples for this thesis of ways in which inadequate abstraction barriers cause pain.

We will discuss the many issues that arise from leaks in this abstraction barrier in the upcoming subsections.

6.2.4 Pattern Matching Compilation Made For Intrinsically-Typed Syntax

The cost of this fourth option is the cleverness required to come up with a version of the pattern matching compilation which, rather than being hindered by types in its syntax, instead puts them to good use. Lacking this cleverness, we were unable to pay the requisite cost, and hence have not much to say in this section.

6.3 Patterns with Type Variables – The Three Kinds of Identifiers

We have one final bit of infrastructure to explain and motivate before we have enough of the structure sketched out to give all of the rest of the engineering challenges: representing the identifiers. Recall from Subsection 5.2.1 (Our Approach in Nine Steps) that we automatically emit an inductive type describing all available primitive functions.

When deciding how to represent identifiers, there are roughly three options we have to choose from:

1. We could use an untyped representation of identifiers, such as Coq strings (as in Anand et al. [16], for example), or integers indexing into some finite map.
2. We could index the expression type over a finite map of valid identifiers, and use dependent typing to ensure that we only have well-typed identifiers.
3. We could have a fixed set of valid identifiers, using types to ensure that we have only valid expressions.

The first option results in expressions that are not always well-typed. As discussed in Chapter 3 and seen in the preceding sections, having leaky abstraction barriers is often worse than having none at all, and we expect that having partially-well-typed expressions would be no exception.

The second option is probably the way to go if we want truly extensible identifier-sets. There are two issues. First, this adds a linear overhead in the number of identifiers—or more precisely, in the total size of the types of the identifiers—because every AST node will store a copy of the entire finite map. Second, because our expression syntax is simply typed, polymorphic identifiers pose a problem. To support identifiers like `fst` and `snd`, which have types $\forall A B, A * B \rightarrow A$ and $\forall A B, A * B \rightarrow B$

respectively, we must either replicate the identifiers with all of the ways they might be applied, or else we must add support in our language for dependent types or for explicit type polymorphism.

Instead, we chose to go with the third option, which we believe is the simplest. The inductive type of identifiers is indexed over the type of the identifier, and type polymorphism is expressed via meta-level arguments to the constructor. So, for example, the identifier code for `fst` takes two type-code arguments `A` and `B`, and has type `ident (A * B → A)`. Hence all fully-applied identifier codes have simple types (such as `A * B → A`), and our inductive type still supports polymorphic constants. An additional benefit of this approach is that unification of identifiers is just pattern matching in Gallina, and hence we can rely on the pattern-matching compilation schemes of Coq’s fast reduction machines, or the OCaml compiler itself, to further speed up our rewriting.

Aside: Why Use Pattern Matching Compilation At All? Given the fact that, after pre-reduction, there is no trace of the decision tree remaining, one might ask why we use pattern matching compilation at all, rather than just leaving it to the pattern-matching compiler of Coq or OCaml to be performant. We have three answers to this question.

The first, perhaps most honest answer, is that it is a historical accident; we prematurely optimized this part of the rewriting engine when writing it.

The second answer is that pattern matching compilation is a good abstraction barrier for factoring out the work of revealing enough structure from the work of unifying a pattern with an expression. Said another way, even though we reduce away the decision tree and its evaluation, there is basically no wasted work; removing pattern matching compilation while preserving all the benefits would effectively just be inlining all of the functions, and there would be no dead code revealed by this inlining.

The third and final answer is that it allows us to easily prune useless work. The pattern matching compilation algorithm naturally prunes away patterns that can be known to not work, given the structure that we’ve revealed. By contrast, if we just record what information we’ve already revealed as we’re performing pattern unification, it’s quite tricky to avoid decomposition which can be known to be useless based on only the structure that’s been revealed already.

Consider, for example, rewriting with two rules whose left-hand-sides are $x + (y + 1)$ and $(a+b)+(c*2)$. When revealing structure for the first rewrite rule, the engine will first decompose the (unknown) expression into the application of the `+` identifier to two arguments, and then decompose the second argument into the application of the `+` identifier to two arguments, and then finally decompose the second inner argument into a literal identifier to check if it is the literal 1. If the decomposition succeeds, but

the literal is not 1 (or if the second inner argument is not a literal at all), then rewriting will fall back to the second rewrite rule. If we are doing structure decomposition in the naïve way, we will then decompose the outer first argument (bound to x in the first rewrite rule) into the application of the identifier $+$ to two arguments. We will then attempt to decompose the secound outer argument into the application of the identifier $*$ to two arguments. Since there is no way an identifier can be both $+$ and $*$, this decomposition will fail. However, we could have avoided doing the work of decomposing x into $a + b$ by realizing that the second rewrite rule is incompatible with the first; this is exatly what pattern-matching compilation and decision-tree evaluation does. [TODO: Does this need more explanation? Is it understandable?] [TODO: be consistent about “pattern matching compilation” vs “pattern-matching compilation”?]

Pattern Matching For Rewriting We now arrive at the question of how to do pattern matching for rewriting with identifiers. We want to be able to support type variables, for example to rewrite `@fst ?A ?B (@pair ?A ?B ?x ?y)` to x . While it would arguably be more elegant to treat term and type variables identically, doing this would require a language supporting dependent types, and we are not aware of any extension of PHOAS to dependent types. Extensions of HOAS to dependent types are known [134], but the obvious modifications of such syntax that in the simply-typed case turn HOAS into PHOAS result in infinite self-referential types in the dependently-typed case.

As such, insofar as we are using intrinsically well-typed syntax at all, we need to treat type variables separately from term variables. We need three different sorts of identifiers:

- identifiers whose types contain no type variables, for use in external-facing expressions and the denotation function,
- identifiers whose types are permitted to contain type variables, for use in patterns, and
- identifiers with no type information, for use in pattern-matching compilation.

The first two are relatively self-explanatory. The third of these is required because pattern-matching compilation proceeds in an untyped way; there’s no obvious place to keep the typing information associated to identifiers in the decision tree, which must be computed before we do any unification, type variables or otherwise. [TODO: is this sufficiently understandable?]

We could, in theory, use a single inductive type of type codes for all three of these. We could parameterize the inductive of type codes over the set of free type variables (or even just over a boolean declaring whether or not type variables are allowed), and

conventionally use the type code for unit in all type-code arguments when building decision trees.

This sort of reuse, however, is likely to introduce more problems than it solves.

The identifier codes used in pattern-matching compilation must be untyped, to match the decision we made for expressions in Section 6.2. Having them conventionally be typed pattern codes instantiated with unit types is, in some sense, just more opportunity to mess up and try to inspect the types when we really shouldn't. There is a clear abstraction barrier here, of having these identifier codes not carry types, and we might as well take advantage of that and codify the abstraction barrier in our code.

The question of type variables is more nuanced. If we are only tracking whether or not a type is allowed to have type variables, then we might as well use two different inductive types; there is not much benefit to indexing the type codes over a boolean rather than having two copies of the inductive, for there's not much that can be done generically in whether or not type variables are allowed. Note also that we must track at least this much information, for identifiers in expressions passed to the denotation function must not have uninstantiated type variables, and identifiers in patterns must be permitted to have uninstantiated type variables.

However, there is some potential benefit to indexing over the set of uninstantiated type variables. This might allow us to write type signatures for functions that guarantee some invariants, possibly allowing for easier proofs. However, it's not clear to us where this would actually be useful; most functions already care only about whether or not we permit type variables at all. Our current code in fact performs a poor approximation of this strategy in some places: we index over the entire pattern where indexing over the free variables of the pattern would suffice.

This unneeded indexing causes an enormous amount of pain, and is yet another example of how poorly designed abstraction barriers incur outsized overhead. Rewrite rule replacements are expressed as dependently-typed towers indexed first over the type variables of a pattern, and then again over the term variables. This design is a historical artifact, from when we expected to be writing rewrite rule ASTs by hand rather than reifying them from Gallina, and found the curried towers more convenient to write. [TODO: should I elaborate more on the pain?] This design, however, is absolutely a mistake, especially given the concession we make in Subsection 6.1.3 (Type Codes) to not track enough typing information to avoid all typechecking.

While indexing over only the set of permitted type variables would simplify proofs significantly, we'd benefit even more by indexing only over whether or not we permit type variables at all. None of our proofs are made simpler by tracking the set of permitted type variables.

6.4 Pre-evaluation Revisited

Having built up enough infrastructure to give a bit more in the way of code examples, we now return to the engineering challenges posed by reducing early, first investigated in Section 6.1.

6.4.1 How Do We Know What We Can Unfold?

We can now revisit Subsection 6.1.4 in a bit more detail.

The `known` argument We noted in Subsection 6.2.3 the `known` argument of the `rIdent` constructor of `rawexpr`. This argument is used to track what sorts of operations can be unfolded early. In particular, if a given identifier has no type arguments (for example, addition on \mathbb{Z} s), and we have already matched against it, then when performing further matches to unify with other patterns, we can directly match it against pattern identifiers. By contrast, if the identifier has not yet been matched against, or if it has unknown type arguments, we cannot guarantee that `matches` will reduce. Tracking this information adds a not-insignificant amount of pain to the code.

Consider the following two cases, where we will make use of both `true` and `false` for the `known` argument.

First, let us consider the simpler case of wanting `known` to be `false`. As a toy example, suppose we are rewriting with the rules `@List.map A B f (x::xs) = f x :: List.map f xs` and `@List.map (option A) (option B) (option_map f) (List.map (@Some A) xs) = xs`. When decomposing structure for the first rewrite rule, we will match on the head identifier to see if it is `List.map`. Supposing that the final argument is not a cons cell, we will fall back to the second rewrite rule. While we know that the first identifier is a `List.map`, we do not know its type arguments. Therefore, when we want to try to substitute with the second rewrite rule, we must match on the type structure of the first type argument to `List.map` to see if it is an option, and, if so, extract the underlying type to put into unification data. However, this decomposition will block on the type arguments to `List.map`, so we don't want to fully unfold it during early reduction. Note that the first rewrite rule is not really necessary in this example; the essential point is that we don't want to be unfolding complicated recursive matches on the type structure that are not going to reduce.³

There are two cases where we want to reduce the `match` on an identifier. One of them is when the identifier is known from the initial η -expansion of identifiers discussed in Subsection 6.1.1 (note that this is distinct from the η -expansion of identifier applications), and the identifier has no type arguments.⁴ The other case is when we

³In the current codebase, removing the first rewrite rule would, unfortunately, result in unfolding of the matching on the type structure, due to an oversight in how we compute the `known` argument. See the next footnote for more details.

⁴In our current implementation we don't actually check that the identifier has no type arguments

have tested an identifier against a pattern identifier, and it has no type arguments. In this case, when we eventually get around to collecting unification data for this identifier, we know that we can reduce away the check on this identifier. Whether or not the overhead is worth it in this second case is unclear; the design of this part of the rewriting engine suffers from the lack of a unified picture about what, exactly, is worth reducing, and what is not.

Gratuitous Dependent Types: How much do we actually want to unfold?

When computing the replacement of a given expression, how much do we want to unfold? Here we encounter a case of premature optimization being the root of, if not evil, at least headaches. The simplest path to take here would be to have unification output a map of type-variable indices to types and a map of expression-variable indices to expressions of unknown types. We could then have a function, not to be unfolded early, which substitutes the expressions into some untyped representation of terms, and then performs a typechecking pass to convert back to a well-typed expression.

Instead, we decided to reduce as much as we possibly could. Following the common practice of eager students looking to use dependent types, we defined a dependently typed data structure indexed over the pattern type which holds the mapping of each pattern type variable to a corresponding type. While this mapping cannot be fully computed at rewrite-rule-compilation time—we may not know enough type structure in the `rawexpr`—we can reduce effectively all of the lookups by turning them into matches on this which *can* be reduced. This, unfortunately, complicates our proofs significantly while likely not providing any measurable speedup, serving only as yet another example of the pain induced by needless dependency at the type level.
[TODO: is this enough? does it need more? should we keep it here at all?]

6.4.2 Revealing “Enough” Structure

We noted in Subsection 6.2.3 that the constructors `rIdent` and `rExpr` hold “alternate” PHOAS expressions. We now discuss the reason for this.

Consider the example where we have two rewrite rules: that $(x + y) + 1 = x + (y + 1)$ and that $x + 0 = x$. If we have the expression $(a + b) + 0$, we would first try to match this against $(x + y) + 1$. If we didn’t store the expression $a + b$ as a PHOAS expression, and had it only as a `rawexpr`, then we’d have to retypecheck it, inserting casts as necessary, in order to get a PHOAS expression to return from unification of $a + b$ with x in $x + 0$.

in this case. This is an oversight, and the correct design would be able to distinguish between “this identifier is known and it has no type arguments”, “this identifier is known but it has unknown type arguments”, and “this identifier is completely unknown”. Failure to distinguish these cases does not seem to cause too much trouble, because the way the code is structured luckily ensures that we only match on the type arguments once, and because everything is CPS’d, this matching does not block further reduction.

Instead of incurring this overhead, we store the undecomposed PHOAS expression in the `rawexpr`, allowing us to reuse it when no more decomposition is needed. This does, however, complicate proofs: we need to talk about matching the revealed and unrevealed structure, sometimes just on the type level, and other times on both the term level and the type level.

6.5 Monads: Missing Abstraction Barriers at the Type Level

We introduce in Subsection 5.4.2 the `UnderLets` monad for let-lifting, which we inline into the definition of the NbE_t value type. We use two other monads in the rewriting engine: the option monad to encode possible failure of rewrite rule side-conditions and substitutions, and the CPS monad discussed in Subsection 6.1.2.

Although we introduce a bit of syntactic sugar for monadic binds in an ad-hoc way, we do not fully commit to a monadic abstraction barrier in our code. This lack of principle incurs pain when we have to deal with mismatched monads in different functions, especially when we haven't ordered the monadic applications in a principled way.

