

PROOF REPAIR

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ABSTRACT

PROOF REPAIR

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Chairs of the Supervisory Committee:

TODO

Computer Science & Engineering

Abstract will go here.

To my family.



I love all of you.

CONTENTS

1	INTRODUCTION	3
2	MOTIVATING PROOF REPAIR	5
2.1	Proof Development	5
2.2	Proof Maintenance	5
2.3	Proof Repair	6
3	PROOF REPAIR BY EXAMPLE	7
3.1	Motivating Example	7
3.2	Approach	9
3.3	Differencing	12
3.4	Transformation	13
3.5	Implementation	14
3.6	Results	18
3.7	Conclusion	24
4	PROOF REPAIR ACROSS TYPE EQUIVALENCES	25
4.1	Motivating Example	25
4.2	Approach	25
4.3	Differencing	26
4.4	Transformation	32
4.5	Implementation	32
4.6	Results	33
4.7	Conclusion	33
5	RELATED WORK	35
5.1	Programs	35
5.2	Proofs	36
6	CONCLUSIONS & FUTURE WORK	41

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INTRODUCTION

Motivation for verifying systems

Era of scale—enter proof engineering [68]

Looking back (Social Processes [28]), development has come a long way, but maintenance is still hard! And this is a problem in practice!

But missed opportunity: automation doesn't understand that proofs evolve

So we build automation that does, and we call this proof repair. Proof repair shows that there is reason to believe that verifying a modified system should often, in practical use cases, be easier than verifying the original the first time around.

Or, in other words (thesis statement): Changes in programs, specifications, and proofs carry information that a tool can extract, generalize, and apply to fix other proofs broken by the same change. A tool that automates this can save work for proof engineers relative to reference manual repairs in practical use cases.

Key technical bit: differencing and program transformations, taking advantage of the rich and structured language proofs are written in.

We implement this in a tool suite for Coq, get some sweet results.

Pave path to the next era of verification

READING GUIDE

How to read this thesis

Mapping of papers to chapters

Authorship statements for included paper materials, to credit coauthors

Expected reader background & where to find more info

2

MOTIVATING PROOF REPAIR

Before we talk more about proof repair, it helps to know what it's like to develop and maintain proofs to begin with, and what happens under the hood when you do that. This chapter gives you that context, then explains the high-level approach to proof repair that builds on that.

2.1 PROOF DEVELOPMENT

Cartoon version of development: program, spec, proof

Proof assistants: short overview of foundations & different options (survey paper), then say focus on Coq

Slightly less brief overview of Coq and its foundations and automation and so on (including proof terms), going through a running example of proof development in Coq

2.2 PROOF MAINTENANCE

Problem is when something changes—change something in running example

There are a lot of development processes people use to make proofs less likely to break to begin with (survey paper)

But still, even with these, the reality: This happens all the time (REPLICA)

And in fact not just after developing a proof, but during development too (REPLICA)

And breaks proofs even for experts (REPLICA)

And it's an extra big problem when you have a large development and the changes are outside of your control

Hence Social Processes

Why automation breaks, even with good development processes

Hence proof repair—smarter automation

2.3 PROOF REPAIR

Name inspired by program repair, but quite different as we'll soon see.

Recall thesis: Changes in programs, specifications, and proofs carry information that a tool can extract, generalize, and apply to fix other proofs broken by the same change. A tool that automates this can save work for proof engineers relative to reference manual repairs in practical use cases.

Proof repair accomplishes this using a combination of differencing and program transformations.

Differencing extracts the information from the change in program, specification, or proof.

The transformations then generalize that information to a more general fix for other proofs broken by the same change.

The details of applying the fix vary by the kind of fix, as we'll soon see.

Crucially, all of this happens over the proof terms in this rich language we saw in the Development section. This is kind of the key insight that makes it all work.

This is great because this language gives us so much information and certainty. This helps us with two of the biggest challenges from program repair. (generals related work)

But it's also challenging because this language is so unforgiving. Plus, in the end, we need these tactic proofs, not just proof terms. So we can't just reuse program repair tools. (generals related work)

So next two chapters will show two tools in our tool suite that work this way, how they handle these challenges, and how they save work.

3

PROOF REPAIR BY EXAMPLE

The first tool (PUMPKIN PATCH) focuses on changes in programs and specifications, though these changes are limited in scope as we'll see later.

What this tool does is, when programs and specifications change and this breaks a lot of proofs, it lets the proof engineer fix just one of those proofs. It then generalizes the example patch into something that can fix other proofs broken by the same change.

So in other words, the information from those changes is carried in the difference between the old and new version of the example patched proof. PUMPKIN PATCH generalizes that information.

Application can be automated in some cases at the end, or it can be manual.

The work saved is shown retroactively on case studies replaying changes from large proof developments in Git. Results for this tool are preliminary compared to what we'll see later, since this was the first prototype.

3.1 MOTIVATING EXAMPLE

Traditional proof automation considers only the current state of theorems, proofs, and definitions. This is a missed opportunity: verification projects are rarely static. Like other software, these projects evolve over time.

With traditional proof automation, the burden of change largely falls on proof engineers. This does not have to be true. Proof automation can view theorems, proofs, and definitions as fluid entities: when a proof or specification changes, a tool can search the difference between the old and new versions for a *reusable patch* that can fix broken proofs.

WITHOUT PROOF REPAIR Experienced Coq programmers use design principles and custom tactics to make proofs resilient to change. These techniques are useful for large proof developments, but they place the burden of change on the programmer. This can be problematic when change occurs outside of the programmer's control.

<pre> Definition IZR (z:Z) : R := match z with Z0 => 0 Zpos n => INR (Pos.to_nat n) Zneg n => - INR (Pos.to_nat n) end. </pre>	<pre> Definition IZR (z:Z) : R := match z with Z0 => 0 Zpos n => IPR n Zneg n => - IPR n end. </pre>
---	---

Figure 1: Old (left) and new (right) definitions of IZR in Coq. The old definition applies injection from naturals to reals and conversion of positives to naturals; the new definition applies injection from positives to reals.

Consider a commit from the Coq 8.7 release [57]. This commit redefined injection from integers to reals (Figure 1). This change broke 18 proofs in the standard library.

The Coq developer who committed the change fixed the broken proofs, then made an additional 12 commits to address the change in `coq-contribs`, a regression suite of projects that the Coq developers maintain as versions change. Many of these changes were simple. For example, the developer wrote a lemma that describes the change:

Lemma `INR_IPR` : $\forall p$, `INR (Pos.to_nat p) = IPR p`.

The developer then used this lemma to fix broken proofs within the standard library. For example, one proof broke on this line:

`rewrite Pos2Nat.inj_sub by trivial.` ✗

It succeeded with the lemma:

`rewrite <- 3!INR_IPR, Pos2Nat.inj_sub by trivial.` ✓

These changes are outside-facing: Coq users have to make similar changes to their own proofs when they update from Coq 8.6 to Coq 8.7. The Coq developer can update some tactics to account for this, but it is impossible to account for every tactic that users could use. Furthermore, while the developer responsible for the changes knows about the lemma that describes the change, the Coq user does not. The Coq user must determine how the definition has changed and how to address the change, perhaps by reading documentation or by talking to the developers.

WITH PROOF REPAIR When a user updates the Coq standard library, a proof repair tool can determine that the definition has changed, then analyze changes in the standard library and in `coq-contribs` that resulted from the change in definition (in this case, rewriting by the lemma). It can extract a reusable patch from those changes, which it can automatically apply within broken user proofs. The user never has to consider how the definition has changed.

3.2 APPROACH

In the example from Section 3.1, we can see how the example change in one proof carries enough information to fix other proofs broken by the same change (namely the rewrite by `INR_IPR`). So a tool can extract that, generalize it, and use it to fix other proofs broken by the same change.

The key insight behind PUMPKIN’s approach is that this is true more generally. To use PUMPKIN, the programmer modifies a single proof script to provide an *example* of how to adapt a proof to a change. PUMPKIN extracts that information into a *patch candidate*—which is localized to the context of the example, but not enough to fix other proofs broken by the change. It then generalizes that candidate into a *reusable patch*: a function that can be used to fix other broken proofs broken by the same change, which PUMPKIN defines as a Coq term.

In other words, looking back to the thesis statement, the information shows up in the difference between versions of the example patched proof. PUMPKIN can extract and generalize that information. Application works with hint databases or is manual. Here is the system diagram for PUMPKIN. The PUMPKIN repository contains a detailed user guide.

As mentioned earlier, PUMPKIN does this using a combination of semantic differencing and program transformations. Differencing looks at the difference between versions of the example patched proof for this information, and finds the candidate. Then, program transformations modify that candidate to produce the reusable proof patch.

And of course all of this happens over proof terms, since tactics might hide necessary information. Of course this is hard to see on the example from Section 3.1, since we were lucky enough to see the difference in tactics here. Let’s look at a toy example for which that isn’t true.

