

Mechanized Verification of the Correctness and Asymptotic Complexity of Programs

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under the supervision of Arthur Charguéraud and François Pottier

Computer programs: cooking recipes, but for computers?

Mom's easy apple pie

- Slice 6 apples
- Mix with 3/4C sugar, 2T flour, 3/4T cinnamon, 1T lemon juice
- Transfer between two pie crusts
- Bake 40 min at 425°F

Computing the lengths of two lists

```
let length_sum l1 l2 =  
  let x = length l1 in  
  let y = length l2 in  
  x + y
```


Computer: cooking recipes, but for computers? (2)

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What does it mean to do the right thing?

“The right thing”: a **specification**, written in a formal language.

What do we expect from a program?

less
confidence



more
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(does not crash)



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An illustrative example: Binary Search

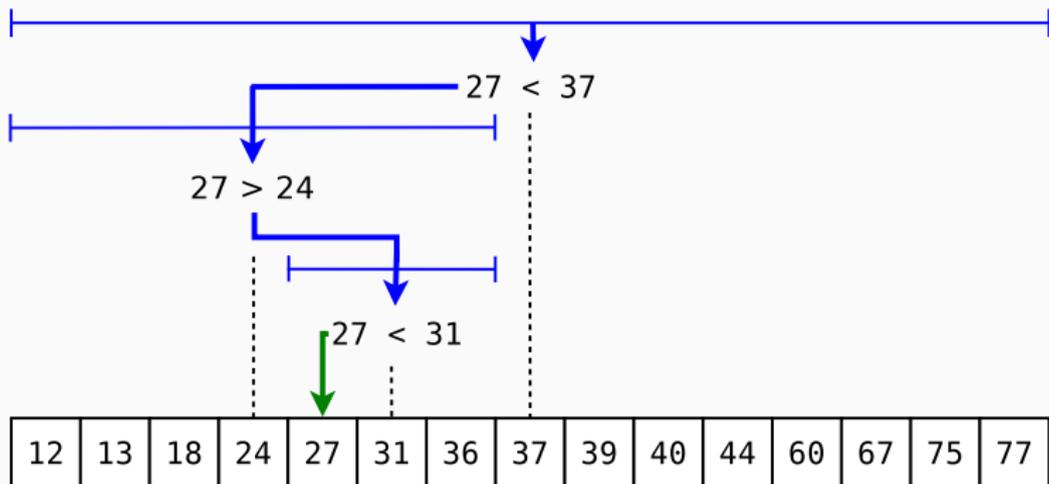
Consider a *sorted* array of integers:

12	13	18	24	27	31	36	37	39	40	44	60	67	75	77
----	----	----	----	----	----	----	----	----	----	----	----	----	----	----

Question: is 27 in the array? If so, at which index?

An illustrative example: Binary Search (2)

At each step, reduce by half the segment to search by comparing 27 with the middle element.