The simplest example of this pain is in our mixing of the option and CPS monads in `eval_decision_tree`. The type of `eval_decision_tree` is $\forall \{T : \text{Type}\} (\text{es} : \text{list rawexpr}) (\text{d} : \text{decision_tree}) (K : \mathbb{N} \rightarrow \text{list rawexpr} \rightarrow \text{option } T)$, `option T`. Recall that the function of `eval_decision_tree` is to reveal structure on the list of expressions `es` by evaluating the decision tree `d`, calling `K` to perform rewriting with a given rewrite rule (referred to by index) whenever it hits a leaf node, and continuing on when `K` fails with `None`. What is the correctness condition for `eval_decision_tree`?

We need two correctness conditions. One of them is that, if `eval_decision_tree` succeeds at all, it is equivalent to calling `K` on some index with some list of expressions which is appropriately equivalent to `es`. (See Subsection 6.7.1 discussion of what, exactly, “equivalent” means in this case.) This is the interpretation correctness condition.

The other correctness condition is significantly more subtle, and corresponds to the property that the rewriter must map related PHOAS expressions to related PHOAS expressions. This one is a monster. We present the code before explaining it to show just how much of a mouthful it is.

```
Lemma wf_eval_decision_tree {T1 T2} G d
  :  $\forall (P : \text{option } T1 \rightarrow \text{option } T2 \rightarrow \text{Prop})$ 
    (HPNone : P None None)
```

```

(ctx1 : list (@rawexpr var1))
(ctx2 : list (@rawexpr var2))
(ctxe : list { t : type & @expr var1 t * @expr var2 t }%type)
(Hctx1 : length ctx1 = length cte)
(Hctx2 : length ctx2 = length cte)
(Hwf : ∀ t re1 e1 re2 e2,
  List.In ((re1, re2), existT _ t (e1, e2))
  (List.combine (List.combine ctx1 ctx2) cte)
  → @wf_rawexpr G t re1 e1 re2 e2)
cont1 cont2
(Hcont : ∀ n ls1 ls2,
  length ls1 = length cte
  → length ls2 = length cte
  → (forall t re1 e1 re2 e2,
    List.In ((re1, re2), existT _ t (e1, e2))
    (List.combine (List.combine ls1 ls2) cte)
    → @wf_rawexpr G t re1 e1 re2 e2)
  → (cont1 n ls1 = None ↔ cont2 n ls2 = None)
  ∧ P (cont1 n ls1) (cont2 n ls2)),
P (@eval_decision_tree var1 T1 ctx1 d cont1)
  (@eval_decision_tree var2 T2 ctx2 d cont2).

```

This is one particular way to express the following meaning: Suppose that we have two calls to `eval_decision_tree` with different PHOAS `var` types, different return types `T1` and `T2`, different continuations `cont1` and `cont2`, different expression lists `ctx1` and `ctx2`, and the same decision tree. Suppose further that we have two lists of PHOAS expressions, and a relation relating elements of `T1` to elements of `T2`. Let us assume the following properties of the expression lists and the continuations: The two lists of untyped `rawexprs` match with each other and the two lists of typed expressions, and all of the types line up. The two continuations, when fed identical indices, and fed lists of `rawexprs` which match with the given lists of typed expressions, either both fail, or both succeed with related outputs. Then we can conclude that the calls to `eval_decision_tree` either both fail, or both succeed with related outputs. Note, importantly, that we connect the lists of `rawexprs` fed to the continuations with the lists `rawexprs` fed to `eval_decision_tree` only via the lists of typed expressions.

Why do we need such complication here? The `eval_decision_tree` makes no guarantee about how much of the expression it reveals, but we must capture the fact that related PHOAS inputs result in the *same* amount of revealing, however much revealing that is. We do, however, also guarantee that the revealed expressions are both related to each other as well as to the original expressions, modulo the amount of revealing. Finally, the continuations that we use assume that enough structure is revealed, and hence are not guaranteed to do the same thing regardless of the level of revealing.

There are a couple of ways that this correctness condition might be simplified, all of which essentially amount to better enforcement of abstraction barriers.

The function that rewrites with a particular rule relies on the invariant that `eval_decision_tree` reveals enough structure. This breaks the abstraction barrier that rewriting with a particular rule is only supposed to care about the expression structure. If we enforced this abstraction barrier, we'd no longer need to talk about whether or not two `rawexprs` had the same level of revealed structure, which would vastly simplify the definition `wf_rawexpr` (discussed more in the upcoming Subsection 6.7.2). Furthermore, we could potentially remove the lists of typed expressions, mandating only that the lists of `rawexprs` be related to each other.

Finally, we could split apart the behavior of the continuation from the behavior of `eval_decision_tree`. Since the behavior of the continuations could be assumed to not depend on the amount of revealed structure, we could prove that invoking `eval_decision_tree` on any such “good” continuation returned a result *equal* to invoking the continuation on the same list of `rawexprs`, rather than merely one equivalent to it modulo the amount of revealing. This would bypass the need for this lemma entirely, allowing us to merely strengthen the previous lemma used for interpretation-correctness.

So here we see that a minor leak in an abstraction barrier (allowing the behavior of rewriting to depend on how much structure has been revealed) can vastly complicate correctness proofs, even forcing us to break other abstraction barriers by inlining the behavior of various monads.

6.6 Rewriting Again in the Output of a Rewrite Rule

We now come to the feature of the rewriter that caused the most pain: allowing some rules to be designated as subject to a second bottomup rewriting pass in their output. This feature is important for allowing users to express one operation (for example, `List.flat_map`) in terms of other operations (for example, `list_rect`) which are themselves subject to reduction.

The technical challenge, here, is that the PHOAS `var` type of the input of normalization by evaluation is not the same as the `var` type of the output. Hence the rewrite-rule replacement phase of rules marked for subsequent rewriting passes must change the `var` type when they do replacement. This can be done, roughly, by wrapping arguments passed in to the replacement rule in an extra layer of `Var` nodes.

However, this incurs severe cost in phrasing and proving the correctness condition of the rewriter. [TODO: should I try to talk about more of this pain?] While most

of the nitty-gritty details are beyond the scope even of this chapter, we will look at one particular implication of supporting this feature in Subsection 6.7.2 (Which Equivalence Relation?).

6.7 Delayed Rewriting in Variable Nodes

We saw in Subsection 6.2.3 that the `rawexpr` inductive has separate constructors for PHOAS expressions and for NbE_t values. The reason for this distinction lies at the heart of fusing normalization by evaluation and pattern matching compilation.

Consider rewriting in the expression `List.map (λ x. y + x) [0; 1]` with the rules $x + 0 = x$, and `List.map f [x ; ... ; y] = [f x ; ... ; f y]`. We want to get out the list `[y; y + 1]` and *not* `[y + 0; y + 1]`. In the bottomup approach, we first perform rewriting on the arguments to `List.map` before applying rewriting to `List.map` itself. Although it would seem that no rewrite rule applies to either argument, in fact what happens is that $(\lambda x. y + x)$ becomes an NbE_t thunk which is waiting for the structure of `x` before deciding whether or not rewriting applies. Hence when doing decision tree evaluation, it's important to keep this thunk waiting, rather than forcing it early with a generic variable node. The `rValue` constructor allows us to do this. The `rExpr` constructor, by contrast, holds expressions which we are allowed to do further matching on.

How does the use of these different constructors show up? Recall from Figure 5-2 in Subsection 5.3.2 that we put constants into η -long application form by calling `reflect` at the base case of `reduce(c)`. When performing this η -expansion, we build up a `rawexpr`. When we encounter an argument with an arrow type, we drop it directly into an `rValue` constructor, marking it as not subject to structure revealing. When we encounter an argument whose type is not an arrow, we can guarantee that there is no thunked rewriting, and so we can put the value into an `rExpr` constructor, marking it as subject to structure decomposition.

One might ask: since we distinguish the creation of `rExpr` and `rValue` on the basis of their type, could we not just use the same constructor for both? The reason we cannot do this is that when revealing structure, we may decompose an expression in an `rExpr` node into an application of an expression to another expression. In this case, the first of these will have an arrow type, and both must be placed into the `rExpr` constructor and be marked as subject to further decomposition. Hence we cannot distinguish these cases just on the basis of the type, and we do in fact need two constructors.

6.7.1 Relating Expressions and Values

First, some background context: When writing PHOAS compiler passes, there are in general two correctness conditions that must be proven about them. The first is a

soundness theorem. In Section 4.1.3, we called this theorem `check_is_even_expr_sound`. For compiler passes that produce syntax trees, this theorem will relate the denotation of the input AST to the denotation of the output AST, and might hence alternatively be called a *semantics preservation* theorem, or an *interpretation correctness* theorem. The second theorem, only applicable to compiler passes that produce ASTs (unlike our evenness checker from Subsection 7.1.2), is a syntactic well-formedness theorem. It will say that if the input AST is well-formed, then the output AST will also be well-formed. As seen in Section 4.1.3, the definition of well-formed for PHOAS relates two expressions with different `var` arguments. Hence most PHOAS well-formedness theorems are proven by showing that a given compiler pass preserves relatedness between PHOASTs with different `var` arguments.

The fact that NbE values contain thunked rewriting creates a great deal of subtlety in relating `rawexprs`. As the only correctness conditions on the rewriter are that it preserves denotational semantics of expressions and that it maps related expressions to related expressions, these are the only facts that hold about the NbE_t values in `rValue`. Since native PHOAS expressions do not permit such thunked values, we can only relate NbE_t values to the interpretation of such expressions. Even this is not straightforward, as we must use an extensional equivalence relation, saying that an NbE_t value of arrow type is equivalent to an interpreted function only when equivalence between the NbE_t value argument and the interpreted function argument implies equivalence of their outputs.

6.7.2 Which Equivalence Relation?

Generalizing the challenge Subsection 6.7.1, it turns out that describing how to relate two (or more!) objects was one of the most challenging parts of the proof effort. All told, we needed approximately *two dozen* ways of relating various objects.

We begin with the equivalence relations hinted at in previous sections.

wf_rawexpr In Section 6.5, we introduced without definition the four-place `wf_rawexpr` relation. This relation, a beefed up version of the PHOAS definition of `related` in Section 4.1.3, takes in two `rawexprs`, two PHOAS expressions (of the same type), and is parameterized over a list of pairs of allowed and related variables, much like the definition of `related`. It requires that both `rawexprs` have the same amount of revealed structure (important only because we broke the abstraction barrier of revealed structure only mattering as an optimization); that the unrevealed structure, the “alternate” expression of the `rApp` and `rIdent` nodes, match exactly with the given expressions; and that the structure that is revealed matches as well with the given expressions. The only nontrivial case in this definition is what to say about NbE_t values match expressions. We say that an NbE_t value is equivalent only to the result of calling NbE’s `reify` function on that value. That this definition suffices is highly non-obvious; we refer the reader to our Coq proofs, performed with no axioms other than functional extensionality, as our justification of sufficiency. That each NbE_t

value must match at least the result of calling NbE’s `reify` function on that value is a result of how we handle unrevealed forms when building up the arguments to an η -long identifier application as discussed briefly in Subsection 6.1.1 (What does this reduction consist of?). Namely, when forming applications of `rawexprs` to NbE_t values during η -expansion, we say that the “unrevealed” structure of an NbE_t value v is `reify v`.

interp_maybe_do_again In Section 6.6, we discussed a small subset of the implications of supporting rewriting again in the output of a rewrite rule. The most easily describable pain caused by this feature shows up in the definition of what it means for a rewrite rule to preserve denotational semantics. At the user-level, this is quite obvious: the left-hand side of the rewrite rule (prior to reification⁵) must equal the right-hand side. However, there are two subtleties to expressing the correctness condition to intermediate representations of the rewrite rule. We will discuss one of them here, and the other in Section 6.8 (What’s the Ground Truth: Patterns Or Expressions?).

At some point in the rewriting process, the rewrite rule must be expressed in terms of a PHOAS expression whose `var` type is either the output `var` type—if this rule is not subject to more rewriting—or else is the NbE_t value type—if the rule is subject to more rewriting. Hence we must be able to relate an object of this type to the denotational interpretation that we are hoping to preserve. There are two subtleties here. The first is that we cannot simply “interpret” the NbE_t values stored in `Var` nodes; we must use the extensional relation described above in Section 6.7 (Delayed Rewriting in Variable Nodes), saying that an NbE_t value of arrow type is equivalent to an interpreted function only when equivalence between the NbE_t value argument and the interpreted function argument implies equivalence of their outputs.

Second, we cannot simply interpret the expression which surrounds the `Var` node, and must instead ensure that the “interpretation” of λ s in the AST is extensional over all appropriately-related NbE_t values they might be passed. Note that it’s not even obvious how to materialize the function they must be extensionally related to. When trying to prove that the application of $(\lambda f x. v_1(f x))$ to NbE_t values v_2 and v_3 is appropriately related to the interpreted integer 5, how do we materialize the interpreted functions equivalent to $(\lambda f x. v_1(f x))$ and v_2 ? The answer is “not very well”, as we were unable to materialize them in a sufficiently constructive manner as to eliminate all uses of the axiom of function extensionality, despite sinking many hours into our attempt to eliminate this axiom.⁶

⁵Note that this reification is a tactic procedure reifying Gallina to PHOAS, *not* the `reify` function of normalization by evaluation dicussed elsewhere in this chapter.

⁶Our current thoughts are that it might be possible to prove that an interpreted function being related to any expression implies that the function respects function extensionality. We invite any brave and masochistic readers to take a stab at eliminating this axiom for us.

Related Miscellanea While delving into the details of all two-dozen ways of relating objects is beyond the scope of this thesis, we mention a couple of other non-obvious design questions that we found challenging to answer.

