

Coalgebraic Logic Programming

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Outline

- 1 Recursion and Corecursion
 - Inductive and Coinductive Types in Coq
 - Terminative and Productive Functions
 - Recursion and Corecursion without types

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- 3 Parallelism

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- 3 Parallelism
- 4 Future directions: Applications to type inference

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- 3 Parallelism
- 4 Future directions: Applications to type inference
- 5 Appendix: LP in Type inference

Today's talk...

...continuation of Thanos'es talk of yesterday:

- about logic programming (LP);
- about first order (= in Thanos'es terms infinite) language for LP;
- about how much we can merge methods of FP and LP...
- may be you will see some references to a possible game semantics.

Inductive Types and Recursive Functions

```
Inductive list (A : Type) : Type :=  
  | nil : list A  
  | cons : A -> list A -> list A.
```

Recursive functions have arguments of inductive types.

```
Fixpoint length (A:Type) (l: list A) : nat :=  
  match l with  
  | nil => 0  
  | _ :: l' => S (length l')  
  end.
```


Coinductive Types and Corecursive Functions

```
CoInductive stream (A:Set) : Set :=  
SCons: A -> stream A -> stream A.
```

Corecursive functions have outputs of coinductive types. (Type of input arguments is not important.)

```
CoFixpoint map (s:Stream A) : Stream B :=  
SCons (f (hd s)) (map (tl s)).
```

Termination

We require all computations to terminate, because of:

- Curry-Howard Isomorphism (propositions $\rightarrow\leftarrow$ types; proofs $\rightarrow\leftarrow$ programs): non-terminating proofs can lead to inconsistency.

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- Curry-Howard Isomorphism (propositions $\rightarrow\leftarrow$ types; proofs $\rightarrow\leftarrow$ programs): non-terminating proofs can lead to inconsistency.
- To decide type-checking of dependent types, we need to reduce expressions to normal form.

Productive Values

Values in co-inductive types are **productive** when all observations of fragments made using recursive functions are guaranteed to be computable in finite time.

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The element of the stream at position n can be found by:

Definition

$$\begin{cases} \text{nth } 0 \text{ (SCons } a \text{ tl)} = a \\ \text{nth (S } n \text{) (SCons } a \text{ tl)} = \text{nth } n \text{ tl} \end{cases}$$

A given stream s is productive if we can be sure that the computation of the list $\text{nth } n \text{ } s$ is guaranteed to terminate, whatever the value of n is.

We call a function *productive at the input value i* , if it outputs a productive value at i .

Deciding Termination: Structural Recursion

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Deciding Productivity: Guardedness

The guardedness condition insures that

- * each corecursive call is made under at least one constructor;
- ** if the recursive call is under a constructor, it does not appear as an argument of any function.

Violation of any of these two conditions makes a function rejected by the guardedness test in Coq.

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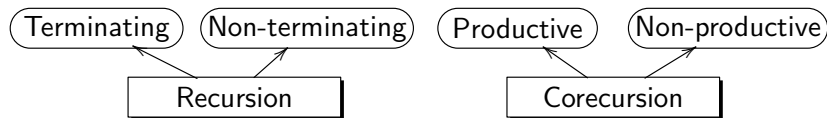
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CoFixpoint map (s:Stream A) : Stream B :=  
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```

To notice:



- The role of types in definition of (co)recursive functions;
- The role of constructors and (co)-pattern matching;

Recursion and Corecursion in Logic Programming

Example

```
nat(0) ←  
nat(s(x)) ← nat(x)  
list(nil) ←  
list(cons x y) ← nat(x), list(y)
```

Example

```
bit(0) ←  
bit(1) ←  
stream(cons (x,y)) ← bit(x), stream(y)
```

SLD-resolution (+ unification and backtracking) behind LP derivations.

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← □
```

The answer is x/O , y/nil , but we can get more substitutions by backtracking. We can backtrack infinitely many times, but each time computation will terminate.

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The answer is x/O , y/nil , but we can get more substitutions by backtracking. We can backtrack infinitely many times, but each time computation will terminate.

Nice, clean semantics: least Herbrand model exists, sound&complete, etc.: see Thanos'es Viva of yesterday.

Corecursion in LP?

Example

`bit(0) ←`

`bit(1) ←`

`stream(scons(x, y)) ←`

`bit(x), stream(y)`

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No answer, as derivation never terminates.

Semantics may go wrong as well: least Herbrand models will contain an infinite term corresponding to stream: so completeness fails.