A tentative binary search implementation

```
(* search in array a for x, in the range [i, j) *)  
(* returns the index of x, or -1 if not found *)  
let rec bsearch (a: int array) x i j =  
  if j <= i then -1 else  
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    if x = a.(k) then k  
    else if x < a.(k) then bsearch a x i k  
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- We can test this program on example input data
- We can formally prove its (total) functional correctness

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- We can test this program on example input data
- We can formally prove its (total) functional correctness
- Yet, something is wrong...

A tentative binary search implementation (2)

On an array containing 1 billion elements:

- A correct binary search should do at most 30 recursive calls ($2^{30} \simeq 1$ billion)
- On some inputs, the code shown performs 1 billion recursive calls

A tentative binary search implementation (3)

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(* search in array a for x, in the range [i, j] *)  
(* returns the index of x, or -1 if not found *)  
let rec bsearch (a: int array) x i j =  
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    if x = a.(k) then k  
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```

buggy, should be k+1

A tentative binary search implementation (4)

In summary, on an array of size n :

- We expect $O(\log n)$ recursive calls;
- But our program does up to n recursive calls.

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Partial correctness	(returns a correct result; might not terminate)
Total correctness	(always returns a correct result)
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Formal verification of correctness and complexity of a program

Step 1

State a **program specification** that characterizes the intended behavior:
functional correctness **and** runtime complexity

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Prove a theorem relating concrete code to the specification

Two kinds of possible human mistakes:

- in math results used in the analysis; or
- when relating the concrete code to the abstract algorithm

Use a **proof assistant** (Coq) to mechanically check every step of the proof

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Option 1: as an upper bound on the wall-clock time.

Useful for embedded systems, but not realistic for commodity hardware.

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Option 2: as a number of cycles for an idealized machine model.



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“Merge sort runs in $10N \log N + 4.92N$. [This bound] can be reduced to $9N \log N$ at the expense of a somewhat longer program.”

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Option 3: as a number of function calls in a high-level language.

More abstract, but still has modularity issues.

How do we specify a program's running time?

Option 4: specify the running time using **asymptotic** complexity.

Describe the “order of growth” of the running time as inputs grow large
e.g. $O(\log n)$, $O(n)$, $O(n \log n)$, $O(n^2)$,

Less precise, but informative enough in many cases.

Advantages of asymptotic complexity specifications

Specifications capturing asymptotic costs:

- have been **widely applied** to a large class of programs and algorithms;
- are **independent** of the machine, runtime system and the details of the implementation;
- allow **modular reasoning**. Abstract over implementation details.

In this thesis

Goal: specify and prove that programs compute a correct result with a bounded asymptotic runtime.

Proofs should be:

- static;
- machine-checked;
- hardware- and runtime- independent;
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Contribution:

A step forward for the verification of the **correctness and complexity** of **imperative, higher-order** programs with **subtle invariants and analysis**, at a **reasonable cost**.

Details of the contribution

1. A formal account of $O()$

Existing:

single-variate O (math, programs), multi-variate O on paper

Contributed:

Coq library for single and multi-variate O ,
with lemmas useful for program analysis

2. A methodology for complexity proofs

Existing:

- manual verification without $O()$ abstraction
- automated analysis restricted to polynomial bounds

Contributed:

- general asymptotic bounds
- with semi-automated cost inference
- implemented as an extension of CFML
(Separation Logic framework in Coq)

3. Case studies

Existing:

polynomial or logarithmic bounds, simple algorithms (quicksort), or interactive verification without O

Contributed:

several algorithms, including a state-of-the-art graph algorithm with nontrivial correctness and complexity

Outline of the rest of the talk

Reasoning with abstract cost functions

Semi-automatic inference of cost functions

Separation Logic with Time Credits

Case study—an Incremental Cycle Detection Algorithm

Reasoning with abstract cost functions

Informal reasoning principles on O can be abused

```
1 let rec bsearch a x i j =  
2   if j <= i then -1 else  
3     let k = i + (j - i) / 2 in  
4     if x = a.(k) then k  
5     else if x < a.(k) then  
6       bsearch a x i k  
7     else  
8       bsearch a x (k+1) j
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Claim:

bsearch a x i j costs $O(1)$.

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By induction on $j - i$:

- $j - i \leq 0$: $O(1)$.

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Where is the catch?

- $j - i \leq 0$: $O(1)$.
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Claim:

bsearch a x i j costs $O(1)$.

Proof:

By induction on $j - i$: ...but which statement are we proving?

- $j - i \leq 0$: $O(1)$.
- $j - i > 0$: $O(1) + O(1) + O(1) = O(1)$.

Meaning of $O(1)$

What we just proved:

$\forall i, j$, $\exists c$, “bsearch a x i j” performs at most c function calls

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$\forall i j$, $\exists c$, “bsearch a x i j” performs at most c function calls

What “ $O(1)$ ” means:

$\exists c$, $\forall i j$, “bsearch a x i j” performs at most c function calls

Meaning of $O(\log n)$

Informal specification: “bsearch a x i j” runs in $O(\log(j - i))$.

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Meaning: there exists a **cost function** f such that,

- for every a, x, i, j , “bsearch a x i j” performs at most $f(j - i)$ function calls
- $f \in O(\lambda n. \log n)$.

Construction of the cost function

Option 1: The user somehow guesses a suitable cost function.
Here, " $\lambda n. 3 \log n + 4$ " works.

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Option 2: Semi-automatically construct the cost function as the proof progresses.