Recall from Subsection 5.3.2 that NbE_t values are Gallina functions on arrow types; dropping the subtleties of the `UnderLets` monad, we had

$$\begin{aligned}\text{NbE}_t(t_1 \rightarrow t_2) &:= \text{NbE}_t(t_1) \rightarrow \text{NbE}_t(t_2) \\ \text{NbE}_t(b) &:= \text{expr}(b)\end{aligned}$$

The PHOAS relatedness condition of Section 4.1.3 (PHOAS) is parameterized over a list of pairs of permitted related variables. Design Question: What is the relation between the permitted related variables lists of the terms of types $\text{NbE}_t(t_1)$, $\text{NbE}_t(t_2)$, and $\text{NbE}_t(t_1 \rightarrow t_2)$. Spoiler: The list for $\text{NbE}_t(t_1)$ is unconstrained, and is prepended to the list for $\text{NbE}_t(t_1 \rightarrow t_2)$ (which is given) to get the list for $\text{NbE}_t(t_2)$. That is, we write

$$\begin{aligned}\text{related-}\text{NbE}_{t_1 \rightarrow t_2}(\Gamma, f_1, f_2) &:= \forall \Gamma' v_1 v_2, \text{related-}\text{NbE}_{t_1}(\Gamma', v_1, v_2) \\ &\quad \rightarrow \text{related-}\text{NbE}_{t_2}(\Gamma' ++ \Gamma, f_1(v_1), f_2(v_2)) \\ \text{related-}\text{NbE}_b(\Gamma, e_1, e_2) &:= \text{related}(\Gamma, e_1, e_2)\end{aligned}$$

Some correctness lemmas do not need full-blown relatedness conditions. For example, in some places, we do not need that a `rawexpr` is fully consistent with its alternate expression structure, only that the types match and that the top-level structure of each alternate PHOAS expression matches the node of the `rawexpr`. Design Question: Is it better to minimize the number of relations and fold these “self-matching” or “goodness” properties into the definitions of relatedness, which are then used everywhere; or is it better to have separate definitions for goodness and relatedness, and have correctness conditions which more tightly pin down the behavior of the corresponding functions. (Non-Spoiler: We don’t have an answer to this one.)

[**TODO:** should I talk about anything else here?]

6.8 What’s the Ground Truth: Patterns Or Expressions?

We mentioned in Subsection 6.7.2 (Which Equivalence Relation?) that there were two subtleties to expressing the interpretation correctness condition for intermediate representations of rewrite rules, and proceeded to discuss only one of them. We discuss the other one here.

We must answer the question, in proving our rewriter correct: What denotational

semantics do we use for a rewrite rule?

In our current framework, we talk about rewrite rules in terms of patterns, which are special ASTs which contain extra pattern variables in both the types and the terms, and in terms of a replacement function, which takes in unification data and returns either failure or else a PHOAST with the data plugged in. While this design is sort-of a historical accident of originally intending to write rewrite rules by hand, there is also a genuine question of how to relate patterns to replacement functions. While we could, in theory, in a better designed rewriter, indirect through the expressions that each of these came from, the functions turning expressions into patterns and replacement rules are likely to be quite complicated, especially with the support for rewriting again described in Section 6.6 (Rewriting Again in the Output of a Rewrite Rule).

The way we currently relate these is that we write an interpretation function for patterns, parameterized over unification data, and relate this to the interpretation of the replacement function applied to unification data, suitably restricted to just the type variables of the pattern in question to make various dependent types line up. Note that this restriction of the unification data would likely be unnecessary if we stripped out all of the dependent types that we don't actually need; c.f. Subsection 6.1.3 (Type Codes). This interpretation function is itself also severely complicated by the use of dependent types in talking about unification data.

6.9 What's The Takeaway?

This chapter has been a brief survey of the engineering challenges we encountered in designing and implementing a framework for building verified partial evaluators with rewriting. We hope that this deep-dive into the details of our framework has fleshed out some of the design principles and challenges we've discussed in previous sections.

If the reader wishes to take only one thing from this chapter, we invite it to be a sense and understanding of just how important good abstraction barriers and API design are to engineering at scale in verified and dependently-typed settings.

Chapter 7

Reification by Parametricity

Fast Setup for Proof by Reflection, in Two Lines of \mathcal{L}_{tac}

[**TODO:** include the trick from <https://github.com/coq/coq/issues/5996#issuecomment-670955273>]

Abstract

We present a new strategy for performing reification in Coq. That is, we show how to generate first-class abstract syntax trees from “native” terms of Coq’s logic, suitable as inputs to verified compilers or procedures in the *proof-by-reflection* style. Our new strategy, based on simple generalization of subterms as variables, is straightforward, short, and fast. In its pure form, it is only complete for constants and function applications, but “let” binders, eliminators, lambdas, and quantifiers can be accommodated through lightweight coding conventions or preprocessing.

We survey the existing methods of reification across multiple Coq metaprogramming facilities, describing various design choices and tricks that can be used to speed them up, as well as various limitations. We report benchmarking results for 18 variants, in addition to our own, finding that our own reification outperforms 16 of these methods in all cases, and one additional method in some cases; writing an OCaml plugin is the only method tested to be faster. Our method is the most concise of the strategies we considered, reifying terms using only two to four lines of \mathcal{L}_{tac} —beyond lists of the identifiers to reify and their reified variants. Additionally, our strategy automatically provides error messages that are no less helpful than Coq’s own error messages.

7.1 Introduction

Proof by reflection [37] is an established method for employing verified proof procedures, within larger proofs. There are a number of benefits to using verified functional programs written in the proof assistant’s logic, instead of tactic scripts. We can often prove that procedures always terminate without attempting fallacious proof steps, and perhaps we can even prove that a procedure gives logically complete answers, for instance telling us definitively whether a proposition is true or false. In contrast, tactic-based procedures may encounter runtime errors or loop forever. As a consequence, those procedures must output proof terms, justifying their decisions, and these terms can grow large, making for slower proving and requiring transmission of large proof terms to be checked slowly by others. A verified procedure need not generate a certificate for each invocation.

The starting point for proof by reflection is *reification*: translating a “native” term of the logic into an explicit abstract syntax tree. We may then feed that tree to verified procedures or any other functional programs in the logic. The benefits listed above are particularly appealing in domains where goals are very large. For instance, consider verification of large software systems, where we might want to reify thousands of lines of source code. Popular methods turn out to be surprisingly slow, often to the point where, counter-intuitively, the majority of proof-execution time is spent in reification – unless the proof engineer invests in writing a plugin directly in the proof assistant’s metalanguage (e.g., OCaml for Coq).

In this paper, we show that reification can be both simpler and faster than with standard methods. Perhaps surprisingly, we demonstrate how to reify terms almost entirely through reduction in the logic, with a small amount of tactic code for setup and no ML programming. Though our techniques should be broadly applicable, especially in proof assistants based on type theory, our experience is with Coq, and we review the requisite background in the remainder of this introduction. In Section 7.2, we summarize our survey into prior approaches to reification and provide high-quality implementations and documentation for them, serving a tutorial function independent of our new contributions. Experts on the subject might want to skip directly to Section 7.3, which explains our alternative technique. We benchmark our approach against 18 competitors in Section 7.4.

7.1.1 Proof-Script Primer

Basic Coq proofs are often written as lists of steps such as `induction` on some structure, `rewrite` using a known equivalence, or `unfold` of a definition. Very quickly, proofs can become long and tedious, both to write and to read, and hence Coq provides \mathcal{L}_{tac} , a scripting language for proofs. As theorems and proofs grow in complexity, users frequently run into performance and maintainability issues with \mathcal{L}_{tac} . Consider the case where we want to prove that a large algebraic expression, involving many `let ... in ...` expressions, is even:

```

Inductive is_even : nat -> Prop :=
| even_0 : is_even 0
| even_SS : forall x, is_even x -> is_even (S (S x)).
Goal is_even (let x := 100 * 100 * 100 * 100 in
              let y := x * x * x * x in
              y * y * y * y).

```

Coq stack-overflows if we try to reduce this goal. As a workaround, we might write a lemma that talks about evenness of `let ... in ...`, plus one about evenness of multiplication, and we might then write a tactic that composes such lemmas.

Even on smaller terms, though, proof size can quickly become an issue. If we give a naive proof that 7000 is even, the proof term will contain all of the even numbers between 0 and 7000, giving a proof-term-size blow-up at least quadratic in size (recalling that natural numbers are represented in unary; the challenges remain for more efficient base encodings). Clever readers will notice that Coq could share subterms in the proof tree, recovering a term that is linear in the size of the goal. However, such sharing would have to be preserved very carefully, to prevent size blow-up from unexpected loss of sharing, and today's Coq version does not do that sharing. Even if it did, tactics that rely on assumptions about Coq's sharing strategy become harder to debug, rather than easier.

7.1.2 Reflective-Automation Primer

Enter reflective automation, which simultaneously solves both the problem of performance and the problem of debuggability. Proof terms, in a sense, are traces of a proof script. They provide Coq's kernel with a term that it can check to verify that no illegal steps were taken. Listing every step results in large traces.

The idea of reflective automation is that, if we can get a formal encoding of our goal, plus an algorithm to *check* the property we care about, then we can do much better than storing the entire trace of the program. We can prove that our checker is correct once and for all, removing the need to trace its steps.

A simple evenness checker can just operate on the unary encoding of natural numbers (Figure 7-1). We can use its correctness theorem to prove goals much more quickly:

```

Fixpoint check_is_even
  (n : nat) : bool
:= match n with
  | 0 => true
  | 1 => false
  | S (S n)
    => check_is_even n
end.

```

Figure 7-1: Evenness Checking

```

Theorem soundness : forall n, check_is_even n = true -> is_even n.
Goal is_even 2000.
Time repeat (apply even_SS || apply even_0). (* 1.8 s *)
Undo.

```

```
Time apply soundness; vm_compute; reflexivity. (* 0.004 s *)
```

The tactic `vm_compute` tells Coq to use its virtual machine for reduction, to compute the value of `check_is_even 2000`, after which `reflexivity` proves that `true = true`. Note how much faster this method is. In fact, even the asymptotic complexity is better; this new algorithm is linear rather than quadratic in `n`.

However, even this procedure takes a bit over three minutes to prove `is_even (10 * 10 * 10 * 10 * 10 * 10 * 10 * 10 * 10)`. To do better, we need a formal representation of terms or expressions.

7.1.3 Reflective-Syntax Primer

Sometimes, to achieve faster proofs, we must be able to tell, for example, whether we got a term by multiplication or by addition, and not merely whether its normal form is 0 or a successor.

A reflective automation procedure generally has two steps. The first step is to *reify* the goal into some abstract syntactic representation, which we call the *term language* or an *expression language*. The second step is to run the algorithm on the reified syntax.

```
Inductive expr :=
| Nat0 : expr
| NatS (x : expr) : expr
| NatMul (x y : expr) : expr.
```

What should our expression language include? At a bare minimum, we must have multiplication nodes, and we must have `nat` literals. If we encode `S` and `0` separately, a decision that will become important later in Section 7.3, we get the inductive type of Figure 7-2.

Before diving into methods of reification, let us write the evenness checker.

```
Fixpoint check_is_even_expr (t : expr) : bool
:= match t with
  | Nat0 => true
  | NatS x => negb (check_is_even_expr x)
  | NatMul x y => orb (check_is_even_expr x) (check_is_even_expr y)
end.
```

Before we can state the soundness theorem (whenever this checker returns `true`, the represented number is even), we must write the function that tells us what number our expression represents, called *denotation* or *interpretation*:

```
Fixpoint denote (t : expr) : nat
:= match t with
  | Nat0 => 0
  | NatS x => S (denote x)
```

```

| NatMul x y => denote x * denote y
end.
```

```

Theorem check_is_even_expr_sound (e : expr)
  : check_is_even_expr e = true -> is_even (denote e).
```

Given a tactic `Reify` to produce a reified term from a `nat`, we can time `check_is_even_expr`. It is instant on the last example.

Before we proceed to reification, we will introduce one more complexity. If we want to support our initial example with `let ... in ...` efficiently, we must also have `let`-expressions. Our current procedure that inlines `let`-expressions takes 19 seconds, for example, on `let x0 := 10 * 10 in let x1 := x0 * x0 in ... let x24 := x23 * x23 in x24`. The choices of representation include higher-order abstract syntax (HOAS) [156], parametric higher-order abstract syntax (PHOAS) [50], and de Bruijn indices [40]. The PHOAS representation is particularly convenient. In PHOAS, expression binders are represented by binders in Gallina, the functional language of Coq, and the expression language is parameterized over the type of the binder. Let us define a constant and notation for `let` expressions as definitions (a common choice in real Coq developments, to block Coq's default behavior of inlining `let` binders silently; the same choice will also turn out to be useful for reification later). We thus have:

```

Inductive expr {var : Type} :=
| Nat0 : expr
| NatS : expr -> expr
| NatMul : expr -> expr -> expr
| Var : var -> expr
| LetIn : expr -> (var -> expr) -> expr.
Definition Let_In {A B} (v : A) (f : A -> B) := let x := v in f x.
Notation "'dlet' x := v 'in' f" := (Let_In v (fun x => f)).
Notation "'elet' x := v 'in' f" := (LetIn v (fun x => f)).
Fixpoint denote (t : @expr nat) : nat
  := match t with
    | Nat0 => 0
    | NatS x => S (denote x)
    | NatMul x y => denote x * denote y
    | Var v => v
    | LetIn v f => dlet x := denote v in denote (f x)
  end.
```

A full treatment of evenness checking for PHOAS would require proving well-formedness of syntactic expressions; for a more complete discussion of PHOAS, we refer the reader elsewhere [50]. Using `Wf` to denote the well-formedness predicate, we could prove a theorem

```
Theorem check_is_even_expr_sound (e : ∀ var, @expr var) (H : Wf e)
: check_is_even_expr (e bool) = true -> is_even (denote (e nat)).
```

To complete the picture, we would need a tactic `Reify` which took in a term of type `nat` and gave back a term of type `forall var, @expr var`, plus a tactic `prove_wf` which solved a goal of the form `Wf e` by repeated application of constructors. Given these, we could solve an evenness goal by writing¹

```
match goal with
| [ |- is_even ?v ]
  => let e := Reify v in
    refine (check_is_even_expr_sound e _ _);
    [ prove_wf | vm_compute; reflexivity ]
end.
```

7.2 Methods of Reification

We implemented reification in 18 different ways, using 6 different metaprogramming facilities in the Coq ecosystem: Ltac, Ltac2, Mtac [82], type classes [181], canonical structures [80], and reification-specific OCaml plugins (quote [58], template-coq [16], ours). Figure 7-3 displays the simplest case: an Ltac script to reify a tree of function applications and constants. Unfortunately, all methods we surveyed become drastically more complicated or slower (and usually both) when adapted to reify terms with variable bindings such as `let-in` or λ nodes.