```
← stream(scons(x, y))  
    |  
← bit(x), stream(y)  
    |  
    ← stream(y)  
        |  
← bit(x1), stream(y1)  
    |  
    ← stream(y1)  
        |  
← bit(x2), stream(y2)  
    |  
    ← stream(y2)  
        |  
        ⋮
```

It can be worse....

Example

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Semantics goes wrong: this time, soundness!

```
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  |  
← bit(x), list(y)  
  |  
  ← list(y)  
    |  
← bit(x1), list(y1)  
  |  
  ← list(y1)  
    |  
← bit(x2), list(y2)  
  |  
  ← list(y2)  
    |  
    ⋮
```

To notice:

- Distinction between (co)inductive type, (co)recursive function over (co)inductive type and a proof by (co)induction is erased.
- Without types guarding (co)recursion, things get messy:
 - ▶ ...not "just" termination, but also semantics
- We do not have a formalism to speak about termination and productivity, or generally, recursion/corecursion.

Note aside: LP has instances of dependent types, mixed induction/coinduction, recursion/corecursion....

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CoALP: what is it about?

- syntactically – first-order logic programming;
- operationally – lazy (co)recursion;
- inspired by coalgebraic fibrational semantics;
- explores the tree-structure of partial proofs – “coinductive trees”;
- uses lazy guarded corecursion using measures of corecursive steps given by coinductive trees (cf. “clocked corecursion”);
- parallel...

Fibrational Coalgebraic Semantics of CoALP in 3 ideas

Operational View of Logic Programming

- 1 Let At be the set of all atoms appearing in a program P . Then P can be identified with a $P_f P_f$ -coalgebra (At, p) , where $p : At \rightarrow P_f(P_f(At))$ sends an atom A to the set of bodies of those clauses in P with head A .

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- 2 Taking $p : At \rightarrow P_f P_f(At)$, the corresponding $C(P_f P_f)$ -coalgebra where $C(P_f P_f)$ is the cofree comonad on $P_f P_f$ is given as follows: $C(P_f P_f)(At)$ is given by a limit of the form

$$\dots \rightarrow At \times P_f P_f(At \times P_f P_f(At)) \rightarrow At \times P_f P_f(At) \rightarrow At.$$

This gives a “tree-like” structure: we call them **&V-trees**.

