

Experiments towards efficient computations

in formalized linear algebra

Maxime Dénès

Marelle Team, INRIA Sophia-Antipolis
ForMath European Project

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- 1 Introduction
- 2 General scheme
- 3 Native compilation
- 4 Primitive structures
- 5 Conclusion

General objective:

Efficient computations in linear algebra proofs.

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Efficient computations in linear algebra proofs.

Motivations

- Proofs relying on (concrete) heavy computations
- Usability of certified algorithms inside provers

Context

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- Calculus of Inductive Constructions

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- Reduction rules encode computations

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- Conversion rule embeds computational steps in reasoning

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Conversion rule:

$$\frac{\Gamma \vdash U : s \quad \Gamma \vdash t : T \quad \Gamma \vdash T \equiv_{\beta\delta\iota\zeta} U}{\Gamma \vdash t : U} \text{conv}$$

In Coq, computational steps are often plugged in proofs through reflection.

$$\Gamma \vdash \text{fP} : \forall(x : T), f(x) = \text{true} \rightarrow P(x)$$

$$\Gamma, x : T \vdash \text{fP}_x(\text{refl_equal true}) : P(x)$$

We distinguish small scale reflection:

```
Lemma existsP : forall P,  
  reflect (exists x, P x) (existsb x, P x).
```

which enables computations in small steps, with fine grained control

and large scale reflection:

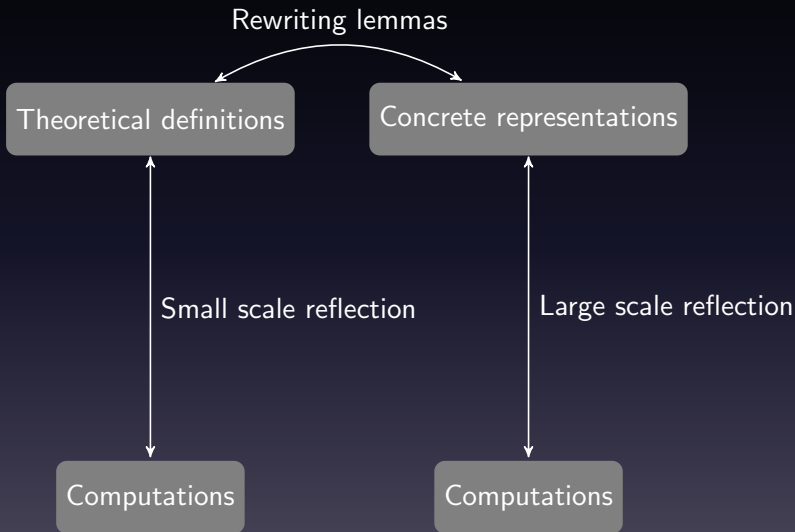
```
Lemma cfred232: cfreducible (Config 11 33 37 H 2 H 13
Y 5 H 10 H 1 H 1 Y 3 H 11 Y 4 H 9 H 1 Y 3 H 9 Y 6 Y 1
Y 1 Y 3 Y 1 Y Y 1 Y).
```

```
Proof. apply check_reducible_valid; by compute. Qed.
```

We would like to perform large scale reflection on objects from linear algebra (matrix,...).

The SSReflect extension (for the Coq proof assistant) provides powerful tools and theories on these objects, but not full computational features.

- 1 Introduction
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We need:

- Efficient techniques for evaluation
- Concrete and adapted representations
- Connections between abstract and concrete descriptions

Example: addition of matrices

```
Definition comp_matrix_add (M N:comp_matrix) :=
  mapi (fun i u => comp_row_add u (get N i)) M.
```

```
Lemma comp_matrix_addE: forall (M N:comp_matrix),
  (matrix_of_comp_matrix M) + (matrix_of_comp_matrix N)
  (matrix_of_comp_matrix (comp_matrix_add M N)).
```

Example: addition of matrices

Goal $m1 + (m2 + m3) = m3 + (m2 + m1)$.