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Semi-automatic synthesis of cost functions

Our approach to this problem

Part 1:

- Synthesize a cost function with the same structure as the code
- For recursive functions, recurrence equations are synthesized
- Accounting details are automatically synthesized
- User input is requested when some over-approximation is required

Part 2:

- In a second step, prove a $O()$ bound for the inferred cost function

Constraint inferred on the cost function f

```
let rec bsearch a x i j =
  if j <= i then -1 else
    let k = i + (j - i) / 2 in
    if x = Array.get a k then k
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```

```
f n >= 1 + (                                     where n = j-i
  if n <= 0 then 0 else
    0 + 1 + max 0 (
      1 + max (f (n/2))
              (f (n - n/2 - 1))
    )
)
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Interactive construction of the cost function f

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if j <= i then -1 else
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$f(j-i) \geq 1 + \dots$

a hole (“...”) is implemented as an evar in Coq

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f (j-i) >= 1 + (if j <= i then ... else ...)
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f (j-i) >= 1 + (
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From cost equation to asymptotic bound

For bsearch, there remains to find a $f \in O(\lambda n \cdot \log n)$ such that:

$$\forall n. f(n) \geq 1 + \begin{cases} 0 & \text{if } n \leq 0 \\ 1 + \max(0, 1 + \max(f(\frac{n}{2}), f(n - \frac{n}{2} - 1))) & \end{cases}$$

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- Use the “Master Theorem”, when applicable (available in Isabelle/HOL, not yet in Coq)
- Substitution method: guess that there is a solution of the form $a \log n + b$, inject it and resolve.

The substitution method in action

$$\exists f : \mathbb{Z} \rightarrow \mathbb{Z}.$$

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monotonic f

$$\wedge \forall n. f(n) \geq 0$$

$$\wedge \forall n. n \leq 0 \implies f(n) \geq 1$$

$$\wedge \forall n. n \geq 1 \implies f(n) \geq f\left(\frac{n}{2}\right) + 3$$

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The substitution method in action

$\exists a b : \mathbb{Z}.$

$$f(n) = a \log n + b$$

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The substitution method in action

$\exists a b c : \mathbb{Z}.$

$f(n) = \text{if } n > 0 \text{ then } a \log n + b \text{ else } c$

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$\wedge c \geq 1$

$\wedge b \geq c + 3 \wedge a \geq 3$

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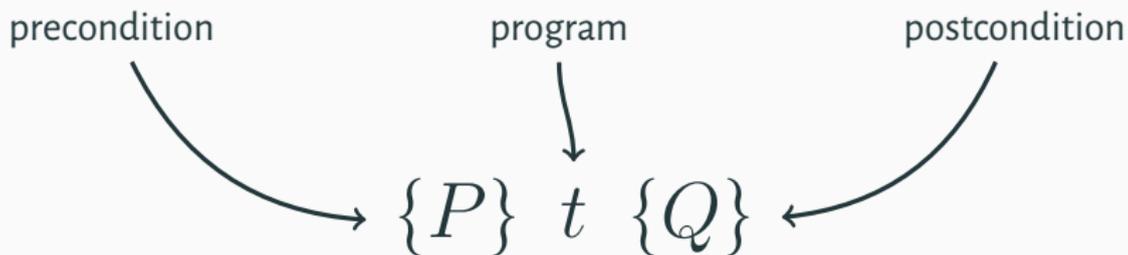
Can be solved automatically.

The user does not have to manually provide values for a , b , and c .