To motivate this workflow, consider using PUMPKIN to search the proofs in Figure 2 for a patch between conclusions. Except we will show a place where the lemma is actually applied. Note that the tactics don’t change even though the terms do—and even though the change could break other proofs.

So what do we do? We invoke the plugin using `old` and `new` as the example change:

```
Patch Proof old new as patch.
```

PUMPKIN first determines the type that a patch from `new` to `old` should have. To determine this, it semantically *diffs* the types and finds this goal type (line 2):

```
∀ n m p, n <= m -> m <= p -> n <= p -> n <= p + 1
```

It then breaks each inductive proof into cases and determines an intermediate goal type for the candidate. In the base case, for example,

<pre> 1 Theorem old: ∀ (n m p : nat), n <= m -> m <= p -> 2 n <= p + 1. (* P p *) 3 Proof. 4 intros. induction H0. 5 - auto with arith. 6 - constructor. auto. 7 Qed. 8 9 fun (n m p : nat) (H : n <= m 10) (H0 : m <= p) => 11 le_ind 12 m 13 (* 14 m *) 15 (fun p0 => n <= p0 + 1) 16 (* P *) 17 (le_plus_trans n m 1 H) 18 (* : P m *) 19 (fun (m0 : nat) (_ : m <= 20 m0) (IHle : n <= m0 + 1) => 21 le_S n (m0 + 1) IHle) 22 p 23 (* 24 p *) 25 H0 </pre>	<pre> 1 Theorem new: ∀ (n m p : nat 2), n <= m -> m <= p -> 3 n <= p. 4 (* P' p *) 5 Proof. 6 intros. induction H0. 7 - auto with arith. 8 - constructor. auto. 9 Qed. 10 11 fun (n m p : nat) (H : n <= 12 m) (H0 : m <= p) => 13 le_ind 14 m 15 (* m *) 16 (fun p0 => n <= p0) 17 (* P' *) 18 H 19 (* : P' m *) 20 (fun (m0 : nat) (_ : m 21 <= m0) (IHle : n <= m0) => 22 le_S n m0 IHle) 23 p 24 (* p *) 25 H0 </pre>
--	---

Figure 2: Two proofs with different conclusions (top) and the corresponding proof terms (bottom) with relevant type information. We highlight the change in theorem conclusion and the difference in terms that corresponds to a patch.

it *diffs* the types and determines that a candidate between the base cases of *new* and *old* should have this type (lines 11 and 12):

```
(fun p0 => n <= p0) m -> (fun p0 => n <= p0 + 1) m
```

It then *diffs* the terms (line 13) for such a candidate:

```
fun n m p H0 H1 =>
  (fun (H : n <= m) => le_plus_trans n m 1 H)
: ∀ n m p, n <= m -> m <= p -> n <= m -> n <= m + 1
```

This candidate is close, but it is not yet a patch. This candidate maps base case to base case (it is applied to *m*); the patch should map conclusion to conclusion (it should be applied to *p*).

This is where the transformations come in. There are four:

1. *Patch specialization* to arguments
2. *Patch abstraction* of arguments or functions
3. *Patch inversion* to reverse a patch
4. *Lemma factoring* to break a term into parts

Here, PUMPKIN *abstracts* this candidate by *m* (line 11), which lifts it out of the base case:

```
fun n0 n m p H0 H1 =>
  (fun (H : n <= n0) => le_plus_trans n n0 1 H)
: ∀ n0 n m p, n <= m -> m <= p -> n <= n0 -> n <= n0 + 1
```

PUMPKIN then *specializes* this candidate to *p* (line 16), the argument to the conclusion of *le_ind*. This produces a patch:

```
patch n m p H0 H1 :=
  (fun (H : n <= p) => le_plus_trans n p 1 H)
: ∀ n m p, n <= m -> m <= p -> n <= p -> n <= p + 1
```

The user can then use *patch* to fix other broken proofs. For example, given a proof that applies *old*, the user can use *patch* to prove the same conclusion by applying *new*:

```
apply old.✓
apply patch. apply new.✓
```

This can happen automatically through hint databases.

This simple example uses only two transformations. The other transformations help turn candidates into patches in similar ways. We discuss all of this in detail later.

CONFIGURATION The components come together to form a proof patch finding procedure:

Pseudocode: find_patch(term, term', direction)

- 1: *diff* types of term and term' for goals
 - 2: *diff* term and term' for candidates
 - 3: **if** there are candidates **then**
 - 4: *factor, abstract, specialize, and/or invert* candidates
 - 5: **if** there are patches **then return** patches
 - 6: **return** failure
-

PUMPKIN infers a *configuration* from the example change. This configuration customizes the highlighted lines for an entire class of changes: It determines what to diff on lines 1 and 2, and how to use the components on line 4.

For example, to find a patch for Figure 2, PUMPKIN used the configuration for changes in conclusions of two proofs that induct over the same hypothesis. Given two such proofs:

$$\begin{array}{l} \forall x, H\ x \rightarrow P\ x \\ \forall x, H\ x \rightarrow P'\ x \end{array}$$

PUMPKIN searches for a patch with this type:

$$\forall x, H\ x \rightarrow P'\ x \rightarrow P\ x$$

using this configuration:

```
1: diff conclusion types for goals
2: diff conclusion terms for candidates
3: if there are candidates then
4:   abstract and then specialize candidates
```

Later we will see real-world examples that demonstrate more configurations.

3.3 DIFFERENCING

The tool should be able to identify the semantic difference between terms. The semantic difference is the difference between two terms that corresponds to the difference between their types. Consider the base case terms in Figure 2 (line 13):

$$\begin{array}{l} \text{le_plus_trans } n\ m\ 1\ H : n \leq m + 1 \\ \text{le_plus_trans } n\ m\ 1\ H : n \leq m \end{array}$$

The semantic differencing component first identifies the difference in their types, or the *goal type*:

$$n \leq m \rightarrow n \leq m + 1$$

It then finds a difference in terms that has that type:

$$\text{fun } (H : n \leq m) \Rightarrow \text{le_plus_trans } n\ m\ 1\ H$$

This is the *candidate* for a reusable patch that the other components modify to find a patch.

Differencing operates over terms and types. Differencing tactics is insufficient, since tactics and hints may mask patches (line 5).¹ Furthermore, differencing is aware of the semantics of terms and types. Simply exploring the syntactic difference makes it hard to identify which changes are meaningful. For example, in the inductive case (line 14), the inductive hypothesis changes:

$$\begin{array}{l} \dots (\text{IHle} : n \leq m0 + 1) \dots \\ \dots (\text{IHle} : n \leq m0) \dots \end{array}$$

¹ Since this is a simple example, replaying an existing tactic happens to work. There are additional examples in the repository (Cex.v).

However, the type of `IH1e` changes for *any* two inductive proofs over `1e` with different conclusions. A syntactic differencing component may identify this change as a candidate. Our semantic differencing component knows that it can ignore this change.

Plus parts of Inside the Core, Testing Boundaries, Future Work

How differencing works in detail

Limitations and whether they're addressed in later tools yet or not

3.4 TRANSFORMATION

PATCH SPECIALIZATION The tool should be able to specialize a patch candidate to specific arguments as determined by the differences in terms. To find a patch for Figure 2, for example, PUMPKIN must specialize the patch candidate to `p` to produce the final patch.

PATCH ABSTRACTION A tool should be able to abstract patch candidates of this form by the common argument:

```
candidate : P' t -> P t
candidate_abs : ∀ t0, P' t0 -> P t0
```

and it should be able to abstract patch candidates of this form by the common function:

```
candidate : P t' -> P t
candidate_abs : ∀ P0, P0 t' -> P0 t
```

This is necessary because the tool may find candidates in an applied form. For example, when searching for a patch between the proofs in Figure 2, PUMPKIN finds a candidate in the difference of base cases. To produce a patch, PUMPKIN must abstract the candidate by the argument `m`. Abstracting candidates is not always possible; abstraction will necessarily be a collection of heuristics.

PATCH INVERSION The tool should be able to invert a patch candidate. This is necessary to search for isomorphisms. It is also necessary to search for implications between propositionally equal types, since candidates may appear in the wrong direction. For example, consider two list lemmas (we write `length` as `len`):

```
old : ∀ l' l, len (l' ++ l) = len l' + len l
new : ∀ l' l, len (l' ++ l) = len l' + len (rev l)
```

If PUMPKIN searches the difference in proofs of these lemmas for a patch from the conclusion of `new` to the conclusion of `old`, it may find a candidate *backwards*:

```
candidate l' l (H : old l' l) :=
  eq_ind_r ... (rev_length l)
: ∀ l' l, old l' l -> new l' l
```

The component can invert this to get the patch:

```
patch l' l (H : new l' l) :=
  eq_ind_r ... (eq_sym (rev_length l))
: ∀ l' l, new l' l -> old l' l
```

We can then use this patch to port proofs. For example, if we add this patch to a hint database [1], we can port this proof:

```
Theorem app_rev_len : ∀ l l',
  len (rev (l' ++ l)) = len (rev l) + len (rev l').
Proof.
  intros. rewrite rev_app_distr. apply old. ✓
Qed.
```

to this proof:

```
Theorem app_rev_len : ∀ l l',
  len (rev (l' ++ l)) = len (rev l) + len (rev l').
Proof.
  intros. rewrite rev_app_distr. apply new. ✓
Qed.
```

Rewrites like `candidate` are *invertible*: We can invert any rewrite in one direction by rewriting in the opposite direction. In contrast, it is not possible to invert the patch PUMPKIN found for Figure 2. Inversion will necessarily sometimes fail, since not all terms are invertible.