We have made detailed walkthroughs and source code of these implementations available² in hope that they will be useful for others considering implementing reification using one of these metaprogramming mechanisms, instructive as nontrivial examples of multiple metaprogramming facilities, or helpful as a case study in Coq performance engineering. However, we do *not* recommend reading these out of general interest: most of the complexity in the described implementations strikes us as needless, with significant aspects of the design being driven by surprising behaviors, misfeatures, bugs, and performance bottlenecks of the underlying machinery as opposed to the task of reification.

```
Ltac f v x := (* reify var term *)
lazymatch x with
| O => constr:(@Nat0 v)
| S ?x => let X := f v x in
            constr:(@NatS v X)
| ?x*?y => let X := f v x in
            let Y := f v y in
            constr:(@NatMul v X Y)
end.
```

¹Note that for the `refine` to be fast, we must issue something like `Strategy -10 [denote]` to tell Coq to unfold `denote` before `Let_In`.

²<https://github.com/mit-plv/reification-by-parametricity>

7.3 Reification by Parametricity

We propose factoring reification into two passes, both of which essentially have robust, built-in implementations in Coq: *abstraction* or *generalization*, and *substitution* or *specialization*.

The key insight to this factoring is that the shape of a reified term is essentially the same as the shape of the term that we start with. We can make precise the way these shapes are the same by abstracting over the parts that are different, obtaining a function that can be specialized to give either the original term or the reified term.

That is, we have the commutative triangle in Figure 7-4.

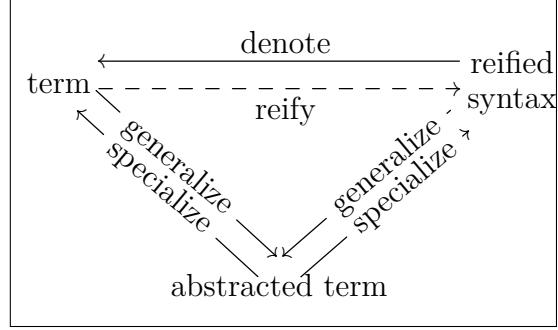


Figure 7-4: Abstraction and Reification

7.3.1 Case-By-Case Walkthrough Function Applications And Constants.

Consider the example of reifying 2×2 . In this case, the *term* is 2×2 or $(\text{mul} (\text{S} (\text{S} \text{O})) (\text{S} (\text{S} \text{O})))$.

To reify, we first *generalize* or *abstract* the term 2×2 over the successor function S , the zero constructor O , the multiplication function mul , and the type \mathbb{N} of natural numbers. We get a function taking one type argument and three value arguments:

$$\lambda N. \lambda(\text{MUL} : N \rightarrow N \rightarrow N) (\text{O} : N) (\text{S} : N \rightarrow N). \text{MUL} (\text{S} (\text{S} \text{O})) (\text{S} (\text{S} \text{O}))$$

We can now specialize this term in one of two ways: we may substitute \mathbb{N} , mul , O , and S , to get back the term we started with; or we may substitute expr , NatMul , NatO , and NatS to get the reified syntax tree

$$\text{NatMul} (\text{NatS} (\text{NatS} \text{NatO})) (\text{NatS} (\text{NatS} \text{NatO}))$$

This simple two-step process is the core of our algorithm for reification: abstract over all identifiers (and key parts of their types) and specialize to syntax-tree constructors for these identifiers.

Wrapped Primitives: “Let” Binders, Eliminators, Quantifiers.

The above procedure can be applied to a term that contains “let” binders to get a PHOAS syntax tree that represents the original term, but doing so would not capture sharing. The result would contain native “let” bindings of subexpressions, not PHOAS let expressions. Call-by-value evaluation of any procedure applied to

the reification result would first substitute the let-bound subexpressions – leading to potentially exponential blowup and, in practice, memory exhaustion.

The abstraction mechanisms in all proof assistants (that we know about) only allow abstracting over terms, not language primitives. However, primitives can often be wrapped in explicit definitions, which we *can* abstract over. For example, we already used a wrapper for “let” binders, and terms that use it can be reified by abstracting over that definition. If we start with the expression

```
dlet a := 1 in a × a
```

and abstract over (`@Let_In N N`), S, O, mul, and N, we get a function of one type argument and four value arguments:

$$\begin{aligned} \Lambda N. \lambda (\text{MUL} : N \rightarrow N \rightarrow N). \lambda(O : N). \lambda(S : N \rightarrow N). \\ \lambda(\text{LETIN} : N \rightarrow (N \rightarrow N) \rightarrow N). \text{LETIN } (S \ O) \ (\lambda a. \text{MUL } a \ a) \end{aligned}$$

We may once again specialize this term to obtain either our original term or the reified syntax. Note that to obtain reified PHOAS syntax, we must include a `Var` node in the `LetIn` expression; we substitute $(\lambda x f. \text{LetIn } x (\lambda v. f (\text{Var } v)))$ for `LETIN` to obtain the PHOAS syntax tree

```
LetIn (NatS Nat0) (\lambda v. NatMul (Var v) (Var v))
```

Wrapping a metalanguage primitive in a definition in the code to be reified is in general sufficient for reification by parametricity. Pattern matching and recursion cannot be abstracted over directly, but if the same code is expressed using eliminators, these can be handled like other functions. Similarly, even though \forall/Π cannot be abstracted over, proof automation that itself introduces universal quantifiers before reification can easily wrap them in a marker definition (`_forall T P := forall (x:T), P x`) that can be. Existential quantifiers are not primitive in Coq and can be reified directly.

Lambdas.

While it would be sufficient to require that, in code to be reified, we write all lambdas with a named wrapper function, that would significantly clutter the code. We can do better by making use of the fact that a PHOAS object-language lambda (`Abs` node) consists of a metalanguage lambda that binds a value of type `var`, which can be used in expressions through constructor `Var : var → expr`. Naive reification by parametricity would turn a lambda of type $N \rightarrow N$ into a lambda of type `expr → expr`. A reification procedure that explicitly recurses over the metalanguage syntax could just precompose this recursive-call result with `Var` to get the desired object-language encoding of the lambda, but handling lambdas specially does not fit in the framework of abstraction and specialization.

First, let us handle the common case of lambdas that appear as arguments to higher-order functions. One easy approach: while the parametricity-based framework does

not allow for special-casing lambdas, it is up to us to choose how to handle functions that we expect will take lambdas as arguments. We may replace each higher-order function with a metalanguage lambda that wraps the higher-order arguments in object-language lambdas, inserting `Var` nodes as appropriate. Code calling the function `sum_upto n f := f(0)+f(1)+⋯+f(n)` can be reified by abstracting over relevant definitions and substituting $(\lambda n f. \text{SumUpTo } n (\text{Abs} (\lambda v. f (\text{Var } v))))$ for `sum_upto`. Note that the expression plugged in for `sum_upto` differs from the one plugged in for `Let_In` only in the use of a deeply embedded abstraction node. If we wanted to reify `LetIn` as just another higher-order function (as opposed to a distinguished wrapper for a primitive), the code would look identical to that for `sum_upto`.

It would be convenient if abstracting and substituting for functions that take higher-order arguments were enough to reify lambdas, but here is a counterexample.

$$\begin{aligned} & \lambda x y. x \times ((\lambda z. z \times z) y) \\ \Lambda N. \lambda(\text{MUL} : N \rightarrow N \rightarrow N). \lambda (x y : N). \text{Mul} x ((\lambda (z : N). \text{Mul} z z) y) \\ & \lambda (x y : \text{expr}). \text{NatMul} x (\text{NatMul} y y) \end{aligned}$$

The result is not even a PHOAS expression. We claim a desirable reified form is

$$\text{Abs}(\lambda x. \text{Abs}(\lambda y. \text{NatMul} (\text{Var } x) (\text{NatMul} (\text{Var } y) (\text{Var } y))))$$

Admittedly, even our improved form is not quite precise: $\lambda z. z \times z$ has been lost. However, as almost all standard Coq tactics silently reduce applications of lambdas, working under the assumption that functions not wrapped in definitions will be arbitrarily evaluated during scripting is already the norm. Accepting that limitation, it remains to consider possible occurrences of metalanguage lambdas in normal forms of outputs of reification as described so far. As lambdas in `expr` nodes that take metalanguage functions as arguments (`LetIn`, `Abs`) are handled by the rules for these nodes, the remaining lambdas must be exactly at the head of the expression. Manipulating these is outside of the power of abstraction and specialization; we recommend postprocessing using a simple recursive tactic script.

7.3.2 Commuting Abstraction and Reduction

Sometimes, the term we want to reify is the result of reducing another term. For example, we might have a function that reduces to a term with a variable number of `let` binders.³ We might have an inductive type that counts the number of `let ... in ...` nodes we want in our output.

```
Inductive count := none | one_more (how_many : count).
```

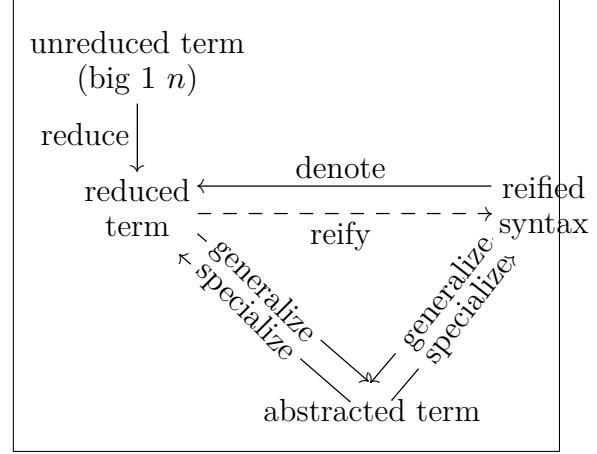
It is important that this type be syntactically distinct from \mathbb{N} for reasons we will see

³More realistically, we might have a function that represents big numbers using multiple words of a user-specified width. In this case, we may want to specialize the procedure to a couple of different bitwidths, then reifying the resulting partially reduced term.

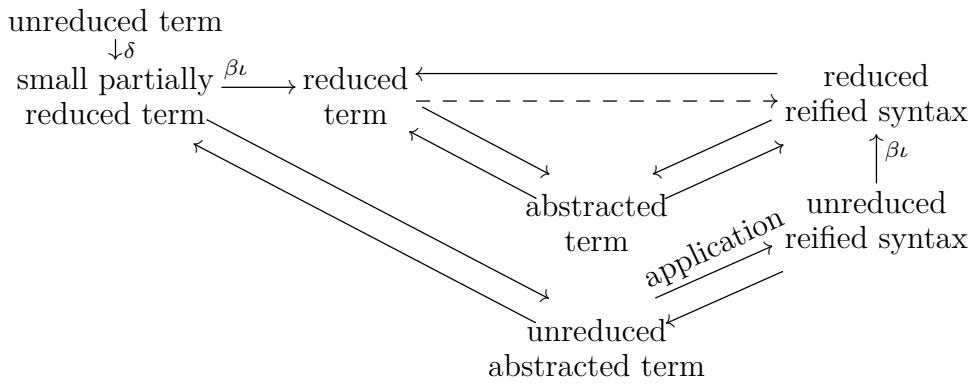
shortly.

We can then define a recursive function that constructs some number of nested let binders:

```
Fixpoint big (x:nat) (n:count)
  : nat
  := match n with
    | none => x
    | one_more n'
      => dlet x' := x * x in
        big x' n'
    end.
```



Our commutative diagram in Figure 7-4 now has an additional node, becoming Figure 7-5. Since generalization and specialization are proportional in speed to the size of the term begin handled, we can gain a significant performance boost by performing generalization before reduction. To explain why, we split apart the commutative diagram a bit more; in reduction, there is a δ or unfolding step, followed by a $\beta\iota$ step that reduces applications of λ s and evaluates recursive calls. In specialization, there is an application step, where the λ is applied to arguments, and a β -reduction step, where the arguments are substituted. To obtain reified syntax, we may perform generalization after δ -reduction (before $\beta\iota$ -reduction), and we are not required to perform the final β -reduction step of specialization to get a well-typed term. It is important that unfolding `big` results in exposing the body for generalization, which we accomplish in Coq by exposing the anonymous recursive function; in other languages, the result may be a primitive eliminator applied to the body of the fixpoint. Either way, our commutative diagram thus becomes



Let us step through this alternative path of reduction using the example of the unreduced term `big 1 100`, where we take 100 to mean the term represented by $\underbrace{(\text{one_more} \dots (\text{one_more}}_{100} \text{ none) \dots)}_{100}$.

