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Handout for part 5:

Complexity: tuple interpretations

1. Monotonic algebras

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Derivation height

A measure of the “cost” of reducing a term to normal form (worst-case).

$$\begin{aligned}\text{add}(x, 0) &\rightarrow x \\ \text{add}(x, \mathbf{s}(y)) &\rightarrow \mathbf{s}(\text{add}(x, y)) \\ \text{mul}(x, 0) &\rightarrow 0 \\ \text{mul}(x, \mathbf{s}(y)) &\rightarrow \text{add}(x, \text{mul}(x, y))\end{aligned}$$

Derivation height:

- $\text{add}(0, \mathbf{s}(0))$: 2 ($\text{add}(0, \mathbf{s}(0)) \rightarrow \mathbf{s}(\text{add}(0, 0)) \rightarrow \mathbf{s}(0)$).
- $\text{mul}(\text{mul}(\mathbf{s}(\mathbf{s}(0)), \mathbf{s}(\mathbf{s}(\mathbf{s}(0)))) , 0)$: 15

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Traditional interpretations (first-order)

Idea:

- map every term s to $\llbracket s \rrbracket \in \mathbb{N}$
- make sure that $s \rightarrow t$ implies $\llbracket s \rrbracket > \llbracket t \rrbracket$

Then: $\llbracket s \rrbracket \geq \text{derivationheight}(s)$!

Approach:

- map every function that takes k arguments to a **monotonic** function in $\mathbb{N}^k \mapsto \mathbb{N}$
- make sure that $\llbracket \ell \rrbracket > \llbracket r \rrbracket$ for all rules $\ell \rightarrow r$

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Bounding derivation height with interpretations to \mathbb{N}

$$\begin{aligned}\text{add}(0, y) &\rightarrow y \\ \text{add}(\mathbf{s}(x), y) &\rightarrow \mathbf{s}(\text{add}(x, y))\end{aligned}$$

Let:

- $\llbracket 0 \rrbracket = 0$
- $\llbracket \mathbf{s}(x) \rrbracket = x + 1$
- $\llbracket \text{add}(x, y) \rrbracket = 1 + y + 2 * x$

We might initially be inclined to choose $\llbracket \text{add}(x, y) \rrbracket = x + y$ – but then we do not have that $\llbracket \ell \rrbracket > \llbracket r \rrbracket$ for the rules. Hence, the interpretation cannot exactly match the “meaning” of the rules:

Then:

$$\begin{aligned} \llbracket \text{add}(0, y) \rrbracket &= 1 + y > \llbracket y \rrbracket \\ \llbracket \text{add}(\mathbf{s}(x), y) \rrbracket &= 3 + y + 2 * x > 2 + y + 2 * x \\ &= \llbracket \mathbf{s}(\text{add}(x, y)) \rrbracket \end{aligned}$$

Hence: $\llbracket \text{add}(\mathbf{s}^n(0), \mathbf{s}^m(0)) \rrbracket = 1 + m + 2 * n$: linear!

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Monotonic algebras: definition

Given: a set \mathcal{A} with a well-founded ordering $>$ (for example: \mathbb{N})

Choose: a function $\llbracket \mathbf{f} \rrbracket$ from \mathcal{A}^k to \mathcal{A} for every \mathbf{f} of arity k

Define: for a given α mapping variables to \mathcal{A} :

- $\llbracket x \rrbracket = \alpha(x)$
- $\llbracket \mathbf{f}(s_1, \dots, s_k) \rrbracket = \llbracket \mathbf{f} \rrbracket(\llbracket s_1 \rrbracket, \dots, \llbracket s_k \rrbracket)$

Prove: $\llbracket \ell \rrbracket > \llbracket r \rrbracket$ for all rules $\ell \rightarrow r$, all α

In practice, since we quantify over α , we essentially view both sides as functions over a given set of variables. This is why we for instance write $\llbracket \text{add}(0, y) \rrbracket = 1 + y$ instead of $1 + \alpha(y)$.

Then: $\llbracket s \rrbracket > \llbracket t \rrbracket$ whenever $s \rightarrow_{\mathcal{R}} t$.

The most common example is to choose the set of natural numbers for \mathcal{A} , but we could also for instance choose the rational numbers (with $x > y$ if $x \geq y + 1$), or pairs of numbers as we will see later.

Consequence: if $\text{tonat}(a) > \text{tonat}(b)$ whenever $a > b$ then $\text{tonat}(\llbracket s \rrbracket) \geq \text{derivationheight}(s)$. (Here, we let tonat be a function that maps each element of \mathcal{A} to a natural number. If $\mathcal{A} = \mathbb{N}$ this is just the identity; if $\mathcal{A} = \mathbb{Q}$ this could for instance be rounding down.)