LEMMA FACTORING The tool should be able to factor a term into a sequence of lemmas. This can help break other problems, like abstraction, into smaller subproblems. It is also necessary to invert certain terms. Consider inverting an arbitrary sequence of two rewrites:

$$t := \text{eq_ind_r } G \dots (\text{eq_ind_r } F \dots)$$

We can view t as a term that composes two functions:

$$\begin{aligned} g &:= \text{eq_ind_r } G \dots \\ f &:= \text{eq_ind_r } F \dots \\ t &:= g \circ f \end{aligned}$$

The inverse of t is the following:

$$t^{-1} := f^{-1} \circ g^{-1}$$

To invert t , PUMPKIN identifies the factors $[f; g]$, inverts each factor to $[f^{-1}; g^{-1}]$, then folds and applies the inverse factors in the opposite direction.

plus parts of PUMPKIN PATCH Inside the Core, Testing Boundaries, Future Work

How the four transformations work in detail

Limitations and whether they're addressed in later tools yet or not

3.5 IMPLEMENTATION

parts of PUMPKIN PATCH Inside the Core, plus more

3.5.1 Tool Details

While our system is a very early prototype under active development, we have made the source code available on Github.² The interested

² <http://github.com/uwplse/PUMPKIN-PATCH/tree/cpp18>

reader can follow along in the repository. Our prototype has no impact on the trusted computing base (Section 3.5.2.1).

3.5.1.1 Semantic Differencing

We implement semantic differencing over *trees*: PUMPKIN compiles each proof term into a tree (`evaluation.ml`). In these trees, every node is a type context, and every edge is an extension to that type context with a new term.³ Correspondingly, type differencing (to identify goal types) compares nodes, and term differencing (to find candidates) compares edges.

The component (`differencing.ml`) uses these nodes and edges to prioritize semantically relevant differences. At the lowest level, it calls a primitive differencing function which checks if it can substitute one term within another term to find a function between their types.

The key benefit to this model is that it gives us a natural way to express inductive proofs, so that differencing can efficiently identify good candidates. Consider, for example, searching for a patch between conclusions of two inductive proofs of theorems about the natural numbers:

```
nat_ind P ... (fun (IH : P n) => ...) : ∀ n, P n
nat_ind P' ... (fun (IH : P' n) => ...) : ∀ n, P' n
```

In each case, the component diffs the terms in the dotted edges of the tree for `nat_ind` (Figure 3) to try to find a term that maps between conclusions of that case:

```
P' 0 -> P 0 (* base case candidate *)
P' (S n) -> P (S n) (* inductive case candidate *)
```

The component also knows that the change in the type of `IH` is inconsequential (it occurs for any change in conclusion). Furthermore, it knows that `IH` cannot show up as a hypothesis in the patch, so it attempts to remove any occurrences of `IH` in any candidate.

When the component finds a candidate, it knows `P'` and `P` as well as the arguments `0` or `(S n)`. This makes it simple to query abstraction for the final patch:

```
∀ n, P' n -> P n
```

The differencing component is *lazy*: it compiles terms into trees one step at a time. It then *expands* each tree as needed to find candidates (`expansion.ml`). For example, consider searching two functions for a patch between conclusions:

```
fun (t : T) => b
fun (t' : T) => b'
```

Differencing introduces a single term of type `T` to a common environment, then expands and recursively diffs the bodies `b` and `b'` in that environment.

³ These trees are inspired by categorical models of dependent type theory [37].

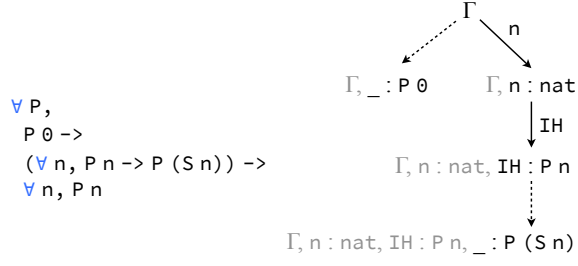


Figure 3: The type of (left) and tree for (right) the induction principle `nat_ind`. The solid edges represent hypotheses, and the dotted edges represent the proof obligations for each case in an inductive proof.

The tool always maintains pointers to easily switch between the tree and AST representations of the terms. This representation enables extensibility.

3.5.1.2 Transformations

PATCH SPECIALIZATION Specialization (`specialize.ml`) takes a patch candidate and some arguments, all of which are Coq terms. It applies the candidate to the arguments, then it $\beta\iota$ -reduces [17] the result using Coq's `Reduction.nf_betaiota` function. It is the job of the patch finding procedure to provide both the candidate and the arguments.

PATCH ABSTRACTION Abstraction (`abstraction.ml`) takes a patch candidate, the goal type, and the function arguments or function to abstract. It first generalizes the candidate, wrapping it inside of a lambda from the type of the term to abstract. Then, it substitutes terms inside the body with the abstract term. It continues to do this until there is nothing left to abstract, then filters results by the goal type. Consider, for example, abstracting this candidate by `m`:

```
fun (H : n <= m) => le_plus_trans n m 1 H
: n <= m -> n <= m + 1
```

The generalization step wraps this in a lambda from some `nat`, the type of `m`:

```
fun (n0 : nat) =>
  (fun (H : n <= m) => le_plus_trans n m 1 H)
: ∀ n0, n <= m -> n <= m + 1
```

The substitution step replaces `m` with `n0`:

```
fun (n0 : nat) =>
  (fun (H : n <= n0) => le_plus_trans n n0 1 H)
: ∀ n0, n <= n0 -> n <= n0 + 1
```

Abstraction uses a list of *abstraction strategies* to determine what subterms to substitute. In this case, the simplest strategy works: The tool replaces all terms that are convertible to the concrete argument

m with the abstract argument $n0$, which produces a single candidate. Type-checking this candidate confirms that it is a patch.

In some cases, the simplest strategy is not sufficient, even when it is possible to abstract the term. It may be possible to produce a patch only by abstracting *some* of the subterms convertible to the argument or function (we show an example of this in Section ??), or the term may not contain any subterms convertible to the argument or function at all. We implement several strategies to account for this. The combinations strategy, for example, tries all combinations of substituting only some of the convertible subterms with the abstract argument. The pattern-based strategy substitutes subterms that match a certain pattern with a term that corresponds to that pattern.

It is the job of the patch finding procedure to provide the candidate and the terms to abstract. In addition, each configuration includes a list of strategies. The configuration for changes in conclusions, for example, starts with the simplest strategy, and moves on to more complex strategies only if that strategy fails. This design makes abstraction simple to extend with new strategies and simple to call with different strategies for different classes of changes.

PATCH INVERSION Patch inversion (`inverting.ml`) exploits symmetry to try to reverse the conclusions of a candidate patch. It first factors the candidate using the factoring component, then calls the primitive inversion function on each factor, then finally folds the resulting list in reverse. The primitive inversion function exploits symmetry. For example, equality is symmetric, so the component can invert any application of `eq_ind` or `eq_ind_r` (any rewrite). Indeed, `eq_ind` and `eq_ind_r` are inverses, and are related by symmetry:

```
eq_ind_r A x P (H : P x) y (H0 : y = x) :=
  eq_ind x (fun y0 : A => P y0) H y (eq_sym H0)
```

If inversion does not recognize that the type is symmetric, it swaps subterms and type-checks the result to see if it is an inverse.

LEMMA FACTORING The lemma factoring component (`factoring.ml`) searches within a term for its factors. For example, if the term composes two functions, it returns both factors:

```
t : X -> Z (* term *)
[f : X -> Y; g : Y -> Z] (* factors *)
```

In this case, the component takes the composite term and x as arguments. It first searches as deep as possible for a term of type $x \rightarrow y$ for some y . If it finds such a term, then it recursively searches for a term with type $y \rightarrow z$. It maintains all possible paths of factors along the way, and it discards any paths that cannot reach z .

The current implementation can handle paths with more than two factors, but it fails when y depends on x . Other components may benefit from dependent factoring; we leave this to future work.

3.5.1.3 Inside the Procedure

The implementation (`patcher.ml4`) of the procedure from Section ?? starts with a preprocessing step which compiles the proof terms to trees (like the tree in Figure 3). It then searches for candidates one step at a time, expanding the trees when necessary.