Separation Logic with Time Credits

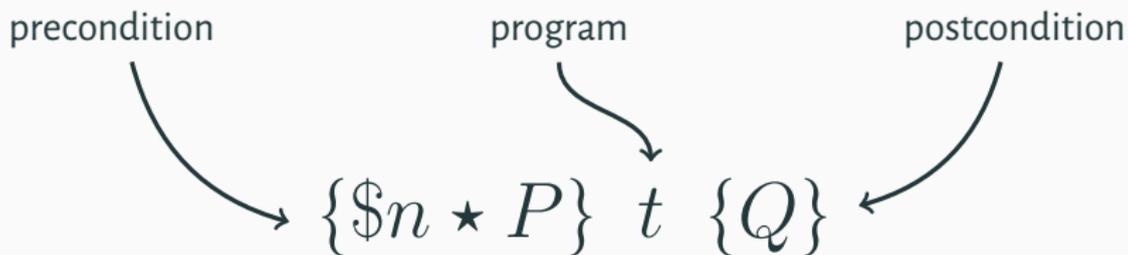
Linking code to cost assertions

Program specifications using Separation Logic



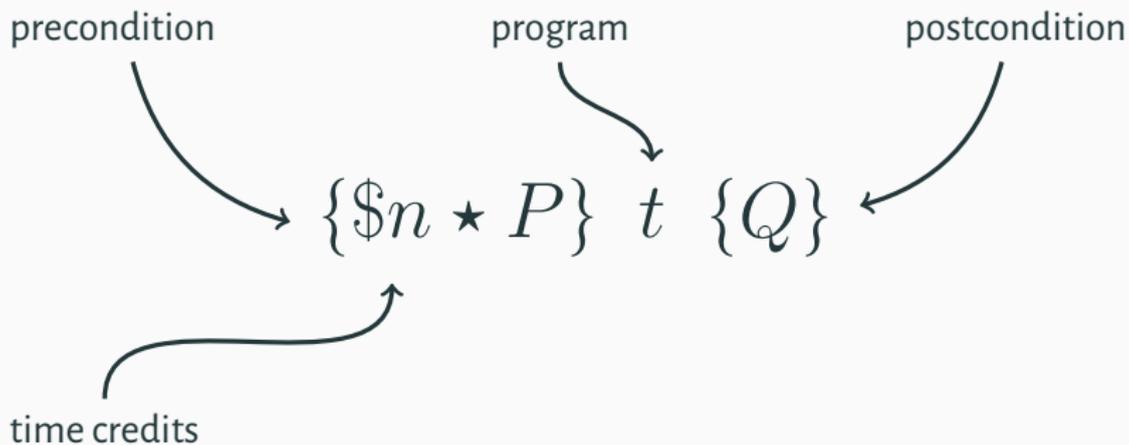
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Program specifications using Separation Logic with Time Credits



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Program specifications using Separation Logic with Time Credits



Time Credits: resources in separation logic

$\$n$

- $\$n$ describes the right to perform n function calls or loop iterations
- $\$(n + m) = \$n \star \$m$
- $\$0 = \text{emp}$

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- $\$n$ describes the right to perform n function calls or loop iterations
- $\$(n + m) = \$n \star \$m$
- $\$0 = \text{emp}$
- Credits are not duplicable: $\$1 \not\Rightarrow \$1 \star \$1$
- Enable amortized complexity analysis

Using time credits in the specification of bsearch

Specification of the complexity of bsearch using time credits:

$$\begin{aligned} & \exists f : \mathbb{Z} \rightarrow \mathbb{Z}. \\ & \left\{ \begin{array}{l} f \in O(\lambda n. \log n) \\ \forall a \ x \ i \ j. \{ \$ (f(j - i)) \star \dots \} (\text{bsearch } a \ x \ i \ j) \{ \dots \} \end{array} \right. \end{aligned}$$

Contribution: Possibly Negative Time Credits

Separation Logic with Time Credits in \mathbb{N} :

$$\begin{aligned} \$0 &\equiv \text{emp} \\ \forall m n \in \mathbb{N}. \quad \$(m + n) &\equiv \$m \star \$n \\ \forall n \in \mathbb{N}. \quad \$n &\Vdash \text{emp} \end{aligned}$$

My extension: Possibly Negative Time Credits in \mathbb{Z} :

$$\begin{aligned} \$0 &\equiv \text{emp} \\ \forall m n \in \mathbb{Z}. \quad \$(m + n) &\equiv \$m \star \$n \\ \forall n \in \mathbb{Z}. \quad \$n \star [n \geq 0] &\Vdash \text{emp} \end{aligned}$$

Corollary: $\$n \equiv \$m \star \$(n - m)$

Possibly Negative Time Credits enable simpler specifications

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  (* returns the index of the first occurrence of v in a *)
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Time Credits in \mathbb{Z} : benefits

- **Simpler specifications**
(when the cost depends on the result)
- **Significant reduction of the number of intermediate side-conditions**
(can accumulate debts and pay them off once at the end)
- **Simpler loop invariants**
(no need to justify that a number of credits is positive at each step)