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Higher-order interpretations to \mathbb{N} : problems

Let’s extend this idea to higher-order rewriting. Here, we quickly run into the problem: what to do with partial applications? For example:

Suppose: $\llbracket \mathbf{s}(x) \rrbracket = x + 1$

Question: What is $\llbracket \mathbf{s} \rrbracket$?

Problem: behaviour matters!

$$\begin{aligned} \text{fold}(F, x, \llbracket \rrbracket) &\rightarrow \llbracket \rrbracket \\ \text{fold}(F, x, \text{cons}(y, l)) &\rightarrow \text{fold}(F, (F \cdot x \cdot y), l) \\ \text{add}(x, 0) &\rightarrow x \\ \text{add}(x, \mathbf{s}(y)) &\rightarrow \mathbf{s}(\text{add}(x, y)) \end{aligned}$$

- What is the derivation height if $F := \lambda x, y. \text{minimum}(x, y)$?
- What if: $F := \lambda x, y. \text{add}(x, y)$?

- What if: $F := \lambda x, y. \text{add}(x, \mathbf{s}(0))$?
- What if: $F := \lambda x, y. \text{add}(x, \mathbf{s}(\mathbf{s}(0)))$?
- What if: $F := \lambda x, y. \text{add}(x, x)$?

All in all, the consequences of using different functions for F cannot really be captured by a number.

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Proposal

Let's interpret terms of function type as functions!

More than that: for each type we have a possibly different interpretation domain. We only fix that function types are interpreted as *monotonic* functions:

Type interpretations:

- For every **base type** ι : a set \mathcal{A}_ι , ordering $>_\iota$ and quasi-ordering \geq_ι
- Define:

$$\begin{aligned} \langle \iota \rangle &= \mathcal{A}_\iota \\ \langle \sigma \Rightarrow \tau \rangle &= \text{“monotonic functions from } \langle \sigma \rangle \text{ to } \langle \tau \rangle \text{”} \\ F >_{\sigma \Rightarrow \tau} G &\text{ if } F(a) >_\tau G(a) \text{ for all } a \in \langle \sigma \rangle \\ F \geq_{\sigma \Rightarrow \tau} G &\text{ if } F(a) \geq_\tau G(a) \text{ for all } a \in \langle \sigma \rangle \end{aligned}$$

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Higher-order monotonic algebras: definition

(Difference to the first-order definition are indicated in **red**.)

Given: a *type interpretation function* as on the previous slide

Choose: a function $\llbracket \mathbf{f} \rrbracket$ in $\langle \sigma \rangle$ for every \mathbf{f} of type σ

Define: for a given α mapping variables to \mathcal{A} :

- $\llbracket x \rrbracket = \alpha(x)$
- $\llbracket \mathbf{f} \rrbracket = \llbracket \mathbf{f} \rrbracket$
- $\llbracket s \cdot t \rrbracket = \llbracket s \rrbracket(\llbracket t \rrbracket)$

(We're ignoring abstractions for now. We will get back to that later!)

Prove: $\llbracket \ell \rrbracket > \llbracket r \rrbracket$ for all rules $\ell \rightarrow r$, all α

In practice, since we quantify over α , we essentially view both sides as functions over a given set of variables.

Then: $\llbracket s \rrbracket > \llbracket t \rrbracket$ whenever $s \rightarrow_{\mathcal{R}} t$.

Consequence: if $\text{tonat}(a) > \text{tonat}(b)$ whenever $a > b$ then $\text{tonat}(\llbracket s \rrbracket) \geq \text{derivationheight}(s)$.

Note that of course, this is also a *termination* technique: if we have a bound on the number of steps, clearly this number is not infinite.

Example:

$$\begin{aligned}
 [] &:: \text{list} \\
 \text{cons} &:: \text{nat} \Rightarrow \text{list} \Rightarrow \text{list} \\
 \text{map} &:: (\text{nat} \Rightarrow \text{nat}) \Rightarrow \text{list} \Rightarrow \text{list} \\
 \text{map}(F, []) &\rightarrow [] \\
 \text{map}(F, \text{cons}(x, l)) &\rightarrow \text{cons}(F \cdot x, \text{map}(F, l))
 \end{aligned}$$

Choose: $\mathcal{A}_\iota = \mathbb{N}$ for all ι

$$\begin{aligned}
 [[]] &= 0 \\
 [\text{cons}](x, y) &= x + y + 1 \\
 [\text{map}](F, x) &= (x + 1) * F(x)
 \end{aligned}$$

Monotonicity: holds. (We can easily see that, for example, if $x > y$ then $[\text{map}](F, x) > [\text{map}](F, y)$, and if $F(x) > G(x)$ for all x then $[\text{map}](F, x) > [\text{map}](G, x)$.)