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- 3 For first order extension: Take *Lawvere Theory* \mathcal{L}_Σ to model the signature Σ (objects are natural numbers, arrows – term arities, composition = substitution), and take $\mathcal{L}_\Sigma \rightarrow Set$ – to model (predicates in) At .

Examples

Program Stream: fibers are term arities. Take the fiber of 1. & V -trees:

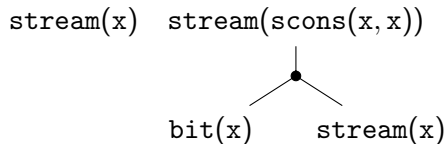
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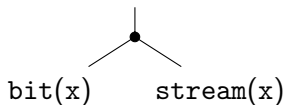
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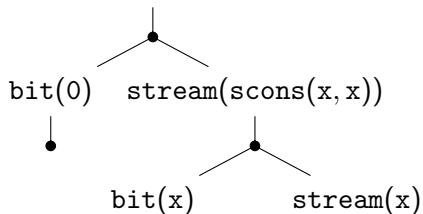
Examples

Program Stream: fibers are term arities. Take the fiber of 1. & V-trees:

stream(x) stream(scons(x, x))



stream(scons(0, scons(x, x)))



Computationally essential:

- ① for coinductive `Stream` program, the $\&V$ -trees are finite!!! – both in depth and in breadth;
- ② each tree gives only a partial computation – it is not like eager SLD-trees we have seen earlier;
- ③ the effect of fibers is best modelled by restricting unification to term-matching (note resemblance to the pattern-matching in Functional setting).

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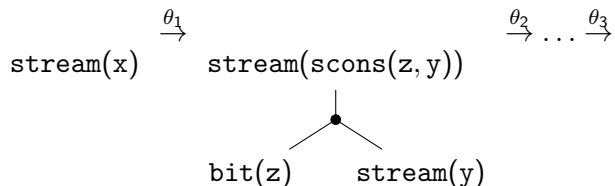
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1. \Rightarrow gives hope for a formalism to describe termination and productivity
2. \Rightarrow hints there may be laziness involved...

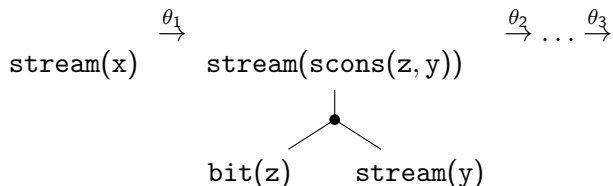
Lazy Corecursion in CoALP: Coinductive trees

`stream(x)` $\xrightarrow{\theta_1}$

Lazy Corecursion in CoALP: Coinductive trees



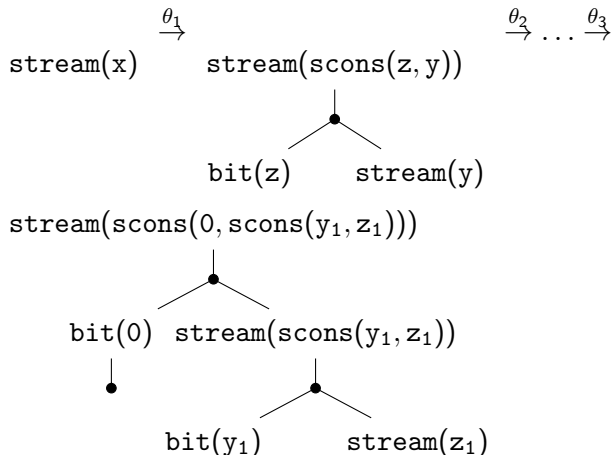
Lazy Corecursion in CoALP: Coinductive trees



Note that transitions θ may be determined in a number of ways:

- using mgus;
- non-deterministically;
- in a distributed/parallel manner.

Lazy Corecursion in CoALP



The above would correspond to one-branch of SLD-derivations we have seen! The main driving force: separation of layers of computations into different dimensions.

Properties:

Komendantskaya, Power, Schmidt: **Coalgebraic Logic Programming: from Semantics to Implementation**, Journal of Logic and Computation, 2014.

- Sound and complete with respect to the coalgebraic semantics;
- Finite computations are sound and complete with respect to the least Herbrand model semantics (so, we can do as much as standard Prolog for sure).
- Adequacy result for observational semantics.