The PUMPKIN prototype exposes the patch finding procedure to users through the Coq command `Patch Proof` `Proof`. PUMPKIN automatically infers which configuration to use for the procedure from the example change. For example, to find a patch for the case study in Section 3.6.1, we used this command:

```
Patch Proof Old.unsigned_range unsigned_range as patch.
```

PUMPKIN analyzed both versions of `unsigned_range` and determined that a constructor of the `int` type changed (Figure 4), so it initialized the configuration for changes in constructors.

Internally, PUMPKIN represents configurations as sets of options, which it passes to the procedure. The procedure uses these options to determine how to compose components (for example, whether to abstract candidates) and how to customize components (for example, whether semantic differencing should look for an intermediate lemma). To implement new configurations for different classes of changes, we simply tweak the options.

3.5.2 Workflow Integration

Needed: hints and so on, any work done since, the Git interface, whatever.

3.5.2.1 Trusted Computing Base

A common concern for Coq plugins is an increase in the trusted computing base. The Coq developers provide a safe plugin API in Coq 8.7 to address this [29]. Our prototype takes this into consideration: While PUMPKIN does not yet support Coq 8.7, it only calls the internal Coq functions that the developers plan to expose in the safe API [45]. Furthermore, Coq type-checks terms that plugins produce. Since PUMPKIN does not modify the type checker, it cannot produce an ill-typed term.

3.6 RESULTS

Needed: key technical results

We used the PUMPKIN prototype to emulate three motivating scenarios from real-world code:

1. **Updating definitions** within a project
(CompCert, Section 3.6.1)

<pre>Record int : Type := mkint { intval: Z; intrange : 0 <= intval < modulus }.</pre>	<pre>Record int : Type := mkint { intval: Z; intrange : -1 < intval < modulus }.</pre>
--	--

Figure 4: Old (left) and new (right) definitions of `int` in CompCert.

2. **Porting definitions** between libraries
(Software Foundations, Section 3.6.2)
3. **Updating proof assistant versions**
(Coq Standard Library, Section 3.6.3)

The code we chose for these scenarios demonstrated different classes of changes. For each case, we describe how PUMPKIN configures the procedure to use the core components for that class of changes. Our experiences with these scenarios suggest that patches are useful and that the components are effective and flexible.

IDENTIFYING CHANGES We identified Git commits from popular Coq projects that demonstrated each scenario. These commits updated proofs in response to breaking changes. We emulated each scenario as follows:

1. *Replay* an example proof update for PUMPKIN
2. *Search* the example for a patch using PUMPKIN
3. *Apply* the patch to fix a different broken proof

Our goal was to simulate incremental use of a patch finding tool, at the level of a small change or a commit that follows best practices. We favored commits with changes that we could isolate. When isolating examples for PUMPKIN, we replayed changes from the bottom up, as if we were making the changes ourselves. This means that we did not always make the same change as the user. For example, the real change from Section 3.6.1 updated multiple definitions; we updated only one.

PUMPKIN is a proof-of-concept and does not yet handle some kinds of proofs. In each scenario, we made minor modifications to proofs so that we could use PUMPKIN (for example, using induction instead of destruction). PUMPKIN does not yet handle structural changes like adding constructors or parameters, so we focused on changes that preserve structure, like modifying constructors. Chapter 4 describes an extension to PUMPKIN that supports changes in structure.

3.6.1 Updating Definitions

Coq programmers sometimes make changes to definitions that break proofs within the same project. To emulate this use case, we identified

<pre> Fixpoint bin_to_nat (b : bin) : nat := match b with B0 => 0 B2 b' => 2 * (bin_to_nat b ') B21 b' => 1 + 2 * (bin_to_nat b') end. </pre>	<pre> Fixpoint bin_to_nat (b : bin) : nat := match b with B0 => 0 B2 b' => (bin_to_nat b') + (bin_to_nat b') B21 b' => S ((bin_to_nat b ') + (bin_to_nat b')) end. </pre>
---	--

Figure 5: Definitions of `bin_to_nat` for Users A (left) and B (right).

a CompCert commit [48] with a breaking change to `int` (Figure 4). We used PUMPKIN to find a patch that corresponds to the change in `int`. The patch PUMPKIN found fixed broken inductive proofs.

REPLAY We used the proof of `unsigned_range` as the example for PUMPKIN. The proof failed with the new `int`:

```

Theorem unsigned_range:
  ∀(i : int), 0 <= unsigned i < modulus.
Proof.
  intros i. induction i using int_ind; auto. ✗

```

We replayed the change to `unsigned_range`:

```

intros i. induction i using int_ind. simpl. omega. ✓

```

SEARCH We used PUMPKIN to search the example for a patch that corresponds to the change in `int`. It found a patch with this type:

```

∀ z : Z, -1 < z < modulus -> 0 <= z < modulus

```

APPLY After changing the definition of `int`, the proof of the theorem `repr_unsigned` failed on the last tactic:

```

Theorem repr_unsigned:
  ∀(i : int), repr (unsigned i) = i.
Proof.
  ... apply Zmod_small; auto. ✗

```

Manually trying `omega`—the tactic which helped us in the proof of `unsigned_range`—did not succeed. We added the patch that PUMPKIN found to a hint database. The proof of the theorem `repr_unsigned` then went through:

```

... apply Zmod_small; auto. ✓

```

3.6.1.1 Configuration

This scenario used the configuration for changes in constructors of an inductive type. Given such a change:

```

Inductive T := ... | C : ... -> H -> T
Inductive T' := ... | C : ... -> H' -> T'

```

PUMPKIN searches two inductive proofs of theorems:

```

∀ (t : T), P t
∀ (t : T'), P t

```

for an isomorphism⁴ between the constructors:

```
... -> H -> H'
... -> H' -> H
```

The user can apply these patches within the inductive case that corresponds to the constructor *C* to fix other broken proofs that induct over the changed type. PUMPKIN uses this configuration for changes in constructors:

-
- 1: *diff* inductive constructors for goals
 - 2: use *all components* to recursively search for changes in conclusions of the corresponding case of the proof
 - 3: **if** there are candidates **then**
 - 4: try to *invert* the patch to find an isomorphism
-

3.6.2 Porting Definitions

Coq programmers sometimes port theorems and proofs to use definitions from different libraries. To simulate this, we used PUMPKIN to port two solutions [2, 6] to an exercise in Software Foundations to each use the other solution’s definition of the fixpoint `bin_to_nat` (Figure 5). We demonstrate one direction; the opposite was similar.

REPLAY We used the proof of `bin_to_nat_pres_incr` from User A as the example for PUMPKIN. User A cut an inline lemma in an inductive case and proved it using a rewrite:

```
assert (∀ a, S (a + S (a + 0)) = S (S (a + (a + 0)))).
- ... rewrite <- plus_n_0. rewrite -> plus_comm.
```

When we ported User A’s solution to use User B’s definition of `bin_to_nat`, the application of this inline lemma failed. We changed the conclusion of the inline lemma and removed the corresponding rewrite:

```
assert (∀ a, S (a + S a) = S (S (a + a))).
- ... rewrite -> plus_comm.
```

SEARCH We used PUMPKIN to search the example for a patch that corresponds to the change in `bin_to_nat`. It found an isomorphism:

```
∀ P b, P (bin_to_nat b) -> P (bin_to_nat b + 0)
∀ P b, P (bin_to_nat b + 0) -> P (bin_to_nat b)
```

APPLY After porting to User B’s definition, a rewrite in the proof of the theorem `normalize_correctness` failed:

```
Theorem normalize_correctness:
  ∀ b, nat_to_bin (bin_to_nat b) = normalize b.
Proof.
  ... rewrite -> plus_0_r. X
```

⁴ If PUMPKIN finds just one implication, it returns that.

Attempting the obvious patch from the difference in tactics—rewriting by `plus_n_0`—failed. Applying the patch that PUMPKIN found fixed the broken proof:

```
... apply patch_inv. rewrite -> plus_0_r.✓
```

In this case, since we ported User A’s definition to a simpler definition,⁵ PUMPKIN found a patch that was not the most natural patch. The natural patch would be to remove the `rewrite`, just as we removed a different `rewrite` from the example proof. This did not occur when we ported User B’s definition, which suggests that in the future, a patch finding tool may help inform novice users which definition is simpler: It can factor the proof, then inform the user if two factors are inverses. Tactic-level changes do not provide enough information to determine this; the tool must have a semantic understanding of the terms.

3.6.2.1 Configuration

This scenario used the configuration for changes in cases of a fixpoint. Given such a change:

```
Fixpoint f ... := ... | g x
Fixpoint f' ... := ... | g x'
```

PUMPKIN searches two proofs of theorems:

$$\begin{array}{l} \forall \dots, P(f \dots) \\ \forall \dots, P(f' \dots) \end{array}$$

for an isomorphism that corresponds to the change:

$$\begin{array}{l} \forall P, P \ x \rightarrow P \ x' \\ \forall P, P \ x' \rightarrow P \ x \end{array}$$

The user can apply these patches to fix other broken proofs about the fixpoint.