Example

$$\begin{aligned}
 [[]] &= 0 \\
 [\text{cons}](x, y) &= x + y + 1 \\
 [\text{map}](F, x) &= (x + 1) * F(x) + 1
 \end{aligned}$$

Goal 1:

$$[[\text{map}(F, [])]] > [[]]$$

That is:

$$(0 + 1) * F(0) + 1 > 0$$

Which is certainly true because $1 > 0$.

Goal 2:

$$[[\text{map}(F, \text{cons}(x, l))]] > [[\text{cons}(F \cdot x, \text{map}(F, l))]]$$

That is:

$$\begin{aligned}
 ((x + l + 1) + 1) * F(x + l + 1) + 1 &> \\
 F(x) + ((l + 1) * F(l) + 1) + 1 &
 \end{aligned}$$

Simplifying the arithmetic, this is:

$$\begin{aligned}
 x * F(x + l + 1) + l * F(x + l + 1) + F(x + l + 1) + F(x + l + 1) + 1 &> \\
 F(x) + l * F(l) + F(l) + 1 &
 \end{aligned}$$

Let's reorganise that a bit!

$$\begin{aligned}
 x * F(x + l + 1) &+ l * F(x + l + 1) &+ F(x + l + 1) &+ F(x + l + 1) &+ 1 \\
 > &+ l * F(l) &+ F(x) &+ F(l) &+ 1
 \end{aligned}$$

Now observe that F is *monotonic*. So for instance $F(x + l + 1) > F(x)$. Hence we quickly see that this inequality indeed holds.

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Exercise

Given:

```

[] :: list
cons :: nat ⇒ list ⇒ list
filter :: (nat ⇒ bool) ⇒ list ⇒ list
helper :: bool ⇒ nat ⇒ list ⇒ list

filter(F, []) → []
filter(F, cons(x,l)) → helper(F · x, x, filter(F, l))
helper(true, x, l) → cons(x, l)
helper(false, x, l) → l

```

Task: show that the following interpretation suffices:

$$\begin{array}{ll}
 [[]] & = 0 & [\text{true}] & = 1 \\
 [\text{cons}](x, y) & = x + y + 1 & [\text{false}] & = 0 \\
 [\text{helper}](b, x, y) & = b + x + y + 1 \\
 [\text{filter}](F, x) & = (x + 1) * (F(x) + 1)
 \end{array}$$

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Bonus exercise

Given:

```

[] :: list
cons :: nat ⇒ list ⇒ list
zip :: (nat ⇒ nat) ⇒ list ⇒ list

zip(F, [], l) = l
zip(F, l, []) = l
zip(F, cons(x, l), cons(y, q)) = cons(F · x · y, zip(F, l, q))

```

Task: find an interpretation that orients these rules!

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Abstraction

Discussion: what should be the interpretation of $\lambda x.s$?

Naive choice: $x \mapsto \llbracket s \rrbracket$

Problem: the naive interpretation for $\lambda x.s$ is not monotonic if x does not occur in s ! For example, this choice would let $\llbracket \lambda x.0 \rrbracket$ be the **constant** function mapping everything to 0 – and thus, it would not be an element of $(\text{nat} \Rightarrow \text{nat})$.

Solution: for each σ, τ , a function $\text{makesm}_{\sigma, \tau}$:

- Input: a monotonic **or** constant function from (σ) to (τ)
- Output: a monotonic function from (σ) to (τ)

- $\mathbf{makesm}_{\sigma,\tau}$ should itself be monotonic!
- we need to have $\llbracket (\lambda x.s) \cdot t \rrbracket > s[x := t]$

The use of \mathbf{makesm} functions may be confusing at first – but essentially, all that this means is that we choose a systematic way of turning a given abstraction into a monotonic function. And in practice, we can usually find a way to define a class of \mathbf{makesm} functions that allows us to *almost* map $\lambda x.s$ to $x \mapsto \llbracket s \rrbracket$ if $x \in FV(s)$ – just adding a cost for the β -reduction. This is demonstrated for $\mathcal{A}_{\mathbf{nat}} = \mathbb{N}$ below.