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What does it tell us beyond LP?

CoALP: the three-dimensional calculus of trees

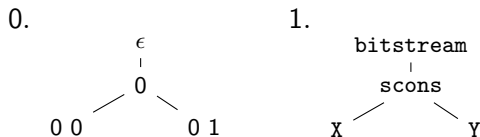
- 1 Dimension 1: term-trees;
- 2 Dimension 2: coinductive trees;
- 3 Dimension 3: derivation trees.

Dimension-1: Term-trees

Take a tree-language \mathbb{N}^* .

Given an $L \in \mathbb{N}^*$, a term tree is a map $L \rightarrow \Sigma$, satisfying term arity restrictions.

Example:



Notation:

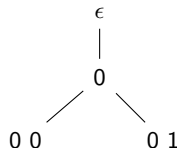
Term (Σ)	<i>finite</i> term trees over Σ
Term ^{∞} (Σ)	<i>infinite</i> term trees over Σ
Term ^{ω} (Σ)	<i>finite and infinite</i> term trees over Σ

Dimension-2: Coinductive trees

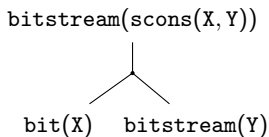
Given an $L \in \mathbb{N}^*$, a coinductive tree is a map $L \rightarrow \mathbf{Term}(\Sigma_P)$, satisfying coinductive tree construction for P .

Example:

0.



2.



Notation:

$\mathbf{CTree}(\mathbf{Term}(\Sigma_P), P)$

$\mathbf{CTree}^\infty(\mathbf{Term}(\Sigma_P), P)$

$\mathbf{CTree}^\omega(\mathbf{Term}(\Sigma_P), P)$

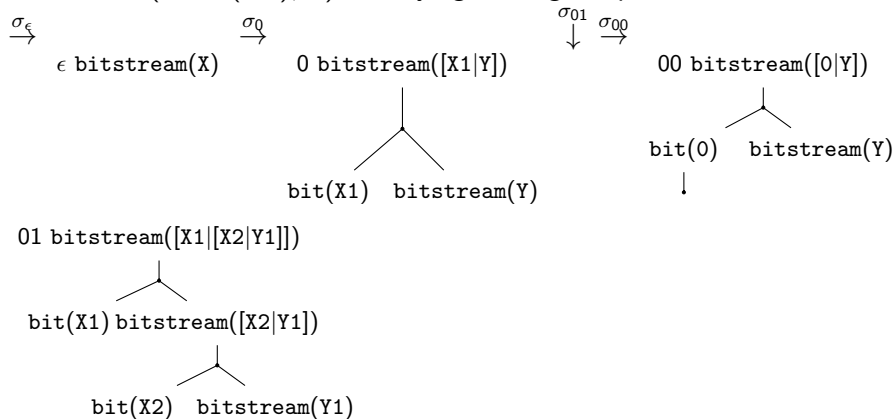
all *finite* coinductive trees over $\mathbf{Term}(\Sigma_P)$

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all *finite and infinite* coinductive trees over $\mathbf{Term}(\Sigma_P)$

Dimension-3: Derivation trees

Given an $L \in \mathbb{N}^*$, a coinductive derivation is a map $L \rightarrow \mathbf{CTree}(\mathbf{Term}(\Sigma_P), P)$, satisfying the mgu requirement.



Dimension-3 notation

$\mathbf{CDer}(\mathbf{CTree}(\mathbf{Term}(\Sigma_P), P))$	all <i>finite</i> coinductive derivations over $(\mathbf{CTree}(\mathbf{Term}(\Sigma_P), P))$
$\mathbf{CDer}^\infty(\mathbf{CTree}(\mathbf{Term}(\Sigma_P), P))$	all <i>infinite</i> coinductive trees over $(\mathbf{CTree}(\mathbf{Term}(\Sigma_P), P))$
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Theory of Productivity for LP

A first-order logic program P is *productive* if

for any term $t \in \mathbf{Term}(\Sigma_P)$, the coinductive tree CT with the root t belongs to $C\mathbf{Tree}(\mathbf{Term}(\Sigma_P), P)$.

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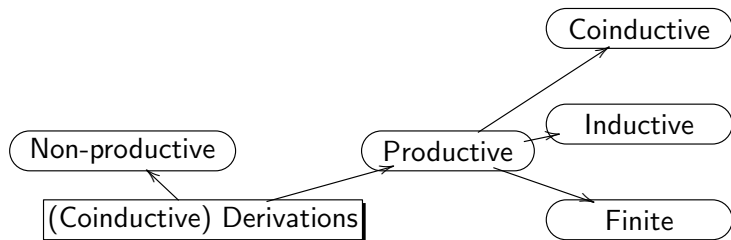
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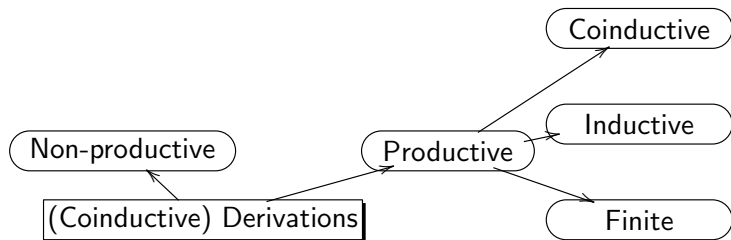
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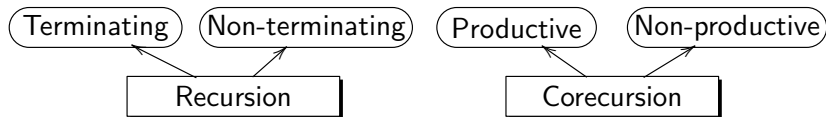