The key feature that differentiates these from the patches we have encountered so far is that these patches hold for *all* P ; for changes in fixpoint cases, the procedure abstracts candidates by P , not by its arguments. PUMPKIN uses this configuration for changes in fixpoint cases:

-
- 1: *diff* fixpoint cases for goals
 - 2: use *all components* to recursively search an intermediate lemma for a change in conclusions
 - 3: **if** there are candidates **then**
 - 4: *specialize* and *factor* the candidate
 abstract the factors by functions
 try to *invert* the patch to find an isomorphism
-

For the prototype, we require the user to cut the intermediate lemma explicitly and to pass its type and arguments. In the future, an improved semantic differencing component can infer both the

⁵ User A uses `*`; User B uses `+`. For arbitrary n , the term $2 * n$ reduces to $n + (n + 0)$, which does not reduce any further.

Definition divide p q := \exists r, p * r = q.	Definition divide p q := \exists r, q = r * p.
--	--

Figure 6: Old (left) and new (right) definitions of divide in Coq.

intermediate lemma and the arguments: It can search within the proof for some proof of a function that is applied to the fixpoint.

3.6.3 Updating Proof Assistant Versions

Coq sometimes makes changes to its standard library that break backwards-compatibility. To test the plausibility of using a patch finding tool for proof assistant version updates, we identified a breaking change in the Coq standard library [49]. The commit changed the definition of `divide` prior to the Coq 8.4 release (Figure 6). The change broke 46 proofs in the standard library. We used PUMPKIN to find an isomorphism that corresponds to the change in `divide`. The isomorphism PUMPKIN found fixed broken proofs.

REPLAY We used the proof of `mod_divide` as the example for PUMPKIN. The proof broke with the new `divide`:

```
Theorem mod_divide:
   $\forall$  a b, b $\neq$ 0 -> (a mod b == 0 <-> (divide b a)).
Proof.
... rewrite (div_mod a b Hb) at 2.✗
```

We replayed changes to `mod_divide`:

```
... rewrite mul_comm. symmetry.
rewrite (div_mod a b Hb) at 2.✓
```

SEARCH We used PUMPKIN to search the example for a patch that corresponds to the change in `divide`. It found an isomorphism:

```
 $\forall$  r p q, p * r = q -> q = r * p
 $\forall$  r p q, q = r * p -> p * r = q
```

APPLY The proof of the theorem `Zmod_divides` broke after rewriting by the changed theorem `mod_divide`:

```
Theorem Zmod_divides:
   $\forall$  a b, b<>0 -> (a mod b = 0 <->  $\exists$  c, a = b * c).
Proof.
... split; intros (c,Hc); exists c; auto.✗
```

Adding the patches PUMPKIN found to a hint database made the proof go through:

```
... split; intros (c,Hc); exists c; auto.✓
```

3.6.3.1 Configuration

This scenario used the configuration for changes in dependent arguments to constructors. PUMPKIN searches two proofs that apply the same constructor to different dependent arguments:

$$\begin{array}{c} \dots (C (P \ x)) \dots \\ \dots (C (P' \ x)) \dots \end{array}$$

for an isomorphism between the arguments:

$$\begin{array}{l} \forall x, P \ x \rightarrow P' \ x \\ \forall x, P' \ x \rightarrow P \ x \end{array}$$

The user can apply these patches to patch proofs that apply the constructor (in this case *study*, to fix broken proofs that instantiate *divide* with some specific *r*).

So far, we have encountered changes of this form as arguments to an induction principle; in this case, the change is an argument to a constructor. A patch between arguments to an induction principle maps directly between conclusions of the new and old theorem without induction; a patch between constructors does not. For example, for *divide*, we can find a patch with this form:

$$\forall x, P \ x \rightarrow P' \ x$$

However, without using the induction principle for *exists*, we can't use that patch to prove this:

$$(\exists x, P \ x) \rightarrow (\exists x, P' \ x)$$

This changes the goal type that semantic differencing determines. PUMPKIN uses this configuration for changes in constructor arguments:

-
- 1: *diff* constructor arguments for goals
 - 2: use *all components* to recursively search those arguments for changes in conclusions
 - 3: if there are candidates **then**
 - 4: *abstract* the candidate
 factor and try to *invert* the patch to find an isomorphism
-

For the prototype, the model of constructors for the semantic differencing component is limited, so we ask the user to provide the type of the change in argument (to guide line 2). We can extend semantic differencing to remove this restriction.

3.7 CONCLUSION

Rehashing thesis and how we do it

What we haven't accomplished yet at this point (parts of PUMPKIN PATCH future work), segue into next chapter

4

PROOF REPAIR ACROSS TYPE EQUIVALENCES

This extension to the suite adds support for a broad class of changes in datatypes, handling a large class of practical repair scenarios. What this tool (PUMPKIN Pi) does is, when datatypes change and this breaks a lot of proofs, it generalizes the change in datatype itself (possibly with some user input) so that it can automatically fix proofs broken by the change in datatype.

So in other words, the information from those changes is carried in the difference between the old and new version of the changed datatype, possibly with some user input.

PUMPKIN Pi generalizes that information and applies it automatically.

The work saved is shown on a lot of case studies (see Table from PUMPKIN Pi).

4.1 MOTIVATING EXAMPLE

PUMPKIN Pi motivating example

4.2 APPROACH

Parts of PUMPKIN Pi intro, problem definition, plus more

Like I mentioned earlier, this also works using differencing and program transformations. And of course all of this happens over proof terms.

Here's the system diagram.

Here, differencing thus looks at the difference between versions of the changed datatype, and finds something called a type equivalence. I'll explain that with examples. Sometimes differencing is automatic, and sometimes it's manual.

Then, program transformation ports proofs across the equivalence directly. So they take care of application.

4.3 DIFFERENCING

General idea from PUMPKIN Pi, explain automatic configurations and so on, segue into DEVoid example

4.3.1 Algebraic Ornaments

An algebraic ornament relates an inductive type A to an indexed version of that type B with a new index of type I_B , where the new index is fully determined by a unique fold over A . For example, `vector` is exactly `list` with a new index of type `nat`, where the new index is fully determined by the `length` function. Consequentially, there are two functions:

```
ltv : list T → Σ(n : nat).vector T n.
vtl : Σ(n : nat).vector T n → list T.
```

that are mutual inverses:

```
∀ (l : list T),          vtl (ltv l) = l.
∀ (v : Σ(n : nat).vector T n), ltv (vtl v) = v.
```

and therefore form the type equivalence from Section ?? . Moreover, since the new index is fully determined by `length`, we can relate `length` to `ltv`:

```
∀ (l : list T), length l = πl (ltv l).
```

In general, we can view an algebraic ornament as a type equivalence:

$$A \vec{i} \simeq \Sigma(n : I_B \vec{i}).B \text{ (index } n \vec{i})$$

where \vec{i} are the indices of A , I_B is a function over those indices, and the `index` operation inserts the new index n at the right offset. Such a type equivalence consists of two functions [73]:

```
promote : A  $\vec{i}$  → Σ(n : I_B  $\vec{i}$ ).B (index n  $\vec{i}$ ).
forget  : Σ(n : I_B  $\vec{i}$ ).B (index n  $\vec{i}$ ) → A  $\vec{i}$ .
```

that are mutual inverses:¹

```
section : ∀ (a : A  $\vec{i}$ ),          forget (promote a) = a.
retraction : ∀ (bΣ : Σ(n : I_B  $\vec{i}$ ).B (index n  $\vec{i}$ )), promote (forget bΣ) = bΣ.
```

An algebraic ornament is additionally equipped with an indexer, which is a unique fold:

```
indexer : A  $\vec{i}$  → I_B  $\vec{i}$ .
```

which projects the promoted index:

```
coherence : ∀(a : A  $\vec{i}$ ), indexer a = πl (promote a).
```

Following existing work [44], we call this equivalence the *ornamental promotion isomorphism*; when it holds and the indexer exists, we say that B is an algebraic ornament of A .

¹ The adjunction condition follows from section and retraction.

`Find ornament` searches for algebraic ornaments between types and is, to the best of our knowledge, the first search algorithm for ornaments.

In their original form, ornaments are a programming mechanism: Given a type A , an ornament determines some new type B . We invert this process for algebraic ornaments: Given types A and B , `DEVOID` searches for an ornament between them. This is possible for algebraic ornaments precisely because the indexer is extensionally unique. For example, all possible indexers for `list` and `vector` must compute the length of a list; if we were to try doubling the length instead, we would not be able to satisfy the equivalence.

`Find ornament` takes two inductive types and searches for the components of the ornamental promotion isomorphism between them:

- **Inputs:** Inductive types A and B , assuming:
 - B is an algebraic ornament of A ,
 - B has the same number of constructors in the same order as A ,
 - A and B do not contain recursive references to themselves under products, and
 - for every recursive reference to A in A , there is exactly one new hypothesis in B , which is exactly the new index of the corresponding recursive reference in B .
- **Outputs:** Functions `promote`, `forget`, and `indexer`, guaranteeing:
 - the outputs form the ornamental promotion isomorphism between the inputs.