Our first step is to unfold `big`, rendered as the arrow labeled δ in the diagram. In Coq, the result is an anonymous fixpoint; here we will write it using the recursor `count_rec` of type $\forall T. T \rightarrow (\text{count} \rightarrow T \rightarrow T) \rightarrow \text{count} \rightarrow T$. Performing δ -reduction, that is, unfolding `big`, gives us the small partially reduced term

```
(λ(x : N). λ(n : count).
  count_rec (N → N) (λx. x) (λn'. λbign'. λx. dlet x' := x × x in bign' x')) 1 100
```

We call this term small, because performing $\beta\iota$ reduction gives us a much larger reduced term:

```
dlet x1 := 1 × 1 in ... dlet x100 := x99 × x99 in x100
```

Abstracting the small partially reduced term over `@Let_In N N`, S, O, mul, and N gives us the abstracted unreduced term

```
ΛN. λ(MUL : N → N → N)(O : N)(S : N → N)(LETIN : N → (N → N) → N).
  (λ(x : N). λ(n : count). count_rec (N → N) (λx. x)
    (λn'. λbign'. λx. LETIN (MUL x x) (λx'. bign' x'))))
  (S O) 100
```

Note that it is essential here that `count` is not syntactically the same as N; if they were the same, the abstraction would be ill-typed, as we have not abstracted over `count_rec`. More generally, it is essential that there is a clear separation between types that we reify and types that we do not, and we must reify *all* operations on the types that we reify.

We can now apply this term to `expr`, `NatMul`, `NatS`, `Nat0`, and, finally, $(\lambda v f. \text{LetIn } v (\lambda x. f(\text{Var } x)))$. We get an unreduced reified syntax tree of type `expr`. If we now perform $\beta\iota$ reduction, we get our fully reduced reified term.

We take a moment to emphasize that this technique is not possible with any other method of reification. We could just as well have not specialized the function to the `count` of 100, yielding a function of type `count → expr`, despite the fact that our reflective language knows nothing about `count`!

This technique is especially useful for terms that will not reduce without concrete parameters, but which should be reified for many different parameters. Running reduction once is slightly faster than running OCaml reification once, and it is more than twice as fast as running reduction followed by OCaml reification. For sufficiently large terms and sufficiently many parameter values, this performance beats even OCaml reification.⁴

⁴We discovered this method in the process of needing to reify implementations of cryptographic primitives [70] for a couple hundred different choices of numeric parameters (e.g., prime modulus of

7.3.3 Implementation in \mathcal{L}_{tac}

`ExampleMoreParametricity.v` in the code supplement mirrors the development of reification by parametricity in Subsection 7.3.1.

Unfortunately, Coq does not have a tactic that performs abstraction.⁵ However, the `pattern` tactic suffices; it performs abstraction followed by application, making it a sort of one-sided inverse to β -reduction. By chaining `pattern` with an \mathcal{L}_{tac} -match statement to peel off the application, we can get the abstracted function.

```
Ltac Reify x :=
match(eval pattern nat, Nat.mul, S, 0, (@Let_In nat nat) in x)with
| ?rx _ _ _ _ =>
  constr:( fun var => rx (@expr var) NatMul NatS Nat0
                           (fun v f => LetIn v (fun x => f (Var x))) )
end.
```

Note that if `@expr var` lives in `Type` rather than `Set`, an additional step involving retyping the term is needed; we refer the reader to `Parametricity.v` in the code supplement.

The error messages returned by the `pattern` tactic can be rather opaque at times; in `ExampleParametricityErrorMessages.v`, we provide a procedure for decoding the error messages.

Open Terms.

At some level it is natural to ask about generalizing our method to reify open terms (i.e., with free variables), but we think such phrasing is a red herring. Any lemma statement about a procedure that acts on a representation of open terms would need to talk about how these terms would be closed. For example, solvers for algebraic goals without quantifiers treat free variables as implicitly universally quantified. The encodings are invariably ad-hoc: the free variables might be assigned unique numbers during reification, and the lemma statement would be quantified over a sufficiently long list that these numbers will be used to index into. Instead, we recommend directly reifying the natural encoding of the goal as interpreted by the solver, e.g. by adding new explicit quantifiers. Here is a hypothetical goal and a tactic script for this strategy:

```
(a b : nat) (H : 0 < b) |- ∃ q r, a = q × b + r ∧ r < b

repeat match goal with
| n : nat |- ?P =>
  match eval pattern n in P with
```

arithmetic). A couple hundred is enough to beat the overhead.

⁵The `generalize` tactic returns \forall rather than λ , and it only works on types.

```

| ?P' _ => revert n; change (_forall nat P')
end
| H : ?A |- ?B => revert H; change (impl A B)
| |- ?G => (* ∀ a b, 0 < b → ∃ q r, a = q × b + r ∧ r < b *)
let rG := Reify G in
refine (nonlinear_integer_solver_sound rG _ _);
[ prove_wf | vm_compute; reflexivity ]
end.

```

Briefly, this script replaced the context variables `a` and `b` with universal quantifiers in the conclusion, and it replaced the premise `H` with an implication in the conclusion. The syntax-tree datatype used in this example can be found in `ExampleMoreParametricity.v`.

7.3.4 Advantages and Disadvantages

This method is faster than all but $\mathcal{L}_{tac}2$ and OCaml reification, and commuting reduction and abstraction makes this method faster even than the low-level $\mathcal{L}_{tac}2$ reification in many cases. Additionally, this method is much more concise than nearly every other method we have examined, and it is very simple to implement.

We will emphasize here that this strategy shines when the initial term is small, the partially computed terms are big (and there are many of them), and the operations to evaluate are mostly well-separated by types (e.g., evaluate all of the `count` operations and none of the `nat` ones).

This strategy is not directly applicable for reification of `match` (rather than eliminators) or `let ... in ...` (rather than a definition that unfolds to `let ... in ...`), `forall` (rather than a definition that unfolds to `forall`), or when reification should not be modulo $\beta\iota\zeta$ -reduction.

7.4 Performance Comparison

We have done a performance comparison of the various methods of reification to the PHOAS language `@expr var` from Figure 7.1.3 in Coq 8.7.1. A typical reification routine will obtain the term to be reified from the goal, reify it, run `transitivity` (`denote reified_term`) (possibly after normalizing the reified term), and solve the side condition with something like `lazy [denote]; reflexivity`. Our testing on a few samples indicated that using `change` rather than `transitivity; lazy [denote]; reflexivity` can be around 3X slower; note that we do not test the time of `Defined`.

There are two interesting metrics to consider: (1) how long does it take to reify the term? and (2) how long does it take to get a normalized reified term, i.e., how long does it take both to reify the term and normalize the reified term? We have chosen to

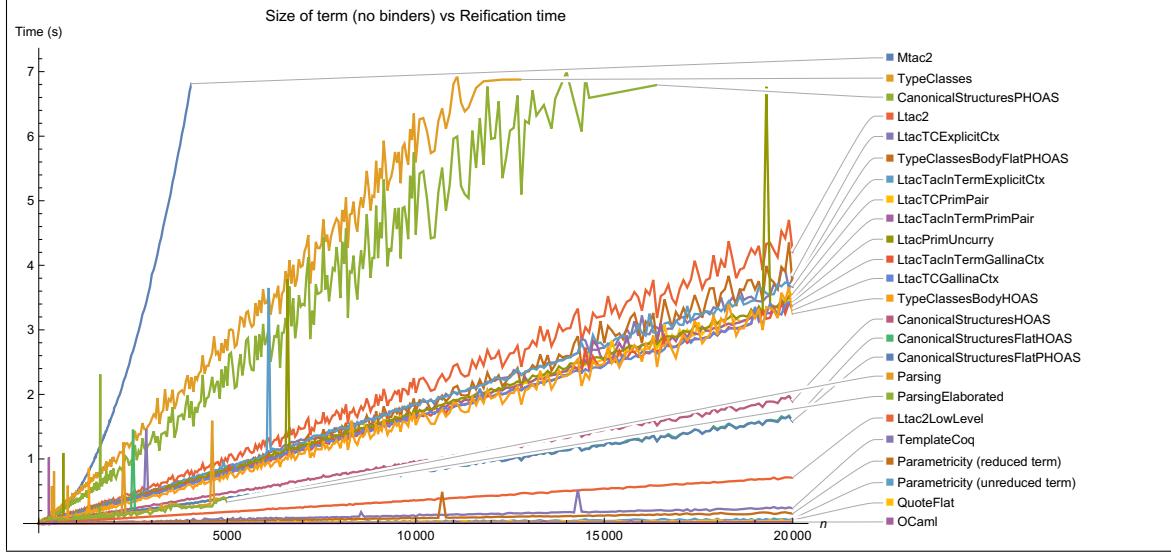


Figure 7-6: Performance of Reification without Binders

consider (1), because it provides the most fine-grained analysis of the actual reification method.

7.4.1 Without Binders

We look at terms of the form $1 * 1 * 1 * \dots$ where multiplication is associated to create a balanced binary tree. We say that the *size of the term* is the number of 1s. We refer the reader to the attached code for the exact test cases and the code of each reification method being tested.

We found that the performance of all methods is linear in term size.

Sorted from slowest to fastest, most of the labels in Figure 7-6 should be self-explanatory and are found in similarly named .v files in the associated code; we call out a few potentially confusing ones:

- The “Parsing” benchmark is “reification by copy-paste”: a script generates a .v file with notation for an already-reified term; we benchmark the amount of time it takes to parse and typecheck that term. The “ParsingElaborated” benchmark is similar, but instead of giving notation for an already-reified term, we give the complete syntax tree, including arguments normally left implicit. Note that these benchmarks cut off at around 5000 rather than at around 20 000, because on large terms, Coq crashes with a stack overflow in parsing.
- We have four variants starting with “CanonicalStructures” here. The Flat variants reify to `@expr nat` rather than to `forall var, @expr var` and benefit from fewer function binders and application nodes. The HOAS variants do

not include a case for `let ... in ...` nodes, while the PHOAS variants do. Unlike most other reification methods, there is a significant cost associated with handling more sorts of identifiers in canonical structures.

We note that on this benchmark our method is slightly faster than template-coq, which reifies to de Bruijn indices, and slightly slower than the quote plugin in the standard library and the OCaml plugin we wrote by hand.

7.4.2 With Binders

We look at terms of the form `dlet a1 := 1 * 1 in dlet a2 := a1 * a1 in ... dlet an := an-1 * an-1 in an`, where n is the size of the term. The first graph shown here includes all of the reification variants at linear scale, while the next step zooms in on the highest-performance variants at log-log scale.

In addition to reification benchmarks, the graph in Figure 7-7 includes as a reference (1) the time it takes to run `lazy` reduction on a reified term already in normal form (“identity lazy”) and (2) the time it takes to check that the reified term matches the original native term (“lazy Denote”). The former is just barely faster than OCaml reification; the latter often takes longer than reification itself. The line for the template-coq plugin cuts off at around 10 000 rather than around 20 000 because at that point template-coq starts crashing with stack overflows.

A nontrivial portion of the cost of “Parametricity (reduced term)” seems to be due to the fact that looking up the type of a binder is linear in the number of binders in the context, thus resulting in quadratic behavior of the retyping step that comes after abstraction in the `pattern` tactic. In Coq 8.8, this lookup will be $\log n$, and so reification will become even faster [151].

7.5 Future Work, Concluding Remarks

We identify one remaining open question with this method that has the potential of removing the next largest bottleneck in reification: using reduction to show that the reified term is correct.

Recall our reification procedure and the associated diagram, from Figure 7.3.2. We perform δ on an unreduced term to obtain a small, partially reduced term; we then perform abstraction to get an abstracted, unreduced term, followed by application to get unreduced reified syntax. These steps are all fast. Finally, we perform