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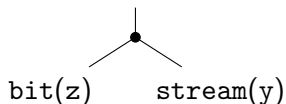
Compare with:



Deciding Productivity: Guardedness

- Dimension 1. Measures of reduction on term trees:
`stream(y)` is a reduction of `stream(scons(x,y))`
- Dimension 2. Reduction on coinductive tree loops:

`stream(scons(z,y))`



- Dimension 3. Discovery of derivation loops.

Example of guardedness issues

$$p(s(X1), X2, Y1, Y2) \leftarrow q(X2, X2, Y1, Y2)$$
$$q(X1, X2, s(Y1), Y2) \leftarrow p(X1, X2, Y2, Y2)$$

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♦

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$$\quad | \\ \quad \vdots$$

Soundness of Corecursion in LP

- CoALP is sound and complete for inductive programs;
- Soundness of coinductive programs is our next step.

Two directions:

- Imposing guardedness conditions, to ensure every coinductive tree is finite.

To ban programs that are not guarded by constructors:

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Unlike termination checks in Coq/Agda cannot be done fully statically (no types to help!), and needs some proof search in Dimension 3.

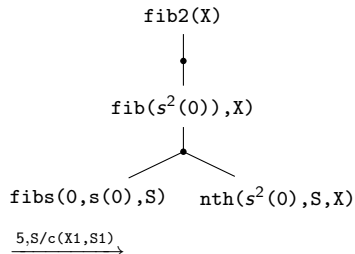
- Determining when it is safe to make a coinductive conclusion (and finding a right coinductive hypothesis).
(Again, the troubles come from un-typed setting.)

Stream of Fibonacci numbers:

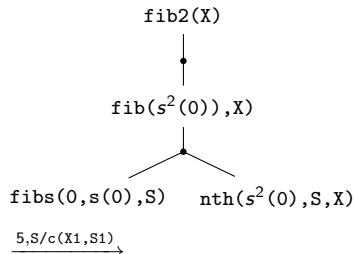
Falls into infinite loops in Prolog and Prolog-like version of CoLP [Gupta et al. 2007] [Both are eager...] Those powerful SAT/SMT solvers would not do it either.

1. `add(0,Y,Y).`
2. `add(s(X),Y,s(Z)) :- add(X,Y,Z).`
3. `fibs(X,Y,cons(X,S)) :- add(X,Y,Z), fibs(Y,Z,S).`
4. `nth(0,cons(X,S),X).`
5. `nth(s(N),cons(X,S),Y) :- nth(N,S,Y).`
6. `fib(N,X) :- fibs(0,s(0),S), nth(N,S,X).`
7. `fib2(X) :- fib(s(s(0)),X).`

Examples of derivations with Fib: lazy step 1

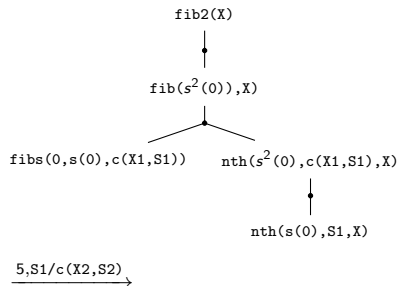


Examples of derivations with Fib: lazy step 1

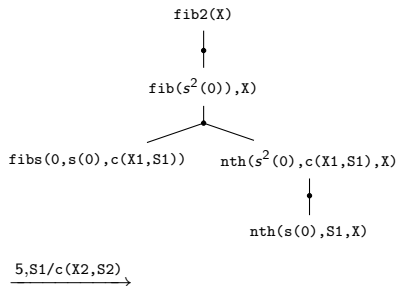


1. add(0, Y, Y).
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add(X, Y, Z).
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4. nth(0, cons(X, S), X).
5. nth(s(N), cons(X, S), Y) :-
nth(N, S, Y).
6. fib(N, X) :- fibs(0, s(0), S),
nth(N, S, X).
7. fib2(X) :- fib(s(s(0)), X).

Examples of derivations with Fib: lazy step 2

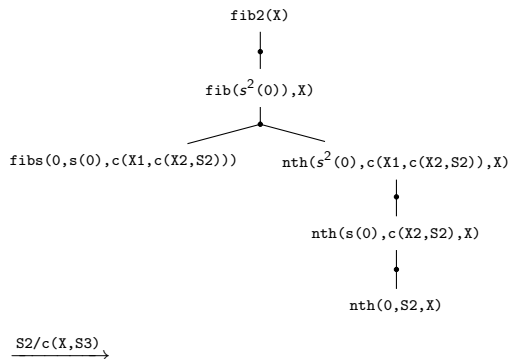