`Find ornament` includes an option to generate a proof that the outputs form the ornamental promotion isomorphism; by default, this option is false, since `Lift` does not need this proof.

Presentation. We present both algorithms relationally, using a set of judgments; to turn these relations into algorithms, prioritize the rules by running the derivations in order, falling back to the original term when no rules match. The default rule for a list of terms is to run the derivation on each element of the list individually.

Notes on Syntax. The language the algorithms operate over is CIC_ω with primitive eliminators; this is a simplified version of the type theory underlying Coq. Figure 7 contains the syntax (which includes variables, sorts, product types, functions, inductive types, constructors, and eliminators), as well as the syntax for some judgments and operations, the rules for which are standard and thus omitted. For simplicity of presentation, we assume variables are names; we assume that all names are fresh. As in Coq, we assume the existence of an

$$\begin{array}{ll}
\langle i \rangle \in \mathbb{N}, \langle v \rangle \in \text{Vars}, \langle s \rangle \in \{ \text{Prop}, \text{Set}, \Gamma \vdash t : T // \text{ type checking} \\
\text{Type}(\langle i \rangle) \} & \Gamma \vdash t_1 \equiv_{\beta\delta i} t_2 // \text{ definitional} \\
& \text{equality} \\
\langle t \rangle ::= \langle v \rangle \mid \langle s \rangle \mid \Pi(\langle v \rangle : \langle t \rangle) . \langle t \rangle \mid & t_\beta // \text{ beta-reduction} \\
\lambda(\langle v \rangle : \langle t \rangle) . \langle t \rangle \mid \langle t \rangle \langle t \rangle \mid & t_{\beta\delta i} // \text{ normalization} \\
\text{Ind}(\langle v \rangle : \langle t \rangle) \{ \langle t \rangle, \dots, \langle t \rangle \} \mid \text{Constr } t [y / x] // \text{ substitution} \\
(\langle i \rangle, \langle t \rangle) \mid & \xi(I, Q, c, C) // \text{ type of} \\
\text{Elim}(\langle t \rangle, \langle t \rangle) \{ \langle t \rangle, \dots, \langle t \rangle \} & \text{eliminator}
\end{array}$$

Figure 7: CIC_ω syntax (left, from existing work [72]) and judgments and operations (right).

$$\begin{array}{l}
A := \\
\quad \text{Ind}(Ty_A : \Pi(\vec{i}_A : \vec{X}_A).s_A) \{C_{A_1}, \dots, C_{A_n}\} \\
B := \text{Ind}(Ty_B : \Pi(\vec{i}_B : \vec{X}_B).s_B) \{C_{B_1}, \dots, C_{B_n}\} \\
\forall 1 \leq i \leq n, \\
\quad E_{A_i} (p_A : P_A) := \\
\quad \quad \xi(A, p_A, \text{Constr}(i, A), C_{A_i}) \\
\quad E_{B_i} (p_B : P_B) := \\
\quad \quad \xi(B, p_B, \text{Constr}(i, B), C_{B_i}) \\
P_A := \\
\quad \Pi(\vec{i}_A : \vec{X}_A)(a : A \vec{i}_A).s_A \\
P_B := \Pi(\vec{i}_B : \vec{X}_B)(b : B \vec{i}_B).s_B \\
\text{index} := \text{insert} (\text{off } A \ B) \\
\text{deindex} := \\
\quad \text{remove} (\text{off } A \ B)
\end{array}$$

Figure 8: Common definitions for both algorithms.

inductive type Σ for sigma types with projections π_l and π_r ; for simplicity, we assume projections are primitive. Throughout, we use \vec{i} and $\{t_1, \dots, t_n\}$ to denote lists of terms, and we use $\vec{i}[j]$ to denote accessing the element of the list \vec{i} at offset j .

Common Definitions. The algorithms assume list insertion and removal functions `insert` and `remove`, plus two functions `DEVOID` implements: `off` computes the offset of the new index of type I_B in B 's indices, and `new` determines whether a hypothesis in a case of the eliminator type of B is new. Figure 8 contains other common definitions, the names for which are reserved: The `index` and `deindex` functions insert an index into and remove an index from a list at the index computed by `off`. Input type A expands to an inductive type with indices of types \vec{X}_A , sort s_A , and constructors $\{C_{A_1}, \dots, C_{A_n}\}$. P_A denotes the type of the motive of the eliminator of A , and each E_{A_i} denotes the type of the eliminator for the i th constructor of A . Analogous names are also reserved for input type B .

The `Find ornament` algorithm implements the specification. It builds on three intermediate steps: one to generate each of `indexer`, `promote`, and `forget`. Figure 9 shows the algorithm for generating `indexer`. The

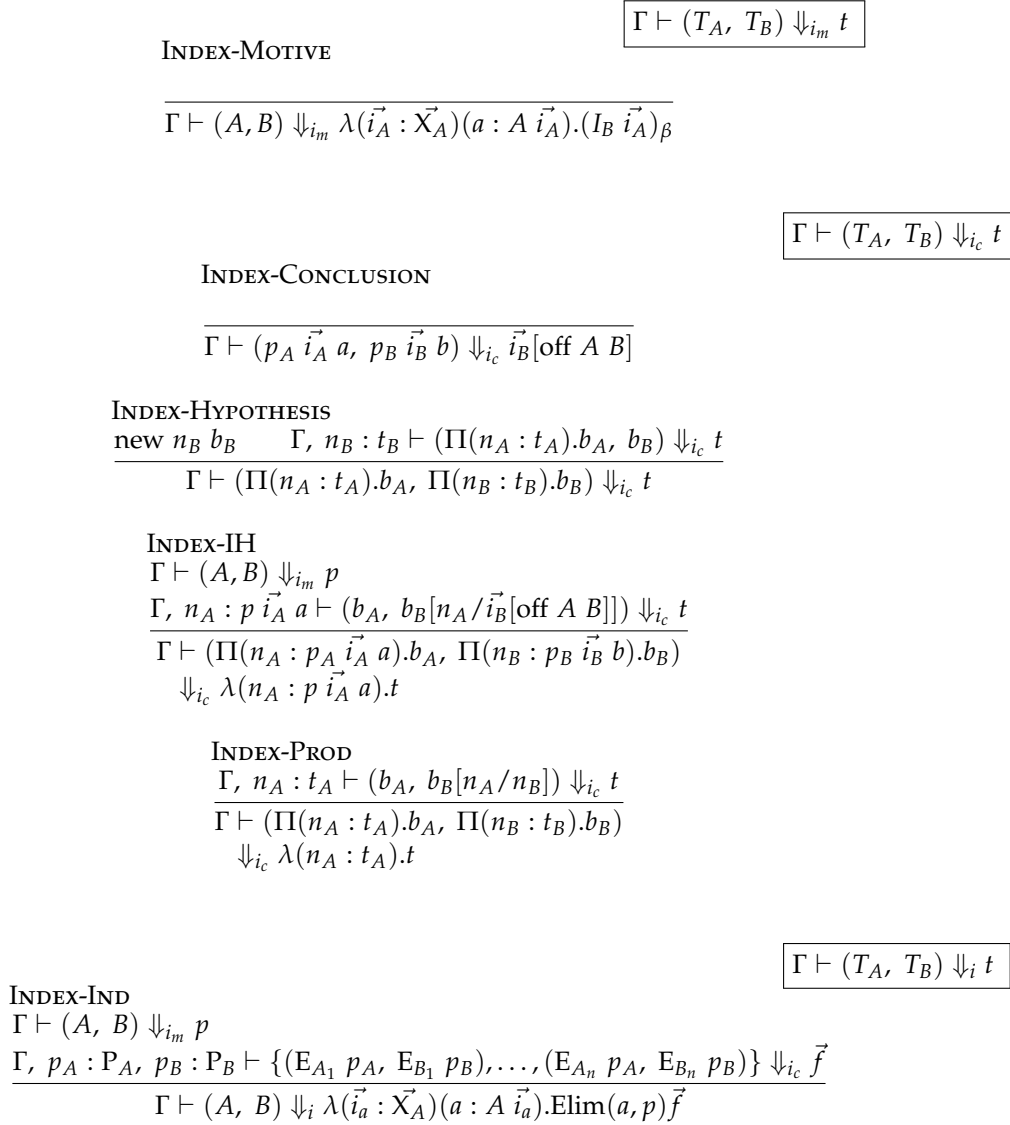


Figure 9: Identifying the indexer function.

algorithms for generating `promote` and `forget` are similar; Figure 10 shows only the derivations for generating `promote` that are different from those for generating `indexer`, and the derivations for generating `forget` are omitted.

4.3.1.1 Searching for the Indexer

Search generates the `indexer` by traversing the types of the eliminators for A and B in parallel using the algorithm from Figure 9, which consists of three judgments: one to generate the motive, one to generate each case, and one to compose the motive and cases.