Example: (for $\sigma, \tau = \mathbf{nat}$ and $\mathcal{A}_{\mathbf{nat}} = \mathbb{N}$):

- if F is constant, then $\mathbf{makesm}_{\sigma,\tau}(F) = x \mapsto F(x) + x + 1$
- otherwise $\mathbf{makesm}_{\sigma,\tau}(F) = x \mapsto F(x) + 1$

This definition works very nicely in practice. The only difficulty is to prove that the above \mathbf{makesm} function is indeed monotonic; in particular, if F is monotonic in x and G is constant, we must show that that $F >_{\mathbf{nat} \Rightarrow \mathbf{nat}} G$ implies that also $\mathbf{makesm}(F) >_{\mathbf{nat} \Rightarrow \mathbf{nat}} \mathbf{makesm}(G)$. To see that this holds, we make the observation that in the natural numbers, if F is a monotonic function, then $F(x+1) > F(x)$, so $F(x+1) \geq F(x) + 1$; by induction, we see that $F(n) \geq F(0) + n$. In a constant function, $G(n) = G(0)$. Thus we see: for all n : $F(n) \geq F(0) + n > G(0) + n = G(n) + n$.

This idea can be generalised to all types, but it takes a bit more definition effort; for example, if $\sigma = \tau = \mathbf{nat} \Rightarrow \mathbf{nat}$ we let $\mathbf{makesm}_{\sigma,\tau}(F) = (G, x) \mapsto F(G, x) + 1$ if F is monotonic in its first argument (G), and $(G, x) \mapsto F(G, x) + G(0) + 1$ if F is constant in its first argument.

2. Tuple interpretations

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An observation

Consider:

- $\llbracket \text{add}(s^n(0), s^m(0)) \rrbracket = 1 + m + 2 * n$
- actual cost of reduction: $n + 1$
- size of normal form: $n + m$
- This does raise the question: are we actually giving a bound to the *sum* of cost and size by using interpretations to \mathbb{N} ?

Idea: separate cost and size already in the interpretation!

Mechanism: map to \mathbb{N}^2 instead of \mathbb{N} .

We let $\langle x, y \rangle > \langle x', y' \rangle$ if $x > x'$ and $y \geq y'$.

Note: we can choose $\text{tonat}(\langle x, y \rangle) = x$. That is, if $a > b$ in \mathbb{N}^2 then $\text{tonat}(a) > \text{tonat}(b)$ – so if we can express $\llbracket s \rrbracket$ as an element $\langle x, y \rangle$ of \mathbb{N}^2 , then x gives a bound on the derivation height of s . We will refer to the first element of the tuple as the **cost component** of the tuple.

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Separating cost and size

$$\begin{aligned} \text{add}(0, y) &\rightarrow y \\ \text{add}(s(x), y) &\rightarrow s(\text{add}(x, y)) \end{aligned}$$

Let:

$$\begin{array}{lcl} \llbracket 0 \rrbracket & = \langle & \text{cost} \qquad \qquad \text{size} \\ & & 0 \qquad \qquad \qquad 0 \rangle \\ \llbracket s(x) \rrbracket & = \langle & x_{\text{cost}} \qquad \qquad x_{\text{size}} + 1 \rangle \\ \llbracket \text{add}(x, y) \rrbracket & = \langle & x_{\text{cost}} + y_{\text{cost}} + x_{\text{size}} \qquad x_{\text{size}} + y_{\text{size}} \rangle \end{array}$$

Then:

$$\begin{aligned} \llbracket \text{add}(0, y) \rrbracket &= \langle 1 + y_1, y_2 \rangle \\ &> \langle y_1, y_2 \rangle &= \llbracket y \rrbracket \\ \llbracket \text{add}(s(x), y) \rrbracket &= \langle 2 + x_1 + y_1 + x_2, 1 + x_2 + y_2 \rangle \\ &> \langle 1 + x_1 + y_1 + x_2, 1 + x_2 + y_2 \rangle &= \llbracket s(\text{add}(x, y)) \rrbracket \end{aligned}$$

Hence: $\llbracket \text{add}(s^n(0), s^m(0)) \rrbracket = \langle 1 + n, n + m \rangle$: precise! (And also intuitive.)

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When interpretations to \mathbb{N} are Not Great

$$a(b(x)) \rightarrow b(a(x))$$

Let:

- $\llbracket \mathbf{a}(x) \rrbracket = 2 * x$
- $\llbracket \mathbf{b}(x) \rrbracket = x + 1$
- $\llbracket \epsilon \rrbracket = 0$

Then:

$$\llbracket \mathbf{a}(\mathbf{b}(x)) \rrbracket = 2 + 2 * x > 1 + 2 * x = \llbracket \mathbf{b}(\mathbf{a}(x)) \rrbracket$$

Hence: $\llbracket \mathbf{a}^n(\mathbf{b}^m(\epsilon)) \rrbracket = 2^n * m$: exponential!