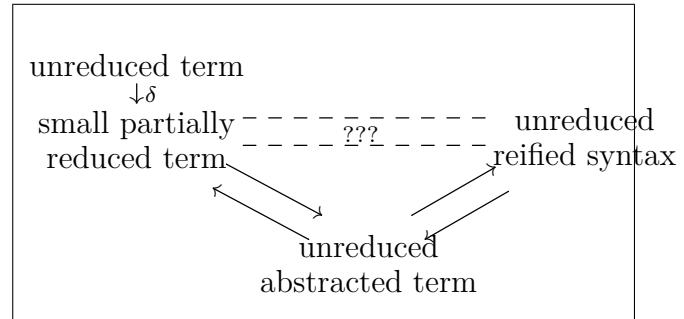


Figure 7-8: Completing the commutative triangle

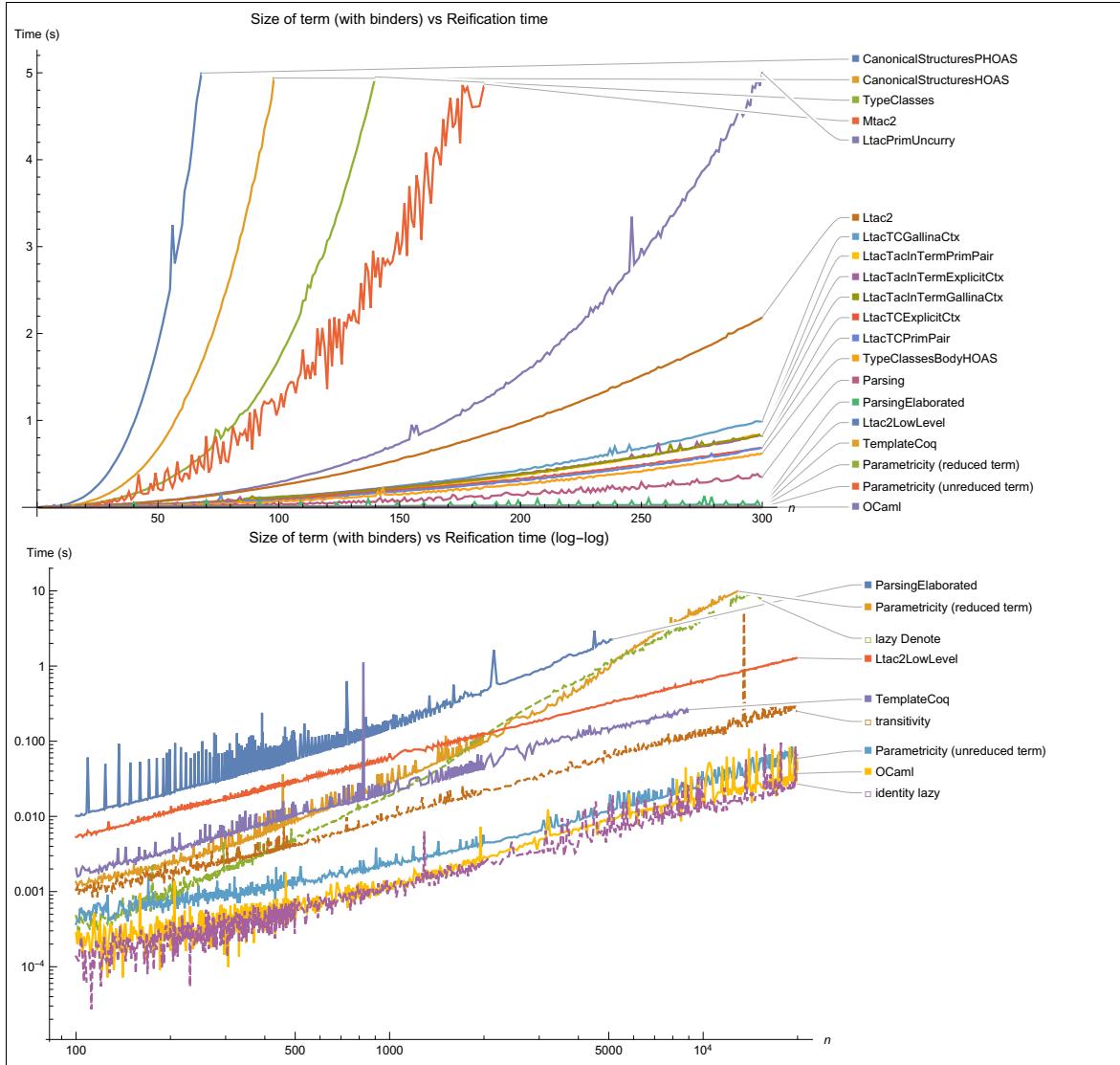


Figure 7-7: Performance of Reification with Binders

$\beta\iota$ -reduction to get reduced, reified syntax and perform $\beta\iota\delta$ reduction to get back a reduced form of our original term. These steps are slow, but we must do them if we are to have verified reflective automation.

It would be nice if we could prove this equality without ever reducing our term. That is, it would be nice if we could have the diagram in Figure 7-8.

The question, then, is how to connect the small partially reduced term with `denote` applied to the unreduced reified syntax. That is, letting F denote the unreduced abstracted term, how can we prove, without reducing F , that

$$F \mathrel{\mathbb{N}} \text{Mul} \mathrel{\text{O}} \text{S} (@\text{Let_In} \mathrel{\mathbb{N}} \mathrel{\mathbb{N}}) = \text{denote } (F \mathrel{\text{expr}} \text{NatMul} \mathrel{\text{Nat0}} \text{NatS} \text{ LetIn})$$

We hypothesize that a form of internalized parametricity would suffice for proving this lemma. In particular, we could specialize F 's type argument with $\mathbb{N} \times \text{expr}$. Then we would need a proof that for any function F of type

$$\forall(T : \text{Type}), (T \rightarrow T \rightarrow T) \rightarrow T \rightarrow (T \rightarrow T) \rightarrow (T \rightarrow (T \rightarrow T) \rightarrow T) \rightarrow T$$

and any types A and B , and any terms $f_A : A \rightarrow A \rightarrow A$, $f_B : B \rightarrow B \rightarrow B$, $a : A$, $b : B$, $g_A : A \rightarrow A$, $g_B : B \rightarrow B$, $h_A : A \rightarrow (A \rightarrow A) \rightarrow A$, and $h_B : B \rightarrow (B \rightarrow B) \rightarrow B$, using $f \times g$ to denote lifting a pair of functions to a function over pairs:

$$\begin{aligned}\text{fst } (F(A \times B)(f_A \times f_B)(a, b)(g_A \times g_B)(h_A \times h_B)) &= FA f_A a g_A h_A \wedge \\ \text{snd } (F(A \times B)(f_A \times f_B)(a, b)(g_A \times g_B)(h_A \times h_B)) &= FB f_B b g_B h_B\end{aligned}$$

This theorem is a sort of parametricity theorem.

Despite this remaining open question, we hope that our performance results make a strong case for our method of reification; it is fast, concise, and robust.

7.6 Acknowledgments and Historical Notes

We would like to thank Hugo Herbelin for sharing the trick with `type of` to propagate universe constraints⁶ as well as useful conversations on Coq's bug tracker that allowed us to track down performance issues.⁷ We would like to thank Pierre-Marie Pédrot for conversations on Coq's Gitter and his help in tracking down performance bottlenecks in earlier versions of our reification scripts and in Coq's tactics. We would like to thank Beta Ziliani for his help in using Mtac2, as well as his invaluable guidance in figuring out how to use canonical structures to reify to PHOAS. We also thank John Wiegley for feedback on the paper.

For those interested in history, our method of reification by parametricity was inspired by the `evm_compute` tactic [129]. We first made use of `pattern` to allow `vm_compute` to replace `cbv-with-an-explicit-blacklist` when we discovered `cbv` was too slow and the blacklist too hard to maintain. We then noticed that in the sequence of doing abstraction; `vm_compute`; application; β -reduction; reification, we could move β -reduction to the end of the sequence if we fused reification with application, and thus reification by parametricity was born.

This work was supported in part by a Google Research Award and National Science Foundation grants CCF-1253229, CCF-1512611, and CCF-1521584.

⁶<https://github.com/coq/coq/issues/5996#issuecomment-338405694>

⁷<https://github.com/coq/coq/issues/6252>

Part IV

Conclusion

Chapter 8

A Retrospective on Performance Improvements

Throughout this thesis, we've looked at the problem of performance in proof assistants, especially those based on dependent type theory, with Coq as our primary tool under investigation. Part I aimed to convince the reader that this problem is interesting, important, challenging, and understudied, as it differs in non-trivial ways from performance bottlenecks in non-dependently-typed languages. Part II discussed design principles to avoid performance pitfalls, and Part III took a deep dive into a particular set of performance bottlenecks and presented a tool, and, we hope, exposed the underlying design methodology, that allows eliminating asymptotic bottlenecks in one important part of proof assistant systems.

In this chapter, we will look instead at the successes of the past decade¹, ways in which performance has improved in major ways. Section 8.1 will discuss specific improvements in the implementation of Coq which resulted in performance gains, paying special attention to the underlying bottleneck being addressed. Those without special interest in the low-level details of proof assistant implementation may want to skip to Section 8.2, which will discuss changes to the underlying type theory of Coq which make possible drastic performance improvements. While we will again have our eye on Coq in Section 8.2, we will broaden our perspective in Section 8.3 to discuss new discoveries of the past decade or so in dependent type theory which enable performance improvements but have not yet made their way into Coq.

¹Actually, the time span we're considering is the course of the author's experience with Coq, which is a bit less than a decade.

8.1 Concrete Performance Advancements in Coq

In this section, we dive into the minutiae: concrete changes to Coq that have measurably increased performance.

8.1.1 Removing Pervasive Evar Normalization

Back when I started using Coq, in version 8.4, almost every single tactic was at least linear in performance in the size of the goal. This included tactics like “add new hypothesis to the context of type `True`” (`pose proof I`) and tactics like “give me the type of the most recently added hypothesis” (`match goal with H : ?T |- _ => T end`). The reason for this was pervasive *evar normalization*.

Let us review some details of the way Coq handles proof scripts. [TODO: if this is described earlier, refer back to it] Coq a partial proof term, where not-yet-given subterms are *existential variables*, or evars, which may show up as goals. For example, when proving the goal `True ∧ True`, after running `split`, the proof term would be `conj ?Goal1 ?Goal2`, where `?Goal1` and `?Goal2` are evars. There are two subtleties:

1. Evars may be under binders. Coq uses a locally nameless representation of terms (c.f. Section 4.1.3), where closed terms use de Bruijn indices, but open terms, i.e., evars, refer to the variables in their context by name. Hence each evar carries with it a named context, which causes a great deal of trouble as described in Section 2.2.3 (Quadratic Creation of Substitutions for Existential Variables).
2. Coq supports backtracking, so we must remember the history of partial proof terms. In particular, we cannot simply mutate partial proof terms to instantiate the evars, and copying the entire partial proof term just to update a small part of it would also incur a great deal of overhead. Instead, Coq never mutates the terms, and instead simply keeps a map of which evars have been instantiated with which terms, called the *evar map*.

There is an issue with the straightforward implementation of evars and evar maps. When walking terms, care must be taken with the evar case, to check whether or not the evar has in fact been instantiated or not. Subtle bugs in unification and other areas of Coq resulted from some functions being incorrectly sensitive to whether or not a term had been built via evar instantiation or given directly.² The fast-and-easy solution used in older versions of Coq was to simply evar-normalize the goal before walking it. That is, every tactic that had to walk the goal for any reason whatsoever would create a copy of the type of the goal—and sometimes the proof context as well—replacing all instantiated evars with their instantiation. Needless to say, this was very expensive when the size of the goal was large.

²See the discussion at Pédrot [154] for more details.

[TODO: include performance graphs?]

As of Coq 8.7, most tactics no longer perform evar normalization, and instead walk terms using a dedicated API which does on-the-fly normalization as necessary. [154] This brought speedups of over 10% to some developments, and improved asymptotic performance of some tactic scripts and interactive proof development.

8.1.2 Delaying the Externalization of Application Arguments

Coq has many representations of terms. There is `constr_expr`, the AST produced by Coq’s parser. Internalization turns `constr_expr` into the untyped `glob_constr` representation of terms by performing name resolution, bound variable checks, notation desugaring, and implicit argument insertion. [56] Type inference fills in the holes in untyped `glob_constrs` to turn them into typed `constrs`, possibly with remaining existential variables. [163] In order to display proof goals, this process must be reversed. The internal representation of `constr` must be “detyped” into `glob_constrs`, which involves primarily just turning de Bruijn indices into names. [67] Finally, implicit arguments must be erased and notations must be re-sugared when externalizing `glob_constrs` into `constr_exprs`, which can be printed relatively straightforwardly. [55, 162]

In old versions, Coq would externalize the entire goal, including subterms that were never printed due to being hidden by notations and implicit arguments. Starting in version 8.5pl2, lazy externalization of function arguments was implemented. [153] This resulted in massive speed-ups to interactive development involving large goals whose biggest subterms were mostly hidden.

Changes like this one can be a game-changer for interactive proof development. The kind of development that can happen when it takes a tenth of a second to see the goal after executing a tactic is vastly different from the kind of development that can happen when it takes a full second or two. In the former case, the proof engine can almost feel like an extension of the coder’s mind, responding to thoughts about strategies to try almost as fast as they can be typed. In the latter case, development is significantly more clunky and involves much more friction.

In the same vein, bugs such as #3691 and #4819, where Coq crawled the entire evar map in `-emacs` mode (used for ProofGeneral/Emacs) looking at all instantiated evars, resulted in interactive proof times of up to half-a-second for every goal display, even when the goal was small and there was nothing in the context. Fixed in Coq 8.6, these bugs, too, got in the way of seamless proof development.

8.1.3 The \mathcal{L}_{tac} Profiler

The most common mistake in performance engineering is to blindly optimize without profiling; you most often spend your time optimizing parts

that aren't actually bottlenecks.

— Charles Leiserson (heavily paraphrased, reconstructed from memory of 6.172)

In old versions of Coq, there was no good way to profile tactic execution. Users could wrap some invocations in `time` to see how long a given tactic took, or could regularly print some output to see where execution hung. Both of these are very low-tech methods of performance debugging, and work well enough for small tactics. For debugging hundreds or thousands of lines of \mathcal{L}_{tac} code, though, these methods are insufficient.

A genuine profiler for \mathcal{L}_{tac} was developed in 2015 and integrated into Coq itself in version 8.6. [190]

For those interested in amusing quirks of implementation details, the profiler itself was relatively easy to implement. Since \mathcal{L}_{tac} already records backtraces for error reporting, it was a relatively simple matter to hook into the stack-trace-recorder and track how much time was spent in each call-stack.

[**TODO:** should I include more text here, extracting from the abstract submitted to CoqPL?]

8.1.4 Compilation to Native Code

Starting in version 8.5, Coq allows users to compile their functional Gallina programs to native code and fully reduce them to determine their output. [34, 63] In some cases, the native compiler is almost $10\times$ faster³ than the optimized call-by-value evaluation bytecode-based virtual machine described in Grégoire and Leroy [85].

The native compiler shines most at optimizing algorithmic and computational bottlenecks. For example, computing the number of primes less than n via the Sieve of Eratosthenes is about $2\times$ to $5\times$ faster in the native compiler than in the VM. [**TODO:** should I include code and/or plots?] By contrast, when the input term is very large compared to the amount of computation, the compilation time can dwarf the running time, eating up any gains that the native compiler has over the VM. This can be seen by comparing the times it takes to get the head of the explicit list of all unary-encoded natural numbers less than, say, 3000, on which the native compiler (1.7s) is about 5% slower than the VM (1.6s) which itself is about $2\times$ slower than built-in call-by-value reduction machine (0.79s) which requires no translation. Furthermore, when the output is large, both the VM and the native compiler suffer from inefficiencies in the readback code. [**TODO:** should I add an example or plots?]

[**TODO:** Should I add more to this section? What should I say?]