Proof.

```
rewrite /m1 /m2 /m3.
```

```
rewrite !comp_matrix_of_funE !comp_matrix_addE.
```

```
by apply:comp_matrix_eqP ; compute.
```

```
Qed.
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- 1 Introduction
- 2 General scheme
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Large scale reflection requires efficient techniques for evaluation of terms, that scale up well to heavy computations.

Weak reduction

Used in most functional languages.

```
# let rec f x = f x;;  
  val f : 'a -> 'b = <fun>  
# f;;  
  - : 'a -> 'b = <fun>  
# f 2;;
```

Weak reduction

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```
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  val f : 'a -> 'b = <fun>
# f;;
  - : 'a -> 'b = <fun>
# f 2;;
```

Strong reduction

Enables comparison of functions body.

```
Coq < Eval compute in (fun (x:nat) => (fun y => y) x).
  = fun x : nat => x
  : nat -> nat
```

How to implement strong reduction ?

Main issue

Which techniques for strong reduction ?

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Abstract machine manipulating a representation of terms

Example: standard evaluator of Coq (Barras, 1999)

- Uses a lazy strategy
- Enables some control over heuristics
- No specialization
- Limited performance
- Better suited to small computations

Bytecode compilation (Grégoire and Leroy, 2002)

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- Uses a modified version of OCaml's virtual machine
- Compiles terms to bytecode

Native code compilation

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Limits of this approach

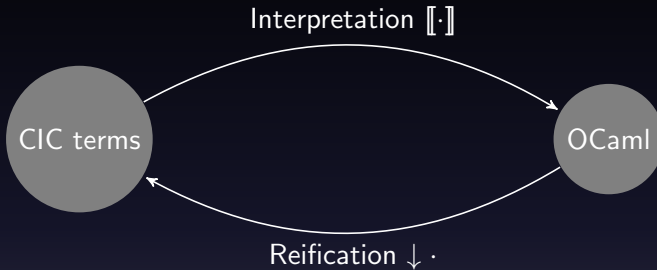
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- Same framework as Coq's VM
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Limits of this approach

- Well suited only for full evaluation involving heavy computations
- No control over the heuristics of the underlying evaluator



Where :

$$\forall t \forall u, t \equiv_{\beta\delta\iota\zeta} u \implies [[t]] = [[u]]$$

$$\forall t \text{ in normal form, } \downarrow [[t]] = t$$

Representation of terms

$$\begin{aligned} \text{Var} &\ni \underline{x} \\ \text{Term} &\ni \underline{t} ::= \underline{x} \mid \underline{t_1} \underline{t_2} \mid \underline{v} \\ \text{Val} &\ni \underline{v} ::= \underline{\lambda x. t} \mid [\tilde{x} \ v_1 \ \dots \ v_n] \end{aligned}$$

Weak reduction of symbolic calculus

$$\begin{aligned} (\underline{\lambda x. t}) \underline{v} &\rightarrow_{\beta_v} \underline{t}[x \leftarrow v] && (\beta_v) \\ [\tilde{x} \ v_1 \ \dots \ v_n] \underline{v} &\rightarrow_{\beta_s} [\tilde{x} \ v_1 \ \dots \ v_n \ v] && (\beta_s) \end{aligned}$$

Strong reduction by iteration of weak reduction

$$\mathcal{N}(t) = \downarrow_0 \llbracket t \rrbracket_0$$

$$\downarrow_n \underline{\lambda x. t} = \lambda \hat{n}. \downarrow_{n+1} ((\underline{\lambda x. t}) [\tilde{n}])$$

$$\downarrow_n [\tilde{x} v_1 \dots v_n] = (n - x) (\downarrow_n \underline{v_1}) \dots (\downarrow_n \underline{v_n})$$

Key ideas of the implementation

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- Representation of accumulators requires to manipulate OCaml's internal objects.
- Objective: encode β_S and β_V rules by OCaml's β -réduction.

Representation of accumulators

Usual OCaml's closures look like the following:



We just change the tag and adapt the code zone, so that the arguments accumulate.