Examples of derivations with Fib: lazy step 2

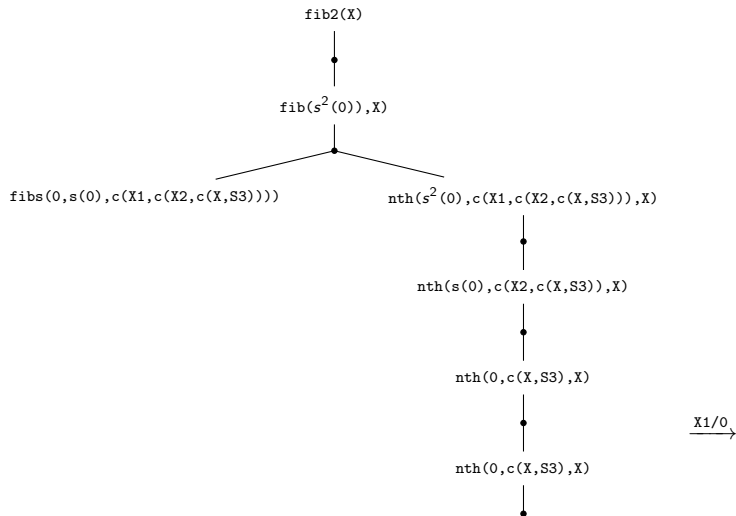


1. $\text{add}(0, Y, Y).$
2. $\text{add}(s(X), Y, s(Z)) :- \text{add}(X, Y, Z).$
3. $\text{fibs}(X, Y, \text{cons}(X, S)) :- \text{add}(X, Y, Z), \text{fibs}(Y, Z, S).$
4. $\text{nth}(0, \text{cons}(X, S), X).$
5. $\text{nth}(s(N), \text{cons}(X, S), Y) :- \text{nth}(N, S, Y).$
6. $\text{fib}(N, X) :- \text{fibs}(0, s(0), S), \text{nth}(N, S, X).$
7. $\text{fib2}(X) :- \text{fib}(s(s(0)), X).$

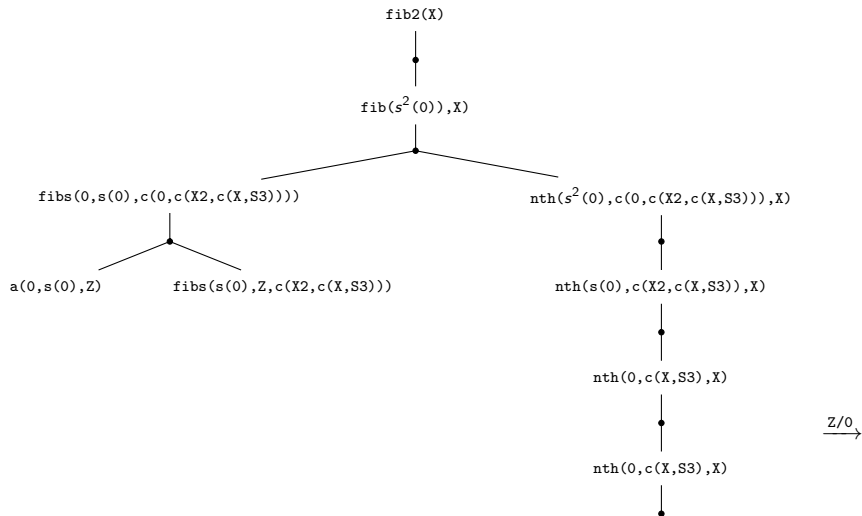
Examples of derivations with Fib: lazy step 3



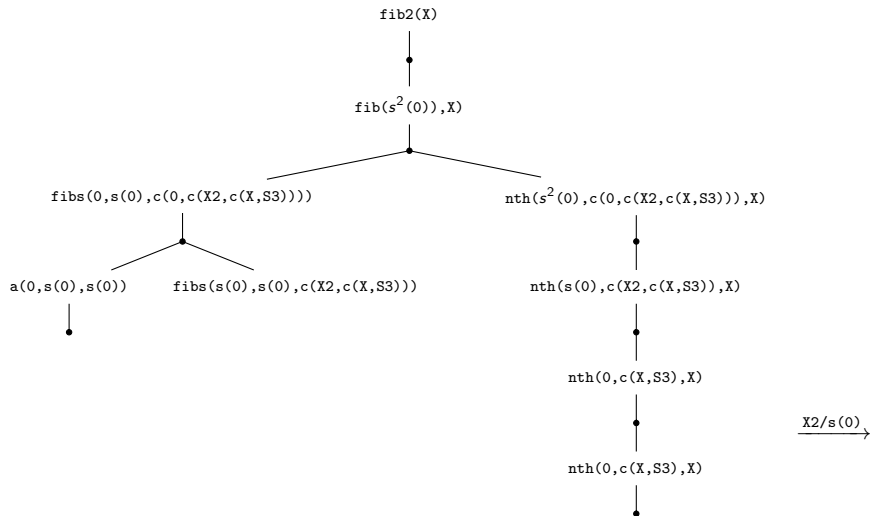
Examples of derivations with Fib: lazy step 4



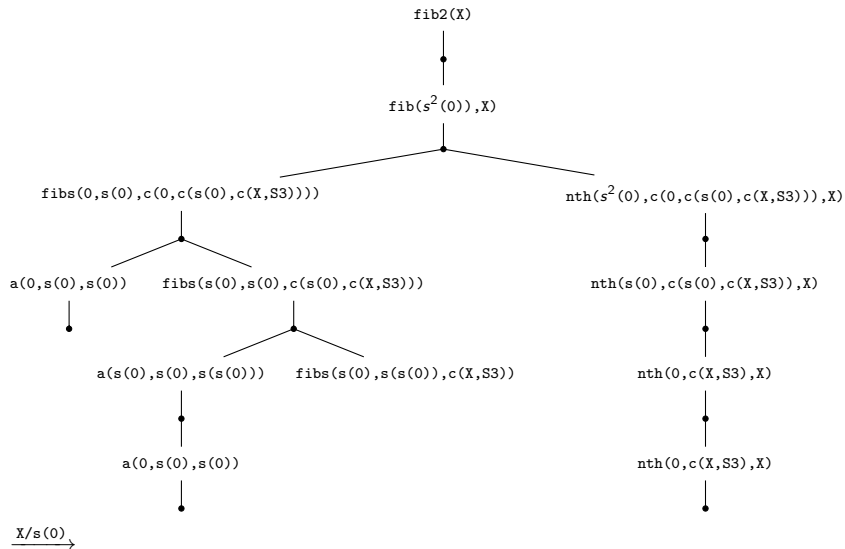
Examples of derivations with Fib: lazy step 5



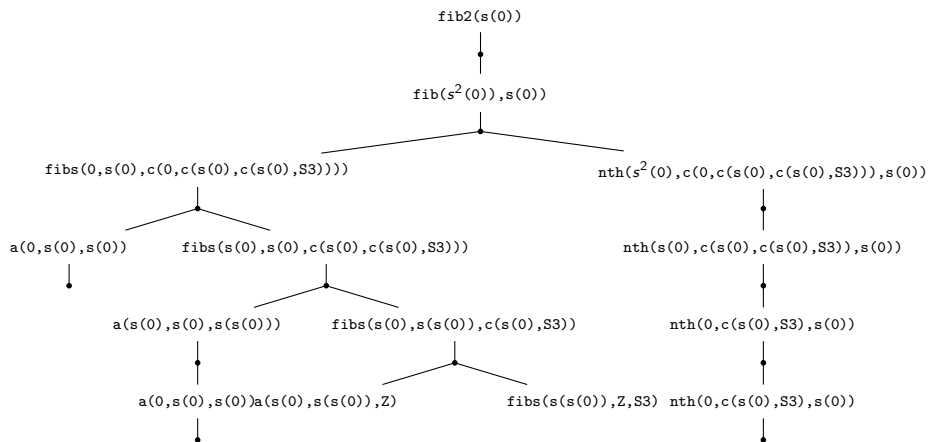
Examples of derivations with Fib: lazy step 6



Examples of derivations with Fib: lazy step 8



Examples of derivations with Fib: lazy step 9



Logic Programming dialects, compared

	Prolog	Parallel Prolog	Co-LP	CoALP
Fib example	No	No	No	Yes
Execution	Eager	Eager	Eager	Lazy
Corecursion				
Mode of execution				
Declarative semantics				
Operational semantics				

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Mode of execution	Sequential	Parallel	Sequential	Parallel
Declarative semantics				
Operational semantics				

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Declarative semantics	lfp	lfp	gfp (restricted)	coalgebraic
Operational semantics				

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Operational semantics	transitions; states: lists of formulae	transitions; states: lists of formulae	transitions; states: lists of formulae	transitions; states: coinductive trees

Outline

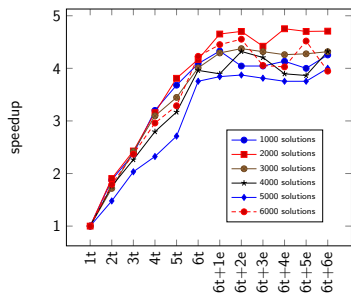
- 1 Recursion and Corecursion
 - Inductive and Coinductive Types in Coq
 - Terminative and Productive Functions
 - Recursion and Corecursion without types
- 2 Coalgebraic Logic Programming
- 3 Parallelism**
- 4 Future directions: Applications to type inference
- 5 Appendix: LP in Type inference

Parallelising CoALP

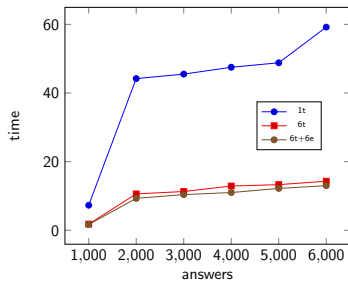
Komendantskaya, Schmidt, Heras: *Exploiting Parallelism in Coalgebraic Logic Programming*, ENTCS, 2014

1. `bit(0).`
2. `bit(1).`
3. `btree(empty).`
4. `btree(tree(L,X,R)) ← btree(L), bit(X), btree(R).`

Parallel CoALP



threads (t) and expand threads (e)