³<https://github.com/coq/coq/pull/12405#issuecomment-633612308>

8.1.5 Primitive Integers and Arrays

Primitive 31-bit integer arithmetic operations were added to Coq in 2007. [184, 17] Although most of Coq merely used an inductive representation of 31-bit integers, the VM included code for compiling these constants to native machine integers.⁴ After hitting memory limits in storing the inductive representations in proofs involving proof traces from SMT solvers, work was started to allow the use of primitive datatypes that would be stored efficiently in proof terms. [65]

Some of this work has since been merged into Coq, including IEEE 754-2008 binary64 floating point numbers merged in Coq 8.11 [132], 63-bit integers merged in Coq 8.10 [66], and persistent arrays [53] merged into Coq 8.13 [64]. Work enabling primitive recursion over these native datatypes is still underway, [62] and the actual use of these primitive datatypes to reap the performance benefits is still to come as of the writing of this thesis.

8.1.6 Primitive Projections for Record Types

Since version 8.5, Coq has had the ability to define record types with projections whose arguments are not stored in the term representation. [179] This allows asymptotic speedups, as discussed in Subsection 3.5.4 (Nested Σ Types). [TODO: what should be said in this section?]

Note that this is a specific instance of a more general theory of implicit arguments [136, 26], and there has been other work on how to eliminate useless arguments from term representations. [39]

8.1.7 Fast Typing of Application Nodes

In Section 2.2.3 (Quadratic Substitution in Function Application), we discussed how the typing rule for function application resulted in quadratic performance behavior when there was in fact only linear work that needed to be done. As of Coq 8.10, when typechecking applications in the kernel, substitution is delayed so as to achieve linear performance. [152] [TODO: does this need more of a recap?] Unfortunately, the pretyping and type inference algorithm is still quadratic, due to the type theory rules used for type inference.

[TODO: should I talk about <https://github.com/coq/coq/pull/263> - Fast lookup in named contexts?] [TODO: should I talk about <https://github.com/coq/coq/pull/6506> - Fast rel lookup?]

⁴The integer arithmetic is 31-bit rather than 32-bit because OCaml reserves the lowest bit for tagging whether a value is a pointer address to a tagged value or an integer.

8.2 Performance-Enhancing Advancements in the Type Theory of Coq

While some of the above performance enhancements touch the trusted kernel of Coq, they do not fundamentally change the type theory. Some performance enhancements require significant changes to the type theory. In this section we will review a couple of particularly important changes of this kind.

8.2.1 Universe Polymorphism

Recall that the main case study of Chapter 3 was our implementation of a category theory library. Recall also from Type Size Blowup: Packed vs. Unpacked Records how the choice of whether to use packed or unpacked records impacts performance; while unpacked records are more friendly for developing algebraic hierarchies, packed records achieve significantly better performance when large towers of dependent concepts (such as categories, functors between categories, and natural transformations between functors) are formalized.

This section addresses a particular feature which allows an entire-library $2\times$ speed-up when using fully-packed records. How is such a large performance gain achievable? Without this feature, called *universe polymorphism*, encoding some mathematical objects requires *duplicating* the entire library! Removing this duplication of code will halve the compile-time.

What are universes?

Universes are type theory's answer to Russell's paradox. [99] Russell's paradox, a famous paradox discovered in 1901, proceeds as follows. A *set* is an unordered collection of distinct objects. Since each *set* is an object, we may consider the set of all sets. Does this set contain itself? It must, for by definition it contains all sets.

So we see by example that some sets contain themselves, while others (such as the empty set with no objects) do not. Let us consider now the set consisting of exactly the sets that do not contain themselves. Does this set contain itself? If it does not, then it fails to live up to its description as the set of *all* sets that do not contain themselves. However, if it does contain itself, then it also fails to live up to its description as a set consisting *only* of sets that do not contain themselves. Paradox!

The resolution to this paradox is to forbid sets from containing themselves. The collection of all sets is too big to be a set, so let's call it (and collections of its size) a proper class. We can nest this construction, as type theory does: We have **Type**₀, the **Type**₁ of all small types, and we have **Type**₁, the **Type**₂ of all **Type**₁s, etc. These subscripts are called *universe levels*, and the subscripted **Type**s are sometimes called *universes*.

Most constructions in Coq work just fine if we simply place them in a single, high-enough, universe. In fact, the entire standard library in Coq effectively uses only three universes. Most of the standard library in fact only needs one universe. We need a second universe for the few constructions that talk about equality between types, and a third for the encoding of a variant of Russell’s paradox in Coq.

However, one universe is not sufficient for category theory, even if we don’t need to talk about equality of types nor prove that `Type : Type` is inconsistent.

The reason is that category theory, much like set theory, talks about itself. [TODO: check if category has been defined before]

Complications from Categories of Categories

In standard mathematical practice, a category \mathcal{C} can be defined [22] to consist of:

- a class $\text{Ob}_{\mathcal{C}}$ of *objects*
- for all objects $a, b \in \text{Ob}_{\mathcal{C}}$, a class $\text{Hom}_{\mathcal{C}}(a, b)$ of *morphisms from a to b*
- for each object $x \in \text{Ob}_{\mathcal{C}}$, an *identity morphism* $1_x \in \text{Hom}_{\mathcal{C}}(x, x)$
- for each triple of objects $a, b, c \in \text{Ob}_{\mathcal{C}}$, a *composition function* $\circ : \text{Hom}_{\mathcal{C}}(b, c) \times \text{Hom}_{\mathcal{C}}(a, b) \rightarrow \text{Hom}_{\mathcal{C}}(a, c)$

satisfying the following axioms:

- *associativity*: for composable morphisms f, g, h , we have $f \circ (g \circ h) = (f \circ g) \circ h$.
- *identity*: for any morphism $f \in \text{Hom}_{\mathcal{C}}(a, b)$, we have $1_b \circ f = f = f \circ 1_a$

Some complications arise in applying the last subsection’s definition of categories to the full range of common constructs in category theory. One particularly prominent example formalizes the structure of a collection of categories, showing that this collection itself may be considered as a category.

The morphisms in such a category are *functors*, maps between categories consisting of a function on objects, a function on hom-types, and proofs that these functions respect composition and identity [124, 22, 194].

The naïve concept of a “category of all categories”, which includes even itself, leads into mathematical inconsistencies which manifest as universe inconsistency errors in Coq, much as with the set of all sets discussed above.

The standard resolution, as with sets, is to introduce a hierarchy of categories, where, for instance, most intuitive constructions are considered *small* categories, and then we also have *large* categories, one of which is the category of small categories. Both definitions wind up with literally the same text in Coq, giving:

```
Definition SmallCat : LargeCategory :=
{| Ob := SmallCategory;
  Hom C D := SmallFunctor C D;
|}.
```

It seems a shame to copy-and-paste this definition (and those of `Category`, `Functor`, etc.) n times to define an n -level hierarchy.

Universe polymorphism is a feature that allows definitions to be quantified over their universes. While Coq 8.4 supports a restricted flavor of universe polymorphism that allows the universe of a definition to vary as a function of the universes of its arguments, Coq 8.5 and later [179] support an established kind of more general universe polymorphism [93], previously implemented only in NuPRL [54]. In these versions of Coq, any definitions declared polymorphic are parametric over their universes.

While judicious use of universe polymorphism can reduce code duplication, careless use can lead to tens of thousands of universe variables which then become a performance bottleneck in their own right.⁵

8.2.2 Judgmental η for Record Types

The same commit that introduced universe polymorphism in Coq 8.5 also introduced judgmental η conversion for records with primitive projections. [179] We have already discussed the advantages of primitive projections in Subsection 8.1.6, and we have talked a bit about judgmental η in Section 3.4.1 (Dodging Judgmental η on Records) and Section 3.4 (Internalizing Duality Arguments in Type Theory).

⁵See, for example, the commit message of a445bc38c172a8f2787afe7d9d193596a0f6e1bd in the HoTT/HoTT library on GitHub, where moving from Coq 8.5 β 2 to 8.5 β 3 incurred a 4 \times slowdown in the file `hit/V.v`, entirely due to performance regressions in universe handling, which were later fixed. This slowdown is likely the one of #4537.

Coq actually had an implementation of full universe polymorphism between versions 8.3 and 8.4, implemented in commit d98dfbcae463f8d699765e2d7004becd7714d6cf and reverted mere minutes later in commit 60bc3cbe72083f4fa1aa759914936e4fa3d6b42e. In-person discussion, either with Matthieu himself or with Bob Harper, revealed that Matthieu abandoned this initial attempt after finding that universe polymorphism was too slow, and it was only by implementing the algorithm of Harper and Pollack [93] that universe polymorphism with typical ambiguity [174, 182, 93], where users need not write universe variables explicitly, was able to be implemented in a way that was sufficiently performant.

[**TODO:** make sure we've talked about judgmental equality and conversion] The η conversion rule for records says that every term x of record type T is convertible with the constructor of T applied to the projections of T applied to x . For example, if x has type $A * B$, then the η rule equates x with $(\text{fst } x, \text{ snd } x)$.

As discussed in Section 3.4, having records with judgmental η conversion allows de-duplicating code that would otherwise have to be duplicated. [**TODO:** should more be said here?]

8.2.3 SProp: The Definitionally Proof Irrelevant Universe

We discussed in Subsection 3.5.2 (Opacity; Linear Dependence of Speed on Term Size) how irrelevant proof arguments tended to cause performance issues just by existing as part of the goal. While there is as-yet no way to erase these arguments, Coq versions 8.10 and later have the ability to define types as judgmentally irrelevant, paving the way for more aggressive erasure. [75, 76] [**TODO:** should more be said here?]

8.3 Performance-Enhancing Advancements in Type Theory at Large

We come now to discoveries and inventions of the past decade or so which have not yet made it into Coq but which show great promise for significant performance improvements.

8.3.1 Higher Inductive Types: Setoids for Free

[**TODO:** is this sentence redundant?] [**TODO:** fix the intro to this subsection]

[**TODO:** fix flow] Recall again that the main case study of Chapter 3 was our implementation of a category theory library.

Equality

Equality, which has recently become a very hot topic in type theory [194] and higher category theory [119], provides another example of a design decision where most usage is independent of the exact implementation details. Although the question of what it means for objects or morphisms to be equal does not come up much in classical 1-category theory, it is more important when formalizing category theory in a proof assistant, for reasons seemingly unrelated to its importance in higher category theory. We consider some possible notions of equality.

Setoids A setoid [33] is a carrier type equipped with an equivalence relation; a map of setoids is a function between the carrier types and a proof that the function respects the equivalence relations of its domain and codomain. Many authors [155, 115, 135, 98, 9, 8, 100, 160, 177, 44, 200] choose to use a setoid of morphisms, which allows for

the definition of the category of set(oid)s, as well as the category of (small) categories, without assuming functional extensionality, and allows for the definition of categories where the objects are quotient types. However, there is significant overhead associated with using setoids everywhere, which can lead to slower compile times. Every type that we talk about needs to come with a relation and a proof that this relation is an equivalence relation. Every function that we use needs to come with a proof that it sends equivalent elements to equivalent elements. Even worse, if we need an equivalence relation on the universe of “types with equivalence relations”, we need to provide a transport function between equivalent types that respects the equivalence relations of those types.

Propositional Equality An alternative to setoids is propositional equality, which carries none of the overhead of setoids, but does not allow an easy formulation of quotient types, and requires assuming functional extensionality to construct the category of sets.

[**TODO:** check if we've defined propositional equality before] Intensional type theories like Coq's have a built-in notion of equality, often called definitional equality or judgmental equality, and denoted as $x \equiv y$. This notion of equality, which is generally internal to an intensional type theory and therefore cannot be explicitly reasoned about inside of that type theory, is the equality that holds between $\beta\delta\iota\zeta\eta$ -convertible terms.

Coq's standard library defines what is called *propositional equality* on top of judgmental equality, denoted $x = y$. One is allowed to conclude that propositional equality holds between any judgmentally equal terms.

Using propositional equality rather than setoids is convenient because there is already significant machinery made for reasoning about propositional equalities, and there is much less overhead. However, we ran into significant trouble when attempting to prove that the category of sets has all colimits, which amounts to proving that it is closed under disjoint unions and quotienting; quotient types cannot be encoded without assuming a number of other axioms.

Higher Inductive Types The recent emergence of higher inductive types allows the best of both worlds. The idea of higher inductive types [194] is to allow inductive types to be equipped with extra proofs of equality between constructors. They originated as a way to allow homotopy type theorists to construct types with non-trivial higher paths. A very simple example is the interval type, from which functional extensionality can be proven [173]. The interval type consists of two inhabitants `zero : Interval` and `one : Interval`, and a proof `seg : zero = one`. In a hypothetical type theory with higher inductive types, the type checker does the work of carrying around an equivalence relation on each type for us, and forbids users from constructing functions that do not respect the equivalence relation of any input type. For

example, we can, hypothetically, prove functional extensionality as follows:

```

Definition f_equal {A B x y} (f : A → B) : x = y → f x = f y.
Definition functional_extensionality {A B} (f g : A → B)
  : (forall x, f x = g x) → f = g
  := λ (H : ∀ x, f x = g x)
    ⇒ f_equal (λ (i : Interval) (x : A)
      ⇒ match i with
        | zero ⇒ f x
        | one ⇒ g x
        | seg ⇒ H x
      end)
    seg.
```

Had we neglected to include the branch for `seg`, the type checker should complain about an incomplete match; the function $\lambda i : \text{Interval} \Rightarrow \text{match } i \text{ with zero} \Rightarrow \text{true} \mid \text{one} \Rightarrow \text{false} \text{ end}$ of type `Interval → bool` should not typecheck for this reason.

The key insight is that most types do not need any special equivalence relation, and, moreover, if we are not explicitly dealing with a type with a special equivalence relation, then it is impossible (by parametricity) to fail to respect the equivalence relation. Said another way, the only way to construct a function that might fail to respect the equivalence relation would be by some eliminator like pattern matching, so all we have to do is guarantee that direct invocations of the eliminator result in functions that respect the equivalence relation.