Generating the Motive. The $(T_A, T_B) \Downarrow_{im} t$ judgment consists of only the derivation INDEX-MOTIVE, which computes the indexer motive from the types A and B (expanded in Figure 8). It does this by constructing a function with A and its indices as premises, and the type I_B in the conclusion with the appropriate indices. Consider `list` and `vector`:

```
list T := Ind (TyA : Type) {...}      vector T := Ind (TyB : Π
  (n : nat). Type) {...}
```

For these types, INDEX-MOTIVE computes the motive:

```
λ (l:list T) . nat
```

Generating Each Case. The $\Gamma \vdash (T_A, T_B) \Downarrow_{ic} t$ judgment generates each case of the indexer by traversing in parallel the corresponding cases of the eliminator types for A and B . It consists of four derivations: INDEX-CONCLUSION handles base cases and conclusions of inductive cases, while INDEX-HYPOTHESIS, INDEX-IH, and INDEX-PROD recurse into products.

INDEX-HYPOTHESIS handles each new hypothesis that corresponds to a new index in an inductive hypothesis of an inductive case of the eliminator type for B . It adds the new index to the environment, then recurses into the body of only the type for which the index already exists. For example, in the inductive case of `list` and `vector`, `new` determines that `n` is the new hypothesis. INDEX-HYPOTHESIS then recurses into the body of only the `vector` case:

```
Π (tl:T) (l:list T) (IHl:pA 1), ...    Π (tv:T) (v:vector T n)
  (IHv:pB n v), ...
```

INDEX-PROD is next. It recurses into product types when the hypothesis is neither a new index nor an inductive hypothesis. Here, it runs twice, recursing into the body and substituting names until it hits the inductive hypothesis for both types:

```
Π (IHl:pA 1), pA (cons tl 1)          Π (IHv:pB n 1), pB
  (S n) (consV n tl 1)
```

INDEX-IH then takes over. It substitutes the new motive in the inductive hypothesis, then recurses into both bodies, substituting the new inductive hypothesis for the index in the eliminator type for B . Here, it substitutes the new motive for p_A in the type of IH_l , extends the environment with IH_l , then substitutes IH_l for n , so that it recurses on these types:

```
pA (cons tl 1)                      pB (S IHl) (consV IHl tl 1)
```

Finally, INDEX-CONCLUSION computes the conclusion by taking the index of motive p_B at off $A B$, here $S IH_l$. In total, this produces a function that computes the length of `cons t 1`:

```
λ (tl:T) (l:list T) (IHl:(λ (l:list T).nat) 1).S IHl
```

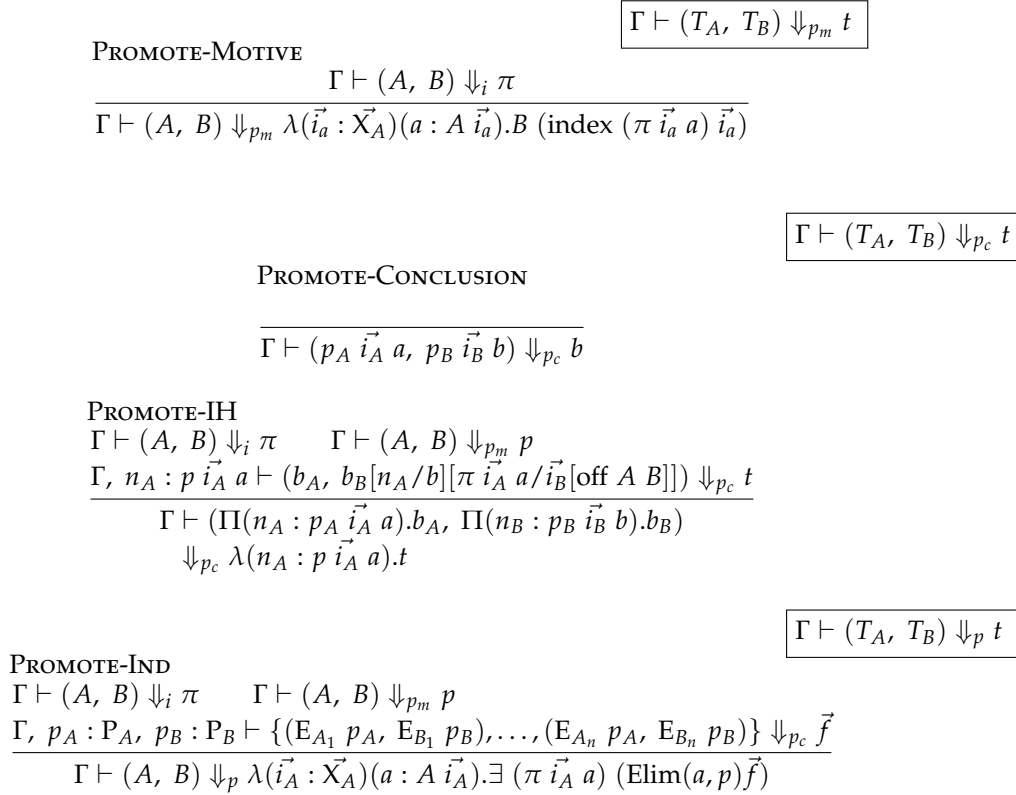



Figure 10: Identifying the promotion function.

Composing the Result. The $\Gamma \vdash (T_A, T_B) \Downarrow_i t$ judgment consists of only INDEX-IND, which identifies the motive and each case using the other two judgments, then composes the result. In the case of `list` and `vector`, this produces a function that computes the length of a list:

```
λ (l:list T).Elim(l, λ (l:list T).nat)
{0, λ (t_l:T) (l:list T) (IH_l:(λ (l:list T).nat) l).S IH_l}
```

4.3.1.2 Searching for Promote and Forget

Figure 10 shows the interesting derivations for the judgment $(T_A, T_B) \Downarrow_p t$ that searches for `promote`: PROMOTE-MOTIVE identifies the motive as B with a new index (which it computes using `indexer`, denoted by metavariable π). When PROMOTE-IH recurses, it substitutes the inductive hypothesis for the term rather than for its index, and it substitutes the new index (which it also computes using `indexer`) inside of that term. PROMOTE-CONCLUSION returns the entire term, rather than its index. Finally, PROMOTE-IND not only recurses into each case, but also packs the result.

The omitted derivations to search for `forget` are similar, except that the domain and range are switched. Consequentially, `indexer` is never needed; FORGET-MOTIVE removes the index rather than inserting it, and FORGET-IH no longer substitutes the index. Additionally,

FORGET-HYPOTHESIS adds the hypothesis for the new index rather than skipping it, and FORGET-IND eliminates over the projection rather than packing the result.

4.3.1.3 *Core Search Algorithm*

The core search algorithm produces `indexer`, `promote`, and `forget`, then composes them into a tuple. This tuple is how DEVOID represents ornaments internally. DEVOID has options (used in `Example.v`) that tell search to generate proofs that its outputs are correct, thereby increasing confidence in and usefulness of those outputs. The proof of coherence is reflexivity. The intuition behind the automation to prove section and retraction (`equivalence.ml`) is that `promote` and `forget` map along corresponding constructors, so inductive cases preserve equalities. Thus, each inductive case of these proofs is generated by a fold that rewrites each recursive reference, with reflexivity as identity.

4.3.2 *Other Search Procedures*

Brief explanation of these and how differencing works for them in detail based on algebraic ornaments

4.3.3 *Designing New Search Procedures*

How hard, how useful

Limitations and whether they're addressed in other tools yet or not

4.4 TRANSFORMATION

Parts of PUMPKIN Pi Transformation, with DEVOID 3.2 and 4.2 as examples, plus some of the beautiful Carlo theory to explain why we go from equivalences to configurations and what that really means

How the transformation works in detail

Limitations and whether they're addressed in other tools yet or not

4.5 IMPLEMENTATION

Parts of PUMPKIN Pi and DEVOID implementation, plus more

4.5.1 *Tool Details*

4.5.2 *Workflow Integration*

PUMPKIN Pi Decompiler and Implementation

4.6 RESULTS

PUMPKIN Pi Case Studies, key technical results

4.7 CONCLUSION

Rehashing thesis and how we do it

What we got here beyond what we had in PUMPKIN PATCH, segue into next chapter

5

RELATED WORK

5.1 PROGRAMS

Program Refactoring

Refactoring [58].

Program Repair

Adapting proofs to changes is essentially program repair for dependently typed languages. Program repair tools for languages with non-dependent type systems [65, 50, 47, 56, 62] may have applications in the context of a dependently typed language. Similarly, our work may have applications within program repair in these languages: Future applications of our approach may repurpose it to repair programs for functional languages.

Ornaments

DEVoid automates discovery of and lifting across algebraic ornaments in a higher-order dependently typed language. In the decade since the discovery of ornaments [54], there have been a number of formalizations and embedded implementations of ornaments [22, 43, 23, 44, 21]. DEVoid is the first tool for ornamentation to operate over a non-embedded dependently typed language. It essentially moves the automation-heavy approach of Ornamentation in ML [74], which operates on non-embedded ML code, into the type theory that forms the basis of theorem provers like Coq. In doing so, it takes advantage of the properties of algebraic ornaments [54]. It also introduces the first search algorithm to identify ornaments, which in the past was identified as a “gap” in the literature [44].