Directions we are exploring

Haskell implementation is nearly finished.

Current task: to find a “right” language to try CoALP-based type inference

- Using CoALP in Hume: for analysis of stream-based networks and/or for type inference;
- Type-inference in Haskell;
- SSReflect: overloading in canonical structures currently requires the use of back-tracking in LP-like algorithm. It could be parallel CoALP execution instead;
- CoALP for global type analysis in object-oriented languages: CoLP is already used for that.
- Formal Verification of CoALP-based type inference

The end

- Komendantskaya, Power, Schmidt: **Coalgebraic Logic Programming: from Semantics to Implementation**, Journal of Logic and Computation, 2014.
- Komendantskaya, Schmidt, Heras: *Exploiting Parallelism in Coalgebraic logic Programming*, ENTCS, 2014.
- A paper on implementing lazy guarded corecursion in CoALP using Haskell is in preparation...
- CoALP webpage has various prototype implementations to play with... <http://staff.computing.dundee.ac.uk/katya/CoALP/>

We will be happy to apply CoALP for TI (or other purposes) in ***YOUR*** language!

Milner, 1978

“A theory of Type Polymorphism in Programming”

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An elegant match between polymorphic λ -calculus and type inference by means of Robinson's unification/resolution algorithm.

Trend in type inference:

improvement in **expressiveness** of the underlying type system, e.g., in terms of

- *Dependent Types*,
- *Type Classes* [Wadler&Blott 89],
- *Generalised Algebraic Types* (GADTs) [Peyton Jones & al, 2006]
- *Dependent Type Classes* [Sozeau & al 08] and
- *Canonical Structures* [Gonthier& al 11].

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Milner-style decidable type inference does not always suffice (e.g. the *principal type* may no longer exist), and TI requires additional inference algorithms.

Implementations of new type inference algorithms include a variety of first-order decision procedures, notably Unification and Logic Programming (LP) [Peyton Jones & al, 2006], Constraint LP [Odersky Sulzmann, Vytiniotis & many more 1999-], LP embedded into interactive tactics (Coq's *eauto*) [Sozeau & al. 08], and LP supplemented by rewriting [Gonthier & al, 11].

Motivation: type inference with Polymorphic types

List Length in Haskell

