

Proofs-as-programs for programmers

Pierre Corbineau and Jean-François MONIN



- Target audience
- Challenges
- Conventional order
- Contents
- Unusual design decisions

Practical motivation

Understand better software pieces

- principled reasoning on programs
- theoretical culture on semantics and notions of computation
- (certified) compilation

But

- Don't expect anything about maths
don't like, hate, are afraid of, not confident
- Don't teach math, but some skills in mathematical thinking
- Science (including programming) = mixture of rigour and creativity

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Coq'Art, SF, author's first courses...

- Propositional logic; first order quantifiers
- Basic data: `bool`, `nat`; simple (recursive) programs
- Induction on `nat`
- Lists (monomorphic, polymorphic); tree-like data-structures
- “Rich” data-types (e.g., `sig` types)
- Inductive relations
- Applications to
 - maths
 - Computer Science: program correctness
 - Computer Science: semantics
 - ...
- Exotic topics, e.g., Curry-Howard correspondence

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Conventional order, assessment

Good points

- simple \rightarrow complex
- theory \rightarrow applications

Drawbacks

- slow: interesting things come late
- mysterious interactions

Possible repair: make things closer to usual math

- different interface
- hide more
- use more automation

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Contract: don't cheat

We don't want students to cheat

We should not cheat as well (as much as possible)

The more explicit and transparent, the better

Use automation only when what happens is perfectly understood

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Math-oriented people

- Bizarre notations (e.g., function application)
- Interesting theories come late
- Why bother with Curry-Howard?

Computer-oriented people

- Requires too much patience
- Amazing and funny things come late

Moreover, Coq's feedback sometimes hard to understand

Two languages: formulae and scripts

Formulae

More or less familiar

Tactics

- Collection of receipes
you face this typical situation, use that tactic.
- Alternative: a little bit of proof-theory?
CIC is complex and subtle, cannot be a starting point.

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Alternative: go fast

But carefully!

Expose interesting / challenging examples as soon as possible

RELY ON PROGRAMMERS INTUITIONS

- trees, easy to visualize
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RELY ON PROGRAMMERS INTUITIONS

- trees, easy to visualize
- (structural) recursive programs on them

- Easy and common data structure
- Abstract Syntax Trees
- Proof trees
structured thinking
- Typing rules
- (Structured, big step, small step) Operational Semantics
- All kinds of inductive definitions

This is consistent with Type Theory

Focus on functional programming,
not on proof-theory

Thank you Curry-Howard

- Implication and universal quantification seen as types for functional programs
- Structural induction seen in the same way as structural recursion
- Natural explanation of implication (\gg truth tables)

Basic types

- `bool` is not that good: interferences with logic
→ take another enumerated type (e.g., traffic colors)
- binary trees and ASTs for simple arithmetic expressions

Functions

- curried syntax
- pattern-matching

Simple theorems and basic tactics

- equalities: `reflexivity` and `rewrite`
- `introduction` of hypotheses or variables, computation steps (`cbn`)
- reasoning by case: `destruct`
- induction on trees

First steps

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Interactive development of functions

Proofs of implicational / universal theorems are functions

Reasoning by case = programming by case

Deep patterns can be useful (with `refine`)

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Deep patterns can be useful (with **refine**)

Basic example: traffic colors

```
Inductive tcolor : Set :=  
  | Green  : tcolor  
  | Orange : tcolor  
  | Red    : tcolor.
```

```
Definition next_col : tcolor -> tcolor :=  
  fun c =>  
    match c with  
    | Green  => Orange  
    | Orange => Red  
    | Red    => Green  
  end.
```

```
Lemma nextnextnext_id :  
  forall c:tcolor, next_col (next_col (next_col c)) = c.
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Challenging example 1, on binary trees

```
Fixpoint revt t : bintree :=
  match t with
  | L c   => L c
  | N l r => N (revt r) (revt l)
  end.
```

Theorem revt_revt : forall t, revt (revt t) = t.

- (for teachers) it is cheap!
much simpler as the corresponding result on lists
- (for students)
try the same exercise in your favorite imperative/OO progr. lang.

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Funny example: functions returning a type, dependent types

Definition `waow` : `tlcolor` -> `Set` :=

```
fun c =>
  match c return Set with
  | Green => nat
  | Orange => tlcolor
  | Red => tlcolor -> nat
end.
```

Definition `waow_waow` : forall c : `tlcolor`, `waow` c :=

```
fun c =>
  match c return waow c with
  | Green => 2
  | Orange => Green
  | Red => fun c' => match c' with Orange => 6 | _ => 1 end
end.
```

Predicate = dependent type, Curry-Howard at work

Lemma nextnextnext_id :

```
forall c:tlcolor, next_col (next_col (next_col c)) = c.
```

Proof.

```
refine (fun c => _).
refine (match c with      (* Reasoning by case ([destruct c]) *)
  | Green => _
  | Orange => _
  | Red => _
end).
all:exact refl_equal.
```

Qed.

Definition id_nextnextnext_pgm :

```
forall c, c = next_col (next_col (next_col c)) :=
fun c => match c with
  | Green => eq_refl
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Challenging example 2: code optimization

```
Inductive aexp : Set :=  
| Cst : nat -> aexp  
| Apl : aexp -> aexp -> aexp  
| Amu : aexp -> aexp -> aexp.
```

```
Fixpoint eval (a : aexp) : nat :=
```

Challenging example 2 (cont'd): syntactic simplification on ASTs

simpl0 Apl
 / \ = a2 (* 0 + a2 = a2 *)
 Cst a2
 |
 0

simpl0 Amu Cst
 / \ = | (* 0 * a2 = 0 *)
 Cst a2 0
 |
 0

simpl0 a = a in all other cases

Challenging example 2 (cont'd): correctness of the optimization

Use ad-hoc case analysis

Lemma `eval_simpl0`: forall a, eval (simpl0 a) = eval a.

Proof.

```
intro a.  
refine ( match a with  
  | Apl (Cst 0) a2 => _  
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  | a'           => eq_refl (eval a')  
end ).
```

Fixpoint `simpl_rec` (a : aexp) :=

Lemma `eval_simpl_rec`: forall a, eval (simpl_rec a) = eval a.

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“Absurd” hypotheses → contagious equalities

Explaining `discriminate` without black magic (nor large eliminations to some extent).

```
Lemma absurd: 5 = 4 -> 15 = 12.
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```
Lemma true_false_eq : true = false -> forall n1 n2 : nat, n1 = n2.
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Lemma eq_eqnatb : forall n1 n2, eqnatb n1 n2 = true -> n1 = n2.
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Then we can get `discriminate` using large eliminations, seen earlier in the `waow` function.

Next step: (small) inversion

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- **constructive / classical logic**
- negation
 - needed lemmas happen to be stated in a positive way
 - deal with “absurd” hypotheses in a positive way as well
- conjunction, disjunction and existential quantifier
(not really needed)
- truth
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Because

- equalities, at the beginning
- custom inductive types, later

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Conclusions

7 years at M1 level (engineering school, ~ 50 students/year)

Feedback was never bad and has improved to "reasonably good" in the recent years.

Show simple amazing things

We tried to introduce complex notions on simple examples.

Rely on (functional) programmers intuition

Yet another reason to teach functional programming to scientists.

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There is room for them next to numbers and other mathematical notions.

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