Programming by Example

Our approach generalizes an example that the programmer provides. This is similar to programming by example, a subfield of program

synthesis [36]. This field addresses different challenges in different logics, but may drive solutions to similar problems in a dependently typed language.

Differencing & Incremental Computation

Existing work in differencing and incremental computation may help improve our semantic differencing component. Type-directed differencing [61] finds differences in algebraic data types. Semantics-based change impact analysis [4] models semantic differences between documents. Differential assertion checking [46] analyzes different versions of a program for relative correctness with respect to a specification. Incremental λ -calculus [12] introduces a general model for program changes. All of these may be useful for improving semantic differencing.

5.2 PROOFS

Proof Reuse

Our approach reimagines the problem of proof reuse in the context of proof automation. While we focus on changes that occur over time, traditional proof reuse techniques can help improve our approach. Existing work in proof reuse focuses on transferring proofs between isomorphisms, either through extending the type system [7] or through an automatic method [53]. This is later generalized and implemented in Isabelle [38] and Coq [76, 71]; later methods can also handle implications. Integrating a transfer tactic with a proof patch finding tool will create an end-to-end tool that can both find patches and apply them automatically.

Proof reuse for extended inductive types [9] adapts proof obligations to structural changes in inductive types. Later work [63] proposes a method to generate proofs for new constructors. These approaches may be useful when extending the differencing component to handle structural changes. Existing work in theorem reuse and proof generalization [31, 66, 41] abstracts existing proofs for reusability, and may be useful for improving the abstraction component. Our work focuses on the components critical to searching for patches; these complementary approaches can drive improvements to the components.

DEVOID identifies and lifts proofs along a specific equivalence similar to that from existing ornaments work [44]. The need to automatically lift functions and proofs across equivalences and other relations is a long-standing challenge for proof engineers [52, 8, 51, 39, 77, 19]. The univalence axiom from Homotopy Type Theory [73] enables transparent transport of proofs; cubical type theory [18] gives univalence a constructive interpretation.

Similarly, our work is related to CoqEAL [19], which transfers functions along arbitrary relations between types. As these relations do not necessarily need to be equivalences, this framework is more general than our work. Similar tradeoffs between automation and generality apply: CoqEAL produces functions that refer to the old type, and does not yet support automatic inference of relations. In addition, CoqEAL currently only supports automatic transfer of functions, and does not yet handle proofs.

These tools may provide an alternative backend for DEVoid. Furthermore, our search algorithm may help discover relations that make these tools easier to use, and our lifting algorithm may help improve automation and efficiency for certain relations in these tools.

The problem that we solve is fundamentally about proof reuse, which applies software reuse principles to ITPs. There is a wealth of work in proof reuse, from tactic languages [32] and logical frameworks [13], to tools for proof abstraction and generalization [67, 42], to domain-specific methodologies [24] and frameworks [25].

DEVoid focuses on the specific problem of reuse when adding fully-determined indices to types. Other approaches to this problem include combinators which definitionally reduce to desirable terms [30] in the language Cedille, and automatic generation of conversion functions in Ghostbuster [55] for GADTs in Haskell. Our work focuses on a type theory different from both of these, in which the properties that allow for such combinators in Cedille are not present, and in which dependent types introduce challenges not present in Haskell.

DEVoid is not the first tool to combine search with reuse. Optician [60] synthesizes bidirectional string transformations; a similar approach may help extend tooling to handle transformations for low-level data. PUMPKIN PATCH [69] searches the difference in proofs for patches that can be used to repair proofs broken by changes; DEVoid uses a similar approach to identify functions that form an equivalence. The resulting tools are complementary: DEVoid supports the addition of indices and hypotheses, which PUMPKIN PATCH does not support; PUMPKIN PATCH supports changes in values, which DEVoid does not support.

Proof Evolution

There is a small body of work on change and dependency management for verification, both to evaluate impact of potential changes and maximize reuse [40, 3] and to optimize build performance [14]. These approaches may help isolate changes, which is necessary to identify future benchmarks, integrate with CI systems, and fully support version updates.

*Proof Refactoring**Proof Repair**Proof Design*

Existing proof engineering work addresses brittleness by planning for changes [75] and designing theorems and proofs that make maintenance less of an issue. Design principles for specific domains (such as formal metatheory [5, 26, 27]) can make verification more tractable. CertiKOS [35] introduces the idea of a deep specification to ease verification of large systems. These design principles and frameworks are complementary to our approach. Even when programmers use informed design principles, changes outside of the programmer’s control can break proofs; our approach addresses these changes.

Proof Automation

We address a missed opportunity in proof automation for ITP: searching for patches that can fix broken proofs. This is complementary to existing automation techniques. Nonetheless, there is a wealth of work in proof automation that makes proofs more resilient to change. Powerful tactics like *crush* [16] can make proofs more resilient to changes. Hammers like Isabelle’s *sledgehammer* [64] can make proofs agnostic to some low-level changes. Recent work [20] paves the way for a hammer in Coq. Even the most powerful tactics cannot address all changes; our hope is to open more possibilities for automation.

Powerful project-specific tactics [16, 15] can help prevent low-level maintenance tasks. Writing these tactics requires good engineering [34] and domain-specific knowledge, and these tactics still sometimes break in the face of change. A future patching tool may be able to repair tactics; the debugging process for adapting a tactic is not too dissimilar to providing an example to a tool.

Rippling [11] is a technique for automating inductive proofs that uses restricted rewrite rules to guide the inductive hypothesis toward the conclusion; this may guide improvements to the differencing, abstraction, and specialization components. The abstraction and factoring components address specific classes of unification problems; recent developments to higher-order unification [59] may help improve these components. Lean [70] introduces the first congruence closure algorithm for dependent type theory that relies only on the Uniqueness of Identity Proofs (UIP) axiom. While UIP is not fundamental to Coq, it is frequently assumed as an axiom; when it is, it may be tractable to use a similar algorithm to improve the tool.

GALILEO [10] repairs faulty physics theories in the context of a classical higher-order logic (HOL); there is preliminary work extend-

ing this style of repair to mathematical proofs. Knowledge-sharing methods [33] can adapt some proofs across different representations of HOL. These complementary approaches may guide extensions to support decidable domains and classical logics.

6

CONCLUSIONS & FUTURE WORK

Reflect on thesis statement and explain how we got it exactly now that you know everything

But I want to spend the rest of this thesis talking about the next era of verification so I can write out a bunch of ideas for students who might want to work with me

THE NEXT ERA: PROOF ENGINEERING FOR ALL

Future Work from many papers, plus research statement, DARPA thoughts, plus more, but trimmed down a lot

What I want in the long run, how this all fits in, is a world of proof engineering for all. From research statement, three rings (four including experts in the center).

And what we have so far with my thesis is a world where it's easier for experts and a bit easier for practitioners, but there's still a lot left to go building on it.

So here are 12 short future project summaries that reach each of these tiers, building that world. Super please contact me if any of these seem fun to you.

Proof Engineering for Experts

Unifying theme: lateral reach. Some examples:

MORE PROOF ASSISTANTS Thoughts from PUMPKIN Pi on Isabelle/HOL, future work from PUMPKIN PATCH.

MORE CHANGES Version updates, isolating large changes (PUMPKIN PATCH), relations more general than equivalences (PUMPKIN Pi).

MORE STYLES ML for decompiler (PUMPKIN Pi, REPLICA) : more for diverse proof styles (PUMPKIN PATCH). Note that this is a WIP, but sketch out project, challenges, future ideas, expectations, evaluation a bit.

Proof Engineering for Practitioners

Unifying theme: usability. Some examples:

AUTOMATION More search procedures for automatic configuration, e-graphs from PUMPKIN Pi, custom unification heuristics.

INTEGRATION IDE & CI integration, HCI for repair.

EVALUATION repair challenge, user studies ideas (PUMPKIN PATCH, REPLICA, panel w/ Benjamin Pierce, QED at large). (maybe look for more ideas, this can be merged with integration if need be).

Proof Engineering for Software Engineers

Unifying theme: mixed methods verification, or the 2030 vision from Twitter thread. Some examples:

GRADUAL VERIFICATION A continuum from testing to verification, tools to help with that.

TOOL-ASSISTED PROOF DEVELOPMENT Tool-assisted development to follow good design principles for verification (James Wilcox conversation, final REPLICA takeaway).

SPECIFICATION INFERENCE Analysis to infer specs (TA1).

Proof Engineering for New Domains

Unifying theme: collaboration, new abstractions for new domains). Some examples:

MACHINE LEARNING Fairification & other ML correctness properties. Some stuff here but more.

CRYPTOGRAPHY Lots of stuff here but not thinking broadly enough. What about cryptographic proof systems? ZK and beyond. Recall email thread.

SOMETHING ELSE Look for more in survey paper, email, DARPA TAs, Twitter. Healthcare perhaps?

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