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Separating cost and size

$$\mathbf{a}(\mathbf{b}(x)) \rightarrow \mathbf{b}(\mathbf{a}(x))$$

Let:

	cost	size
$\llbracket \mathbf{a}(x) \rrbracket$	$\langle x_{cost} + x_{size} \rangle$	$\langle x_{size} \rangle$
$\llbracket \mathbf{b}(x) \rrbracket$	$\langle x_{cost} \rangle$	$\langle x_{size} + 1 \rangle$
$\llbracket \epsilon \rrbracket$	$\langle 0 \rangle$	$\langle 0 \rangle$

Then:

$$\llbracket \mathbf{a}(\mathbf{b}(x)) \rrbracket = \langle x_1 + x_2 + 1, x_2 + 1 \rangle > \langle x_1 + x_2, x_2 + 1 \rangle = \llbracket \mathbf{b}(\mathbf{a}(x)) \rrbracket$$

Hence: $\llbracket \mathbf{a}^n(\mathbf{b}^m(\epsilon)) \rrbracket = (n * m, m)$: precise!

Of course, we can't always get precision. But we invariably get tighter interpretations by using tuples than single numbers.

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Tuple interpretations

Definition: monotonic algebras with $\mathcal{A}_\iota = \mathbb{N}^{K[\iota]}$ for all ι (where $K[\iota]$ is a positive integer for all ι).

\implies both for first- and higher-order!

This is a specific implementation of a well-known method (monotonic algebras), that adds a surprising amount of power over other variations. In the bigger picture, tuple interpretations can be seen as a generalisation of the method of *matrix interpretations*: this method also considers tuples over \mathbb{N} as the interpretation domain, but restrict the shape of the interpretation functions $\llbracket \mathbf{f} \rrbracket$.

Of course, there is no reason to stop here. We could have tuples over *other* sets than \mathbb{N} – for example, using the set of integers \mathbb{Z} as the second set in the component (as only the first needs to admit a wellfounded ordering), a set such as $\mathbb{N} \cup \{\infty\}$, or even some impromptu set $\{a, b, c\}$ with $a > b$ and $a > c$ but b, c not comparable. There are uses for all these examples. We could also use tuples only for *some* base types, and still allow, for instance, a base type $\text{list}(\mathbb{N} \Rightarrow \mathbb{N})$ to be mapped to a function space such as $(\mathbb{N} \Rightarrow \mathbb{N})$. However, for this lecture, we will limit interest to tuples of the form \mathbb{N}^k .

Example sort interpretations:

- $\{\text{nat}\} = \mathbb{N}^2$ (cost, size of normal form)
- $\{\text{list}\} = \mathbb{N}^3$ (cost, list length, size of greatest element)

- $\{\text{bool}\} = \mathbb{N}^1$ (cost)

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Example: interpreting list functions

$$\begin{aligned} \text{append}([], l) &\rightarrow l \\ \text{append}(\text{cons}(x, l), q) &\rightarrow \text{cons}(x, \text{append}(l, q)) \\ \text{sum}([]) &\rightarrow 0 \\ \text{sum}(\text{cons}(x, l)) &\rightarrow \text{add}(x, \text{sum}(l)) \end{aligned}$$

Interpretations:

- $\{\text{list}\} = \mathbb{N}^3$ (cost, list length, maximum element)
- $\llbracket [] \rrbracket = \langle 0, 0, 0 \rangle$
- $\llbracket \text{cons}(x, l) \rrbracket = \langle x_{\text{cost}} + l_{\text{cost}}, l_{\text{len}} + 1, \max(x_{\text{size}}, l_{\text{max}}) \rangle$
- $\llbracket \text{append}(l, q) \rrbracket = \langle \text{cost}, \text{length}, \text{maximum} \rangle$, where:
 - maximum = $\max(l_{\text{max}}, q_{\text{max}})$
 - length = $l_{\text{len}} + q_{\text{len}}$
 - cost = $l_{\text{cost}} + q_{\text{cost}} + l_{\text{len}} + 1$
- $\llbracket \text{sum}(l) \rrbracket = \langle \text{cost}, \text{size} \rangle$, where:
 - size = $l_{\text{len}} * l_{\text{max}}$
 - cost = $l_{\text{cost}} + 2 * l_{\text{len}} + l_{\text{len}} * l_{\text{max}} + 1$