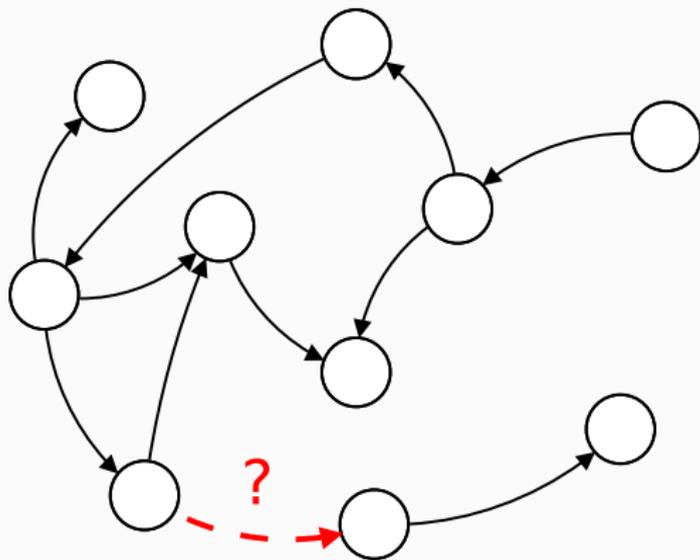
Case Study: an Incremental Cycle Detection Algorithm

Our main case study

Verification of a state-of-the-art **incremental cycle detection algorithm** due to Bender, Fineman, Gilbert and Tarjan (2016).

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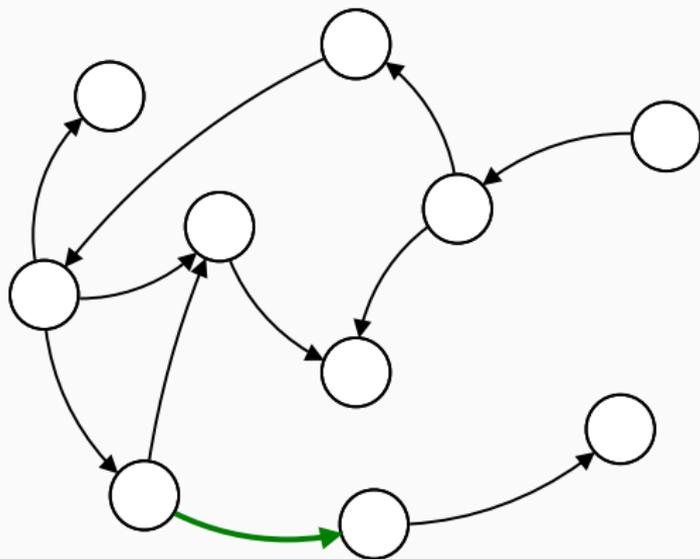
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The problem: checking for acyclicity of a dynamically constructed graph

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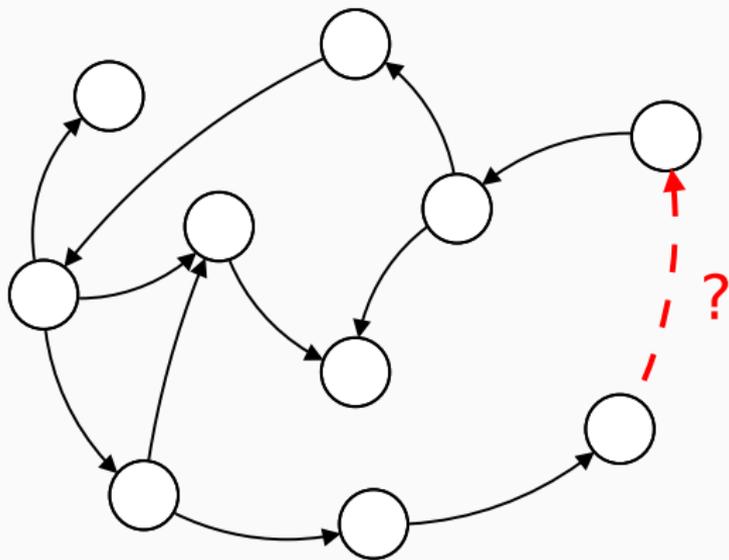
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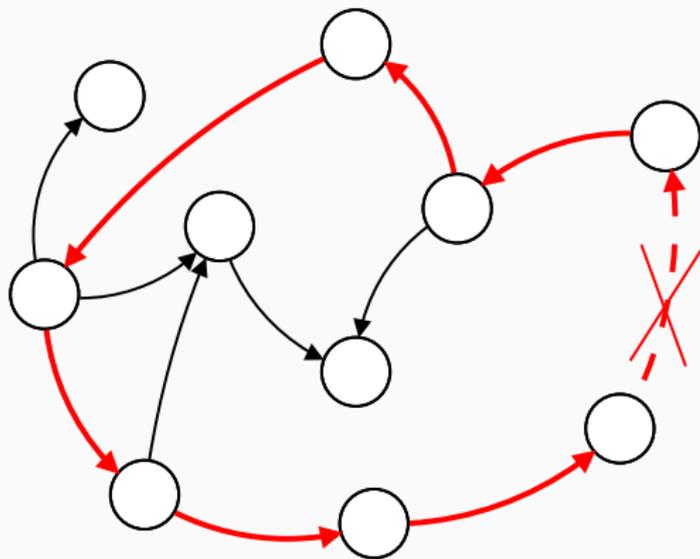
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The problem: checking for acyclicity of a dynamically constructed graph