```
length :: [a] -> Integer
length [] = 0
length (x:xs) = 1 + length xs
```

Logic program for type inference

```
cons(X) ← X = Y → list(Y) → list(Y).
plus(X) ← X = int → int → int.
nil(X) ← X = list(Y).
length(X) ← (X = Y → Z) & nil(Y) & Z = int & cons(W) &
             (W = W1 → W2 → Y) & plus(U) &
             (U = int → Z → Z) & W2 = Y.
```

Query: $\text{length}(X)$?

Answer (any existing PROLOG version): $X = \text{list}(-) \rightarrow \text{int}$.

Trend to do more by type-inference:

... session types,

... writing contracts by means of types:

Example

Vytiniotis et al. "HALO: Haskell to Logic Through Denotational Semantics" [POPL'13]

```
f xs = head (reverse (True : xs))
```

```
g xs = head (reverse xs)
```

Both f and g are well typed and "can't go wrong" in Milner's sense, but g will crash for empty list, and f will never crash.

Contract:

$$\text{reverse} \in (xs : CF) \rightarrow \{ys \mid \text{null } xs \Leftrightarrow \text{null } ys\}$$

Requires strong first-order type inference engines: Z3, Vampire, E...

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- Would it pay-off to get more conceptually elegant on type inference side? – especially bearing in mind the big emphasis on type inference in more expressive type systems.
- Would our “Coalgebraic Logic programming” grow to become a type-inference specific theorem prover (with stronger theoretical background and motivation than state-of-the-art SAT/SMT-solvers)?