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Higher-order tuple interpretations: an example

$$\begin{aligned} [] &:: \text{list} \\ \text{cons} &:: \mathbb{N} \Rightarrow \text{list} \Rightarrow \text{list} \\ \text{map} &:: (\mathbb{N} \Rightarrow \mathbb{N}) \Rightarrow \text{list} \Rightarrow \text{list} \\ \text{map}(F, []) &\rightarrow [] \\ \text{map}(F, \text{cons}(x, l)) &\rightarrow \text{cons}(F \cdot x, \text{map}(F, l)) \end{aligned}$$

Let:

- $\llbracket [] \rrbracket = \langle 0, 0, 0 \rangle$
- $\llbracket \text{cons}(x, l) \rrbracket = \langle x_{\text{cost}} + l_{\text{cost}}, l_{\text{len}} + 1, \max(x_{\text{size}}, l_{\text{max}}) \rangle$
- $\llbracket \text{map}(F, l) \rrbracket = \langle \text{cost}, \text{length}, \text{maximum} \rangle$, where:
 - length: l_{len}
 - maximum: $F(\langle l_{\text{cost}}, l_{\text{max}} \rangle)_s$
 - cost: $(l_{\text{len}} + 1) * (F(\langle l_{\text{cost}}, l_{\text{max}} \rangle)_{\text{cost}} + 1)$

Exercise

1. Find an interpretation, with $(\text{nat}) = \mathbb{N}^2$, for the following system:

$$\begin{aligned} \text{minus}(x, 0) &\rightarrow x \\ \text{minus}(\mathbf{s}(x), \mathbf{s}(y)) &\rightarrow \text{minus}(x, y) \\ \text{quot}(0, \mathbf{s}(y)) &\rightarrow 0 \\ \text{quot}(\mathbf{s}(x), \mathbf{s}(y)) &\rightarrow \mathbf{s}(\text{quot}(\text{minus}(x, y), \mathbf{s}(y))) \end{aligned}$$

Warning: do not take $x_{\text{size}} - y_{\text{size}}$ for the size of $\text{minus}(x, y)$! Doing this would break the monotonicity requirement: we must have $\llbracket \text{minus}(a, b) \rrbracket > \llbracket \text{minus}(a, c) \rrbracket$ if $b > c$, which implies $\llbracket \text{minus}(a, b) \rrbracket_{\text{size}} \geq \llbracket \text{minus}(a, c) \rrbracket_{\text{size}}$ if $b_{\text{cost}} > c_{\text{cost}}$ and $b_{\text{size}} \geq c_{\text{size}}$.

Side note: the fact that we can do this at all illustrates the power of tuple interpretations. This was a motivating example for dependency pairs, since it cannot be handled with any well-founded ordering that has $\text{minus}(x, y) \succeq y$. Thus, termination *cannot* be proved using RPO or interpretations to \mathbb{N} , nor can it be proved with a method like matrix interpretations due to the duplication of x in the last rule. Yet, here we do not only prove its termination, but also find a bound to its complexity.

2. Find an interpretation for the following HTRS, where $\text{zip} :: (\text{nat} \Rightarrow \text{nat}) \Rightarrow \text{list} \Rightarrow \text{list}$.

$$\begin{aligned} \text{zip}(F, [], l) &= l \\ \text{zip}(F, l, []) &= l \\ \text{zip}(F, \text{cons}(x, l), \text{cons}(y, q)) &= \text{cons}(F \cdot x \cdot y, \text{zip}(F, l, q)) \end{aligned}$$

A more challenging higher-order tuple interpretation

$$\begin{aligned} \text{fold}(F, x, []) &\rightarrow [] \\ \text{fold}(F, x, \text{cons}(y, l)) &\rightarrow \text{fold}(F, (F \cdot x \cdot y), l) \end{aligned}$$

Interpretation:

$$\llbracket \text{fold}(F, x, l) \rrbracket = \langle \text{cost}, \text{size} \rangle$$

Where:

- $\text{cost} = 1 + l_{\text{cost}} + F(\langle 0, 0 \rangle)_{\text{cost}} + \text{Helper}[F, \langle l_{\text{cost}}, l_{\text{max}} \rangle]^{l_{\text{len}}(x)}_{\text{cost}}$
- $\text{size} = \text{Helper}[F, \langle l_{\text{cost}}, l_{\text{max}} \rangle]^{l_{\text{len}}(x)}_{\text{size}}$
- And $\text{Helper}[F, y] = x \mapsto \langle F(x, y)_{\text{cost}}, \max(x_{\text{size}}, F(x, y)_{\text{size}}) \rangle$.