Interpretation of terms of λ -calculus

$$\begin{aligned} \llbracket t_1 t_2 \rrbracket_n &= \llbracket t_1 \rrbracket_n \llbracket t_2 \rrbracket_n \\ \llbracket x \rrbracket_n &= \widehat{n - x} \\ \llbracket \lambda x. t \rrbracket_n &= \text{fun } \widehat{n} \rightarrow \llbracket t \rrbracket_{n+1} \end{aligned}$$

Then we extend the calculus with inductive types, fixpoints, sorts, . . .

- 1 Introduction
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For the sake of efficiency, large scale reflection requires primitive structures natively interpreted in the underlying language (e.g. integers, arrays).

```
Register int : Set as int31_type.  
Register add : int -> int -> int as int31_add.
```

```
[[add]] = fun x y → if (is_int x) && (is_int y)  
  then Int31.add x y else mk_id_accu "add" x y
```

Examples of applications :

- Primality tests
- SAT prover trace verification (Armand et al., 2010)
- Computations on concrete matrices (arrays)

Native arrays

- Seen inside Coq as usual functional arrays

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- Interpreted by persistent arrays, based on OCaml's imperative features (Conchon and Filliâtre, 2007)

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- Seen inside Coq as usual functional arrays
- Interpreted by persistent arrays, based on OCaml's imperative features (Conchon and Filliâtre, 2007)
- Provided primitive operations (fold, map, ...)

Back to our matrices example

```
Definition comp_matrix := array (array T).
```

```
Definition comp_matrix_add (M N:comp_matrix) :=  
map_i (fun i u => comp_row_add u (get N i)) M.
```

```
Goal m1 + (m2 + m3) = m3 + (m2 + m1).
```

```
Proof.
```

```
rewrite /m1 /m2 /m3.
```

```
rewrite !comp_matrix_of_funE !comp_matrix_addE.
```

```
by apply:comp_matrix_eqP ; compute.
```

```
Qed.
```



```
Goal m1 + (m2 + m3) = m3 + (m2 + m1).
```

```
Proof.
```

```
rewrite /m1 /m2 /m3.
```

```
rewrite !comp_matrix_of_funE !comp_matrix_addE.
```

```
by apply:comp_matrix_eqP ; native_compute.
```

```
Qed.
```

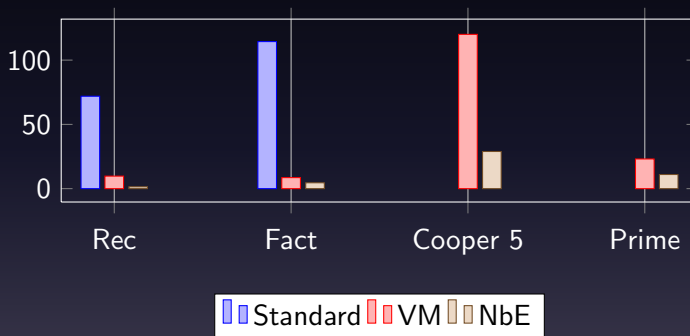
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- 2 General scheme
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A few benchmarks

	Rec	Fact
Standard	71,89s (100 %)	114,4s (100 %)
VM Coq	9,75s (13,6 %)	8,58s (7,5 %)
NbE	1,46s (2,0 %)	4,56s (3,99 %)

	Cooper 5	Prime
Standard	untested	untested
VM Coq	120,03s (100 %)	23.08s (100 %)
NbE	28,9s (24 %)	10.99s (48 %)

A few benchmarks



What is (almost) done

- Native compilation of CIC terms
- Primitive integers and arrays in Coq's terms
- Compilation to native OCaml integers

Work in progress

- Universe constraints, coinductives and cofixpoints
- Compilation to native OCaml arrays
- Concrete implementation of SSReflect matrices and their operations

Perspectives

- Fully computable formalization of standard linear algebra algorithms
- Linear algebra proofs relying on heavy computations (suggestions welcome)