Productive Corecursion in Logic Programming

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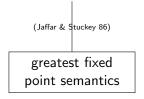
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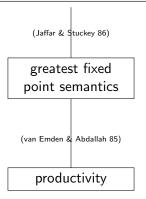
Outline

- Motivation
 - Overview of motivation
 - Background knowledge for understanding motivation
 - Problem description
- Productive Corecursion
 - Loop detection rule review
 - Productivity guarantee
- Conclusion
- 4 Future Work & Implementation

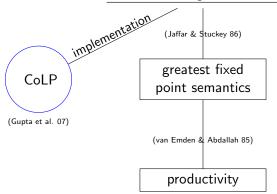
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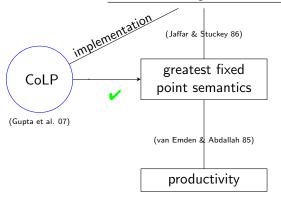




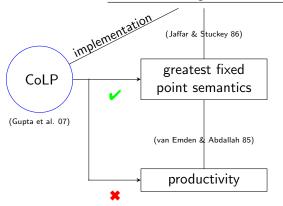


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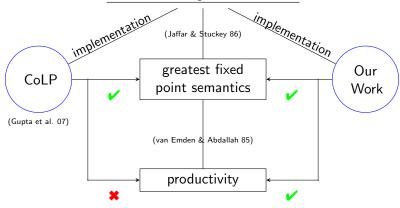


✓: sound



: sound

*****: not sound



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Term:=Constant | Variable | Functor (<List of Terms>)

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 $Definite \ clause := Term \leftarrow Set \ of \ Terms$

Goal clause:=List of Terms

Program:=Set of Definite clauses

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the least fixed point is the smallest set closed forward under the program.

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nat(0)

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$$nat(0)$$

 $nat(s(X)) \leftarrow nat(X)$

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Example

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nat(0)

nat(s(X)) \leftarrow nat(X)
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The least fixed point is $\{nat(0), nat(s(0)), nat(s(s(0))), \dots \}$.