A more challenging higher-order tuple interpretation

$$\begin{aligned} \text{add}(0, y) &\rightarrow y \\ \text{add}(\mathbf{s}(x), y) &\rightarrow \text{add}(x, \mathbf{s}(y)) \\ \text{fold}(F, x, []) &\rightarrow [] \\ \text{fold}(F, x, \text{cons}(y, l)) &\rightarrow \text{fold}(F, (F \cdot x \cdot y), l) \\ \text{sum}(l) &\rightarrow \text{fold}(\lambda x. \lambda y. \text{add}(x, y), 0, l) \end{aligned}$$

Method: Plug $\llbracket \lambda x. \lambda y. \text{add}(x, y) \rrbracket$ into the interpretation for **fold**.

Interpreting λ : use $\text{makesm}_{\ell, \sigma_1 \Rightarrow \dots \Rightarrow \sigma_m \Rightarrow \kappa} =$

$$\begin{cases} (F, x, y_1, \dots, y_m) \mapsto (F(x, \vec{y})_1 + 1 + x_1, F(x, \vec{y})_2, \dots, F(x, \vec{y})_{K[\kappa]}) & \text{if } F \text{ is constant} \\ (F, x, y_1, \dots, y_m) \mapsto (F(x, \vec{y})_1 + 1, F(x, \vec{y})_2, \dots, F(x, \vec{y})_{K[\kappa]}) & \text{if } F \text{ is monotonic} \end{cases}$$

3. Complexity notions

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Derivational and runtime complexity (first-order)

Derivational complexity:

$n \mapsto$ “maximum derivation height for a term of size n ”

Downside: can easily get large; e.g.: `mul(mul(mul(mul(s(s(0))), s(s(0))), s(s(0))), s(s(0))), s(s(0)))`

Runtime complexity:

$n \mapsto$ “maximum derivation height for a **basic** term of size n ”

Basic term: `function(data, ..., data)`

Example: `mul(s(s(s(s(s(0))))), s(s(s(s(s(s(0))))))`

Connection with computational complexity: depends

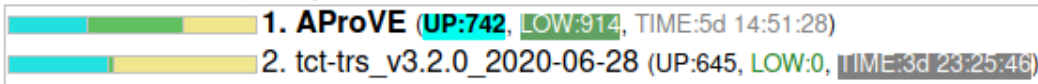
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Termination (and complexity) competition

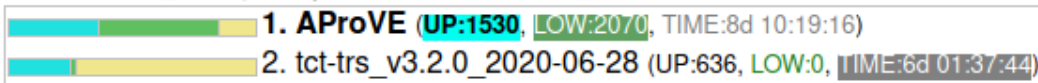
In the annual termination competition, there are categories for both runtime and derivational complexity of first-order term rewriting (both with a general reduction strategy, and focused on innermost reduction).

Complexity Analysis

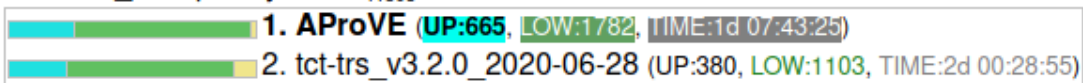
Derivational_Complexity: TRS ₄₁₄₉₉



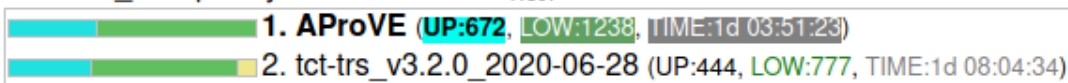
Derivational_Complexity: TRS Innermost ₄₁₅₀₀



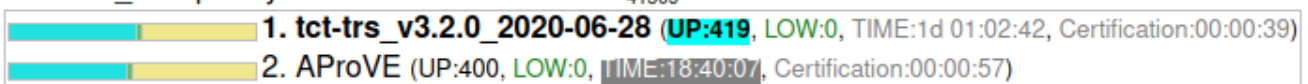
Runtime_Complexity: TRS ₄₁₅₀₈



Runtime_Complexity: TRS Innermost ₄₁₅₀₇



Runtime_Complexity: TRS Innermost Certified ₄₁₅₀₉



Complexity of higher-order term rewriting

Open question: do derivational and runtime complexity even make sense for higher-order rewriting?

$$\begin{aligned} \text{fold}(F, x, []) &\rightarrow [] \\ \text{fold}(F, x, \text{cons}(y, l)) &\rightarrow \text{fold}(F, (F \cdot x \cdot y), l) \end{aligned}$$

Recall:

- What if: $F := \lambda x, y. \text{minimum}(x, y)$?
- What if: $F := \lambda x, y. \text{add}(x, y)$?
- What if: $F := \lambda x, y. \text{add}(x, x)$?