As with the choice involved in defining categories, using propositional equality with higher inductive types rather than setoids derives many of its benefits from not having to deal with all of the overhead of custom equivalence relations in constructions that do not need them. In this case, we avoid the overhead by making the type checker or the metatheory deal with the parts we usually do not care about. Most of our definitions do not need custom equivalence relations, so the overhead of using setoids would be very large for very little gain.

8.3.2 Univalence and Isomorphism Transport

When considering higher inductive types, the question “when are two types equivalent?” arises naturally. The standard answer in the past has been “when they are syntactically equal”. The result of this is that two inductive types that are defined in the same way, but with different names, will not be equal. Voevodsky’s univalence principle gives a different answer: two types are equal when they are isomorphic. This principle, encoded formally as the *univalence axiom*, allows reasoning about isomorphic types as easily as if they were equal.

Tabareau et al. built a framework on top of the insights of univalence, combined with parametricity [165, 196], for automatically porting definitions and theorems to equivalent types. [188, 189]

What is the application to performance? As we saw, for example, in Section 6.2 (NbE vs. Pattern Matching Compilation: Mismatched Expression APIs and Leaky Abstraction Barriers), the choice of representation of a datatype can have drastic consequences on how easy it is to encode algorithms and write correctness proofs. These design choices can also be intricately entwined with both the compile-time and run-time performance characteristics of the code. One central message of both Chapter 6 and Chapter 3 is that picking the right API really matters when writing code with dependent types. The promise of univalence, still in its infancy, is that we could pick the right API for each algorithmic chunk, prove the APIs isomorphic, and use some version of univalence to compose the APIs and reason about the algorithms as easily as if we had used the same interface everywhere.

[**TODO:** should I talk about the difference between isomorphism and equivalence?] [**TODO:** what more should be said here?]

8.3.3 Cubical Type Theory

One important detail we elided in the previous subsections is the question of computation. Higher inductive types and univalence are much less useful if they are opaque to the type checker. The proof of function extensionality, for example, relies on the elimination rule for the interval having a judgmental computation rule.⁶

[**TODO: define point vs path constructors**] Higher inductive types whose eliminators compute on the point constructors can be hacked into dependently-typed proof assistants by adding inconsistent axioms and then hiding them behind opaque APIs so that inconsistency cannot be proven. [123, 31] This is unsatisfactory, however, on two counts:

1. The eliminators do not compute on path constructors. For example, the interval eliminator would compute on `zero` and `one`, but not on `seg`.
2. Adding these axioms compromises the trust story.

Cubical type theory is the solution to both of these problems, for both higher inductive types and univalence. [52] Unlike most other type theories, computation in cubical type theory is implemented by appealing to the category theoretic model,

⁶We leave it as a fun exercise for the advanced reader to figure out why the Church encoding of the interval, where $\text{Interval} := \forall P \ (\text{zero} : P) \ (\text{one} : P) \ (\text{seg} : \text{zero} = \text{one})$, P , does not yield a proof of functional extensionality.

and the insights that allow such computation are slowly making their way into more mainstream dependently-typed proof assistants. [195]

Chapter 9

Concluding Remarks

9.A transcript bits from Adam

And then they'll be the conclusion which I'm thinking of having as a like what are the next steps in performance of previous distance and I think this will. My current inclination is to like, Sort of point towards the paper that under has been talking about writing that's like okay, so there's a sense in which in order to do program transformation and rewriting we took the entire non-trusted part and we threw it out.

The like. And like part of that is because most uses of the non-trusted part you just cobble something together that works but if you're like tracking every single time you invoke conversion and you're like carefully piecing together something it should be possible to make something that scales. And it's not clear if that's currently even the case.

Unlike investigating that it's sort of the next wave of. Performance issues to look at. Okay. Well when you get to the conclusion of this sort of document you've pretty freehand to speculate on things and go where your heart takes you so I'm not too worried about I haven't feel like that part of the relatively quick for you to write and free of difficult choice points.

Yeah, that seems that seems mostly true. I feel like I'll have a little bit of trouble with the like first paragraph on the last paragraph of the conclusion. I'd like the transition points and they'll like actually tying it up but the body of it seems I don't expect to have that much trouble, okay?

9.B transcript bits from Rajee

[TODO: Decide between options, maybe add more text]

Option 1 Perhaps this thesis has inspired you to write your own performance system and we remind you about the things you should look out for when implementing it.

Option 2 The End

[TODO: insert category theory diagram of an End here]

Option 3 (best so far) What are the next steps in proof assistant performance. There's a paper that Andres has floated writing that I think is a good next paper to write.

Ah, that is something like okay, so you've like, Followed all the tenants that I've laid out to like have fast APIs you're like very careful about where you're having called two things. And then you start hitting so brief historical perspective. I've described a bunch of like quadratic or exponential behaviors where like you're hitting.

Areas of the system that aren't scaling nicely. There was a previous generation to this where pretty much everything was quadratic or exponential in like everything and so you couldn't do anything beyond a certain scale because everything would start blowing up on you I see and there was someone before me named George got there who when working on he was the one who led the team at Microsoft Research to you formalized the four color theorem.

I think now not the four color theorem the odd order theorem. In call okay took them about ten years. I think you've mentioned this yeah oh and he went on the cocktails and they fixed these like everything is terrible and everything. So now we're heading like problems that maybe maybe are more fundamental to proof assistance.

But like then you you design your things carefully and you're careful about which parts the system you use and you'd like count for every step. And then you start hitting the next class of problems, which is I have a couple thousand things. A couple thousand variables and I want to introduce them all oops adding a couple thousand like adding n variables is quadratic or cubic in the number of variables that I'm introducing that's unfortunate.

Um, or you're like I want to like change my goal state oops making a new goal state is linear in how many variables there that's sad now. I'm now by running time is quadratic in the number of goal states or something mm-hmm.

And like you hit all of these like the fundamental building blocks. Are too slow. And. That's sort of the next area to investigate of like how do you build a proof assistant so like what are the fundamental building blocks? How are they too slow? Huh the how do we know there are two slow what what are the factors that they're too slow and

like can we show that there's like no way to get anything to actually scale without completely re-implementing the profanion because that's basically what I what I said for program transformation.

I'm like look the existing thing it's quadratic it's real sad let's throw it out and write a new one and stick it in the part of the system that's fast. So like yeah, you can do that for all your proofs you can throw out the entire pure system and write a new one.

But like, Would be nice if you didn't have to do that we say that again. So like you're like, okay, I was trying to do this thing no just the last sentence, oh it would be nice if we didn't have to do that yeah. So the alternative is to the to the alternative is that you figure out what the primitives are what they're too slow and why they're too slow and how do you design a proof assistant like a proof engine with primitives that are actually performant that if you're carefully accounting for all of the primitive steps that you're doing in your proof then you can actually get a proof with reasonable performance.

Like all the things that I've been describing are. You slap something together and it works on small things and then you increase your you try to scale it and it's suddenly stops working because of exponential behavior. And like, Maybe there isn't a hope of fixing that if you slap something together.

But if you're like carefully engineering your proof, you should be able to avoid that. What is the careful part like can you describe that or is it just like the thing so? Okay, so here's how here's how beginners pure things in caulk. Their teacher tells them what they're trying to prove.

They look at what they're trying to prove they look at the list of things they can do they're like, oh I'm trying to prove for all X something. I know a tactic to use. I'm gonna use interest. Oh I'm trying I have a conjunction in my hypothesis that I know a tactic to use I'm gonna use destruct.

I'm trying to prove something about natural numbers, how do I prove something about natural numbers by induction? Where you have this very simple pattern match that are matching program that's running in a brain that you're like how do I do this thing one step at a time? I'm just gonna try a thing and see what works we have some arithmetic, let's try simple let's see if cock knows how to make it simpler there's a tactic called simple without the, Okay, um, sometimes it makes things much nicer.

Sometimes it makes things explode, sometimes it runs forever and gives you nothing it doesn't actually ever run forever pretty much. But running for a year is about as good as running forever.

And so you'll try it and if it works then you're like great it worked. I can keep going

yeah and if it doesn't work then you're like, oh I guess it didn't work, let me try something else instead. And like this is this is how beginners implement proofs and like the way I do proofs is I'm like, okay, let me figure out why this thing should be true.

And let me figure out what gets me closer to my understanding of why it should be true and then I run the same kind of simple program that um that beginners run that's like, oh this should be true by induction on this variable. I'm not just doing induction randomly.

I know why I'm doing induction on what and I'm like, oh I have this conjunction. I can split it apart. I have this disjunction I can split it apart and like I keep making steps and at each point. I'm like, am I still convinced that this theorem is true?

And if I have ever I'm like, oh doesn't these seem like this true anymore that I'd like backup but otherwise I just keep going as long as I'm convinced that the theorem is still reasonable. Where you say something like you do things by figuring out why something should be true is that like.

Is that like constructing approved sketching your head and then doing it versus someone being like oh I know what tactic to implement them, therefore. I will try to construct yeah it's like using a proof method to generate a proof versus knowing what you want to prove and then writing it need to or something oh where is this something different like how does it apply to the engineering case?

I think it's something like that okay, so the thing that I'm doing is I'm like do I believe that this is true when I explain to a very intelligent five-year-old why this is true. And then I'll make steps unlike if at any point. I hit a theorem that I or like I hit a state where I'm like.

This is doesn't seem true anymore. That I'll like back up but I and like I have a big sense of the proof in my head okay, oh but it's like I'm like, okay this this should follow by arithmetic. So then I do a bunch of arithmetic like things and eventually hopefully I get out a thing that's true, but it's like if I want to prove that something is true by arithmetic.

I can just like look at my thing take a step that makes the thing simpler and if the thing still seems true that I'm like great I made it simpler now what and like I can keep taking steps to make it simpler until it's done and I don't have to have a like entire proof in my head.

That's interesting. Don't yeah yeah, this is because I got lined by line feedback on my prefixes. I go along it's great. Yeah. The problem with doing things this way is that they don't scale yeah it seems hard like. It seemed like I feel like with most problems you have to kind of have a proof in your head and then use the syntax to

like.

Make it so let me let me also clarify yeah the sorts of proofs currently that you need to do in caulk or way simpler and the more tedious and the sorts of proofs that you're thinking of here's an example of a proof that you might have to do in caulk, um, this is this is like on the interesting end of proofs okay, oh if you, Have a loop that adds up all the numbers between one and m.

It's the same thing as multiplying n times, m, plus one dividing by two, okay? This is the interesting proof here's another interesting proof, we're merged sword and bubble sort give you the same list if you give them the same list then. In both cases do just do it you you prove so the things you need to prove is you need to prove that they're included or do you just run both things and say no you can't run both things because you need to prove that it's true for every single list, right?

So yeah so the way that you would prove this is you define what it means to be sorted better to be what it means to be a stable sorting of a particular list, maybe you don't need that. I think you can just define what it means to be sorted and what it means for like two lists to be the same up to permutation and you're like for any list there's a unique list, that is the same up to permutation and also sorted.

Look both of these sorting methods produce that list. Okay, oh and like this is at the interesting end the like standard end or things like, um,

If I have a binary tree that holds numbers. And I add one to all the leaves then I take the sum of all the leaves. The number that I get is the number of leaves plus the sum of all the leaves before I added one.

You know, it's these sort of like trivial structural properties.

A lot of time is that proving trivial structural properties. I see so so it's not like you like there's like a it's often not like you're missing a concept and understanding how to generate the proof or something but you need like an elegant way to like structure the proof or something and that's the part where you're saying that the beginner would just be like here's tactic.

I will apply it anywhere like oh what's the good structure to do this or something? I mean, even I'm not what's the good structure to this? I'm like what's what's the structure to do this that isn't wrong? Okay, the beginner is like, Like cargo culting Margo what cargo holding that means?

I maybe it's originated from Richard Feynman and that. There are some places where the. Ah, like livelihood of the tribes depended on like airplane deliveries of cargo. And. Like there was always a ritual associated with the cargo showing up where you like wave the lights in the air so that the airplane can land on the landing strip.

And so you can like there there were some cults so I hear I don't know how accurate this is that developed around this where people would waive the lights in the air hoping that this would make the airplanes in the cargo show up. I see right so you can do the same sort of thing with programming you're like, oh I found the program that does the thing.

I want maybe I can take the code and maybe it'll do the thing. I want. And like you'll also include all the other care and like you're like, why is this other code here? Well, because it was in this other program that did the thing that I want. I don't know if I need it.

So like why are you doing this proof like this because the example proof that I saw had this structure and it worked through the system was like, okay. Yeah, right. So I am I'm at a more advanced level where I'm like, okay, I know that's a proof things about binary trees.

I'm gonna go by induction on the binary tree and I'll do something with the number of leaves and I'll figure the rest of the details out as I go.

And this is how most proofs get done. You do them piecewise. And you like don't account for like how much work call has to do at each step, you're like try it. So work it works then it's good doesn't work that it's bad. And like you could carefully in your head design the entire proof and like carefully account for how much work you expect cock to have to do at each step and make sure that cock shouldn't have to do any work that you don't think it should do.

But very few people designed proofs like this. But it should be possible to get a proof that is fast if you design it like this and like the next wave of performance issues is going to be that even when you're designing proofs like this things are still not fast enough.

And so then that's Cox problem. That's the like, how do you design a proof assistant with good enough primitives? Right, yeah, right. I'm like, look you can just. Throughout throw out the prevention. Do this other thing instead works nicely.

But like it would be nice if you don't have to do that. To get things to scale.

And yeah, that's that's the sort of next step. Right. Now, that's not like a nicer conclusion. Next step is all these very inspiring.

Part V

Extra Material

[TODO: What do I name this part?]

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[**TODO:** maybe include all code in an appendix?]