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Example

```
\label{eq:nat(0)} \begin{split} & \mathsf{nat}(\mathsf{s}(\mathsf{X})) \leftarrow \mathsf{nat}(\mathsf{X}) \\ & \mathsf{The least fixed point is } \{\mathsf{nat}(\mathsf{0}),\, \mathsf{nat}(\mathsf{s}(\mathsf{0})),\, \mathsf{nat}(\mathsf{s}(\mathsf{s}(\mathsf{0}))),\, \dots \}. \\ & \mathsf{The greatest fixed point is } \{\mathsf{nat}(\mathsf{0}),\, \mathsf{nat}(\mathsf{s}(\mathsf{0})),\, \mathsf{nat}(\mathsf{s}(\mathsf{s}(\mathsf{0}))),\, \dots \} \cup \\ \{\mathsf{nat}(\mathsf{s}(\mathsf{s}(\ldots)))\}. \end{split}
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```

Formulae computed by non-terminating derivations are in greatest fixed points. (Jaffar & Stuckey 86; van Emden & Abdallah 85)

4 D > 4 A > 4 B > 4 B > B 9 9 9

(LP: van Emden & Abdallah 86; Komendantskaya et al. 16;

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A productive non-terminating derivation does useful computations while looping rather than just looping.

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 $nat(X) \leftarrow nat(X)$

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Example

 $nat(X) \leftarrow nat(X)$ has non-productive derivation:



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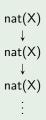
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Example

 $\mathsf{nat}(X) \leftarrow \mathsf{nat}(X) \\ \mathsf{has} \ \mathsf{non\text{-}productive} \ \mathsf{derivation} \colon$



Example

 $\mathsf{nat}(\mathsf{s}(\mathsf{X})) \leftarrow \mathsf{nat}(\mathsf{X})$

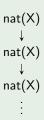
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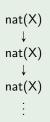
 $nat(s(X)) \leftarrow nat(X)$ computes the first limit ordinal $nat(s(s(\dots)))$:

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$$\begin{array}{c} \mathsf{nat}(\mathsf{X}) \\ \downarrow^{\mathsf{X} \mapsto s(X_2)} \\ \mathsf{nat}(X_2) \\ \downarrow^{\mathsf{X}_2 \mapsto s(X_3)} \\ \mathsf{nat}(X_3) \\ \vdots \end{array}$$

Now consider finite implementation of non-terminating SLD derivations.

Since regular formulae have cyclic derivations, finding a cycle (loop) is suffice for knowing the whole derivation.

Definition (Gupta et al. 07)

CoLP = SLD resolution + loop detection rule

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A goal succeeds if it unifies with its ancestor goal.

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Theorem (Coinductive soundness of coLP) (Gupta et al. 07)

Successful coLP derivations only compute formulae in greatest fixed points.



 $\mathsf{nat}(\mathsf{s}(\mathsf{X})) \leftarrow \mathsf{nat}(\mathsf{X})$ defines the first limit ordinal $\mathsf{s}(\mathsf{s}(\dots))$.

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SLD derivation (non-terminating)

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SLD derivation (non-terminating) $G_0 \quad \mathsf{nat}(\mathsf{X}) \\ \qquad \qquad \downarrow^{\mathsf{X} \mapsto s(\mathsf{X}_2)} \\ G_1 \quad \mathsf{nat}(\mathsf{X}_2) \\ \qquad \qquad \downarrow^{\mathsf{X}_2 \mapsto s(\mathsf{X}_3)} \\ \mathsf{nat}(\mathsf{X}_3) \\ G_2 \qquad \vdots$

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SLD derivation (non-terminating)
$$G_0 \quad \text{nat}(X) \\ \qquad \downarrow \quad x \mapsto s(X_2)$$

$$G_1 \quad \text{nat}(X_2) \\ \qquad \downarrow \quad x_2 \mapsto s(X_3)$$

$$\text{nat}(X_3)$$

$$G_2 \qquad \vdots$$

CoLP derivation (terminating)

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SLD derivation (non-terminating)
$$G_0 \quad \operatorname{nat}(X) \\ \downarrow^{X \mapsto s(X_2)} \\ G_1 \quad \operatorname{nat}(X_2) \\ \downarrow^{X_2 \mapsto s(X_3)} \\ \operatorname{nat}(X_3) \\ G_2 \qquad \vdots$$

$$\begin{array}{c} \mathsf{CoLP} \; \mathsf{derivation} \\ \; \; (\mathsf{terminating}) \\ G_0 \; \; \mathsf{nat}(\mathsf{X}) \\ \; \; \; \downarrow \; \; \mathsf{x} \mapsto \mathsf{s}(\mathsf{X}_2) \\ G_1 \; \; \mathsf{nat}(\mathsf{X}_2) \\ \; \; \downarrow \; \; \mathsf{X}_2 \mapsto \mathsf{X} \; (\mathsf{G}_1 \; \mathsf{unifies} \; \mathsf{G}_0) \\ G_2 \; \; \Box \\ \end{array}$$

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Assume some successful coLP derivation that computes an infinite formula. Problem 1: It is not guaranteed that there exists a corresponding non-terminating SLD derivation.

e.g. For program $p(f(X),X) \leftarrow p(X,X)$ and goal p(f(X),X), coLP computes $p(f(f(\dots)),f(f(\dots)))$ but here is no non-terminating SLD derivation.

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$$G_0: p(f(X),X)$$

$$\downarrow G_1: p(X,X)$$

$$\downarrow X \mapsto f(f(...)) \text{ by unifying } G_1 \text{ with } G_0$$
 $G_2: \square$

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SLD derivation

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$$\downarrow$$
 $G_1: p(X,X)$

 G_1 does not unify head of clause

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Assume some successful coLP derivation that computes an infinite formula.

Problem 2: There exists a corresponding non-terminating SLD derivation but it computes a different formula.

e.g. For program $q(f(X),Y) \leftarrow q(X,h(Y))$ and goal q(f(X),Y), coLP computes $q(f(f(\dots)),h(h(\dots)))$ but the corresponding non-terminating SLD derivation computes $q(f(f(\dots)),Y)$.

CoLP derivation

$$\begin{array}{ll} G_0: & \mathsf{q}(\mathsf{f}(\mathsf{X}),\mathsf{Y}) \\ \downarrow \\ G_1: & \mathsf{q}(\mathsf{X},\mathsf{h}(\mathsf{Y})) \\ & \downarrow X \mapsto f(f(\ldots)), Y \mapsto h(h(\ldots)) \text{ by unifying } G_1 \text{ with } G_0 \\ G_2: & \square \end{array}$$

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$$G_0: q(f(X),Y)$$

$$\downarrow \qquad \qquad \downarrow$$

$$G_1: q(X,h(Y))$$

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```
G_0: q(f(X),Y)
  SLD derivation
                             G_1: q(X,h(Y))
(non-terminating)
                                         \downarrow X \mapsto f(X_2)
                           G_2: q(X_2,h(h(Y)))
                                         \downarrow X_2 \mapsto f(X_3)
                         G_3: q(X_3,h(h(h(Y))))
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                \downarrow X_2 \mapsto f(X_3)
G_3: q(X_3,h(h(h(Y))))
```



Can we have an implementation of infinite SLD derivation,

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Our answer is affirmative.

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- Case 2 One is a variant or instance of the other. e.g. nat(X) and nat(s(X));

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instead of

Definition (Loop detection rule) (Gupta et al. 07)

A goal succeeds if it unifies with its ancestor goal.

We also characterized a class of logic programs whose non-terminating SLD derivations, if any, are guaranteed to be productive.

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Example (Rewriting)

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Example (Rewriting)

 From goal nat(s(X)) and clause nat(Y)← nat(s(Y)), derive nat(s(s(Y))).

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- 2 free from existential variables,

is guaranteed to be productive for its non-terminating SLD derivations, if any.

Why termination for rewriting plays a role?

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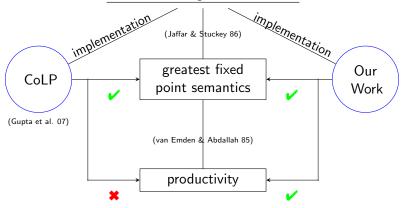
Theorem (our main result: Productivity Semi-decision)

Productivity is semi-decidable for programs characterized above, by SLD resolution combined with our loop detection rule.

- Motivation
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non-terminating SLD derivation



✓: sound

x: not sound

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Streams of consecutive numbers, e.g. 1 2 3 ... or 99 100 101 ..., are defined by the corecursive clause $from(X,[X|T]) \leftarrow from(s(X),T)$.



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Example (Fibonacci stream) (Komendantskaya et al. 15)

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Streams of Fibonacci numbers, e.g. 1 1 2 3 5 8 ... or 10 4 14 18 32 ..., are defined by a corecursive clause that has an existential variable. $\mathsf{fibs}(X,Y,[X|S]) \leftarrow \mathsf{add}(X,Y,Z), \, \mathsf{fibs}(Y,Z,S).$

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Implementation is available at GitHub / coalp / Productive-Corecursion

Thanks!