Higher-order derivational complexity?

Idea: naively extend the definition of derivational complexity

Result:

$$\begin{aligned} \text{add}(x, 0) &\rightarrow x \\ \text{add}(x, \text{s}(y)) &\rightarrow \text{s}(\text{add}(x, y)) \end{aligned}$$

- $(\lambda x. \text{add}(x, x)) \cdot (\text{s}(\text{s}(0)))$
- $(\lambda x. \text{add}(x, x)) \cdot ((\lambda x. \text{add}(x, x)) \cdot (\text{s}(\text{s}(0))))$
- $(\lambda x. \text{add}(x, x)) \cdot ((\lambda x. \text{add}(x, x)) \cdot ((\lambda x. \text{add}(x, x)) \cdot (\text{s}(\text{s}(0)))))$
- ...

Conclusion: exponential complexity at a minimum, even for very simple systems.

Runtime complexity: a simple extension

Runtime complexity:

$n \mapsto$ “maximum derivation height for a **basic** term of size n ”

Basic term: `function(data, ..., data)`

Question: is it interesting to look at λ -functions over constructors?

- `map($\lambda x. \text{s}(x)$, some lst)`?
- `maketree($\lambda x_{\text{nat}}, y_{\text{tree}}. \text{node}(x, y, y)$, some natural number)`

A notion of runtime complexity like this would be well-defined, and give reasonable bounds. However, where runtime complexity makes sense in first-order rewriting if we are interested in “start terms” for a program, the concept of instantiating higher-order functions by constructors or functions that are built from constructors doesn’t seem to have much practical relevance.

Choice: data must be a **first-order** term.

Thus, we let the start terms for higher-order runtime complexity analysis be *exactly the same* as those for runtime analysis of first-order term rewriting. Yet, higher-order function calls may arise during the evaluation of the start terms, so their analysis is still needed. This actually seems representative of full program analysis.

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Higher-order runtime complexity example

```
add(0, y) → y
add(s(x), y) → add(x, s(y))
fold(F, x, []) → []
fold(F, x, cons(y, l)) → fold(F, (F · x · y), l)
sum(l) → fold(λx.λy.add(x, y), 0, l)
```

Basic terms:

- `add(s(s(s(s(s(0))))), s(s(s(s(s(s(0))))))`
- `sum(cons(s(s(0)), cons(0, cons(s(s(0))), [])))`

Runtime complexity: $n \mapsto \mathcal{O}(n^2)$ (actually: length * max)

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Exercises

1. Compute a bound on the runtime complexity of the following system.

```
map(F, []) → []
map(F, cons(x, l)) → cons(F · x, map(F, l))
doublemap(l) → map(double, l)
double(0) → 0
double(s(x)) → s(s(double(x)))
```

2. Compute a bound on the runtime complexity of the following system.

```
add(x, 0) → x
add(x, s(y)) → s(add(x, y))
zip(F, [], l) = l
zip(F, l, []) = l
zip(F, cons(x, l), cons(y, q)) = cons(F · x · y, zip(F, l, q))
zipadd(l, q) → zip(λx.λy.add(y, x), l, q)
```

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A higher-order complexity notion?

Extending the first-order runtime complexity notion to higher-order rewriting is a good start, but it doesn't really capture the higher-order nature. And indeed, tuple interpretations give us much more

information, that we could use for both time and space bounds. Even just sticking to time (or: computation cost) bounds, it would be nice if we could express the complexity of functions, rather than full programs; for example:

Idea:

- complexity of **map** is $\mathcal{O}(n * F(n))$?
- complexity of **fold** is $\mathcal{O}(F^n(n))$?

However, this is speculative; there is no clear definition of what it would mean. We could likely define something, but would it be useful?

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Basic Feasible Functions

But there *is* a higher-order version of PTIME! This is defined in terms of Turing Machines.

Idea:

- Oracle Turing Machines: these take n functions, k binary words
- to compute function i :
 - copy input to tape i
 - go to special state
 - output is written on tape $n + i$
- \implies function **cost** is assumed zero, but function **output size** is important
- Question: is the execution time limited by a **higher-order polynomial** over $F_1, \dots, F_n, w_1, \dots, w_k$?

Relevance: this is exactly determined by the existence of a higher-order polynomially-bounded tuple interpretation, provided we impose some restrictions on the interpretation of binary words.