Minimal OCaml interface

```
type add_edge_result =  
  | EdgeAdded  
  | EdgeCreatesCycle
```

```
val add_edge_or_detect_cycle :  
  graph -> vertex -> vertex -> add_edge_result
```

Our main case study (2)

A state-of-the-art algorithm:

- non-trivial implementation (200 lines of compact OCaml code)
- subtle complexity analysis
- used in Coq (universe constraints) and Dune (build dependencies)

Incremental Cycle Detection: Complexity

Naive algorithm: $O(m)$ traversal at each arc insertion.
Inserting m arcs costs $O(m^2)$.

Using Bender et al.'s algorithm, inserting m arcs costs:

$$O(m \cdot \min(\sqrt{m}, n^{2/3}))$$

Or:

- $O(m\sqrt{m})$ for sparse graphs;
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Specifies the **cost of a sequence of operations**.

No closed formula for the amortized cost of a single operation.

Toplevel specification (functional correctness only)

“IsDAG g G ”: a **Separation Logic predicate** describing the algorithm’s data structure, at address g , representing the graph G .

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$$\forall g G v w. \text{ let } m := |\text{edges } G| \text{ in}$$
$$\text{ let } n := |\text{vertices } G| \text{ in}$$
$$v, w \in \text{vertices } G \wedge (v, w) \notin \text{edges } G \implies$$
$$\left\{ \text{IsDAG } g G \right\}$$
$$(\text{add_edge_or_detect_cycle } g v w)$$
$$\left\{ \begin{array}{l} \lambda \text{ res. match res with} \\ \quad | \text{EdgeAdded} \Rightarrow \text{IsDAG } g (G + (v, w)) \\ \quad | \text{EdgeCreatesCycle} \Rightarrow [w \xrightarrow{*}_G v] \end{array} \right\}$$

Toplevel specification (correctness and complexity)

“IsDAG g G ”: a **Separation Logic predicate** describing the algorithm’s data structure, at address g , representing the graph G .

$$\begin{aligned} \exists \psi. & \quad \psi \in O(m \cdot \min(\sqrt{m}, n^{2/3}) + n) \wedge \\ \forall g G v w. & \quad \text{let } m := |\text{edges } G| \text{ in} \\ & \quad \text{let } n := |\text{vertices } G| \text{ in} \\ & \quad v, w \in \text{vertices } G \wedge (v, w) \notin \text{edges } G \implies \\ & \quad \left\{ \text{IsDAG } g G \star \$(\psi(m + 1, n) - \psi(m, n)) \right\} \\ & \quad (\text{add_edge_or_detect_cycle } g v w) \\ & \quad \left\{ \begin{array}{l} \lambda \text{ res. match res with} \\ \quad | \text{EdgeAdded} \Rightarrow \text{IsDAG } g (G + (v, w)) \\ \quad | \text{EdgeCreatesCycle} \Rightarrow [w \xrightarrow{*}_G v] \end{array} \right\} \end{aligned}$$

Case Study: Summary

Final result

- A formally verified OCaml library for incremental cycle detection
- Succinct specification
- Robust proof (no hardcoded constants or manual accounting)
- Code has been integrated in Dune, fixing some complexity bugs

Contributions

- State-of-the-art result on verified graph algorithms
- A crucial improvement to the algorithm to make it truly incremental

Conclusion

Summary

In this talk:

- Motivation for the verification of complexity using O
- Cost functions and their inference
- Possibly Negative Time Credits
- A large case study

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More in the manuscript:

- Specific challenges related to multivariate O
- Summation lemmas for the analysis of for-loops
- More case studies

Perspectives

Further automation

- in Coq: high-level reasoning on synthesized cost expressions (master theorem, simplification procedures)
- integration with automated complexity analysis tools
- integration of the approach in more automated verification tools

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Implement support to allow extracting concrete complexity bounds

Even more challenging applications:

- space complexity
- concurrent programs
- cache-oblivious algorithms