
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?

32 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 $n + i$
 - \Rightarrow 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.

1. Monotonic algebras

²

Derivation height

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

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

Derivation height:

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

Traditional interpretations (first-order)

Idea:

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

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

Approach:

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

Bounding derivation height with interpretations to \mathbb{N}

$$\text{add}(\mathbf{s}(x), y) \rightarrow \mathbf{s}(\text{add}(x, y))$$

Let:

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

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.

29 Higher-order runtime complexity example

$$\begin{array}{lcl} \text{add}(0, y) & \rightarrow & y \\ \text{add}(\mathbf{s}(x), y) & \rightarrow & \text{add}(x, \mathbf{s}(y)) \\ \text{fold}(\mathbf{F}, x, \mathbf{l}) & \rightarrow & \mathbf{l} \\ \text{fold}(\mathbf{F}, x, \text{const}(y, l)) & \rightarrow & \text{fold}(\mathbf{F}, (F \cdot x \cdot y), l) \\ \text{sum}(\mathbf{l}) & \rightarrow & \text{fold}(\lambda x. \lambda y. \text{add}(x, y), 0, l) \end{array}$$

Basic terms:

- $\text{add}(\mathbf{s}(\mathbf{s}(\mathbf{s}(\mathbf{s}(0))))), \mathbf{s}(\mathbf{s}(\mathbf{s}(\mathbf{s}(\mathbf{s}(0))))))$
- $\text{sum}(\text{cons}(\mathbf{s}(0)), \text{cons}(0, \text{cons}(\mathbf{s}(\mathbf{s}(0))), \mathbf{l})))$

Runtime complexity: $n \rightarrow O(n^2)$ (actually: length * max)

30 Exercises

Exercises

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

$$\begin{array}{lcl} \text{map}(\mathbf{F}, \mathbf{l}) & \rightarrow & \mathbf{l} \\ \text{map}(\mathbf{F}, \text{cons}(x, \mathbf{l})) & \rightarrow & \text{cons}(F \cdot x, \text{map}(\mathbf{F}, \mathbf{l})) \\ \text{doublemap}(\mathbf{l}) & \rightarrow & \text{map}(\text{double}, \mathbf{l}) \\ \text{double}(\mathbf{0}) & \rightarrow & 0 \\ \text{double}(\mathbf{s}(x)) & \rightarrow & \mathbf{s}(\text{double}(x)) \end{array}$$

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

$$\begin{array}{lcl} \text{add}(x, 0) & \rightarrow & x \\ \text{add}(x, \mathbf{s}(y)) & \rightarrow & \mathbf{s}(\text{add}(x, y)) \\ \text{zip}(\mathbf{F}, \mathbf{l}, \mathbf{l}) & = & \mathbf{l} \\ \text{zip}(\mathbf{F}, \mathbf{l}, \mathbf{l}) & = & \mathbf{l} \\ \text{zip}(\mathbf{F}, \mathbf{l}, \mathbf{l}) & = & \mathbf{l} \\ \text{zip}(\mathbf{F}, \text{cons}(x, \mathbf{l}), \text{cons}(y, \mathbf{l})) & = & \text{cons}(F \cdot x \cdot y, \text{zip}(\mathbf{F}, \mathbf{l}, \mathbf{l})) \\ \text{ziipaad}(\mathbf{l}, \mathbf{l}) & \rightarrow & \text{zip}(\lambda x. \lambda y. \text{add}(y, x), \mathbf{l}, \mathbf{l}) \end{array}$$

31 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

26 Complexity of higher-order term rewriting

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

$$\begin{array}{lcl} \text{fold}(F, x, \text{cons}(y, l)) & \rightarrow & \text{f}\text{old}(F, x \cdot y, l) \\ \text{fold}(F, x, \text{cons}(y, l)) & \rightarrow & \text{f}\text{old}(F, (F \cdot x \cdot y), l) \end{array}$$

Recall:

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

27 Higher-order derivational complexity?

Idea: naively extend the definition of derivational complexity

Result:

$$\begin{array}{ll} \text{add}(x, 0) & \rightarrow x \\ \text{add}(x, s(y)) & \rightarrow s(\text{add}(x, y)) \\ \\ \bullet & (\lambda x. \text{add}(x, x)) \cdot (s(0)) \\ \bullet & (\lambda x. \text{add}(x, x)) \cdot ((\lambda x. \text{add}(x, x)) \cdot (s(0))) \\ \bullet & (\lambda x. \text{add}(x, x)) \cdot ((\lambda x. \text{add}(x, x)) \cdot ((\lambda x. \text{add}(x, x)) \cdot (s(s(0)))))) \\ \dots & \end{array}$$

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

28 Runtime complexity: a simple extension

Runtime complexity:

$n \rightarrow$ “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`($\lambda x. s(x)$, some `lst`)?

- `maketree`($\lambda^n \text{nat}, \text{tree}.\text{node}(x, y, y)$, 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.

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

Then:

$$\begin{array}{ll} \text{add}(0, y) & = 1 + y & > y \\ \text{add}(s(x), y) & = 3 + y + 2 * x & > 2 + y + 2 * x \\ & & = s(\text{add}(x, y)) \end{array}$$

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

5

Monotonic algebras: definition

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

Choose: a function $[f]$ from \mathcal{A}^k to \mathcal{A} for every f of arity k

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

- $x = \alpha(x)$
- $f(s_1, \dots, s_k) = [f](s_1, \dots, s_k)$

Prove: $\ell > r$ 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 $\text{add}(0, y) = 1 + y$ instead of $1 + \alpha(y)$.

Then: $s > t$ whenever $s \rightarrow_R t$.

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 $x > y + 1$), or pairs of numbers as we will see later.

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

6

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: $s(x) = x + 1$

Question: What is s ?

Problem: behaviour matters!

$$\begin{array}{ll} \text{fold}(F, x, \text{nil}) & \rightarrow \text{nil} \\ \text{fold}(F, x, \text{cons}(y, l)) & \rightarrow \text{fold}(F, (F \cdot x \cdot y), l) \\ \text{add}(x, 0) & \rightarrow x \\ \text{add}(x, s(y)) & \rightarrow s(\text{add}(x, y)) \end{array}$$

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

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

Proposal

7

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:

$$\begin{aligned} \sigma \Rightarrow \tau &= \mathcal{A}_\tau \\ F >_{\sigma \Rightarrow \tau} G &\text{ if } F(a) >_\tau G(a) \text{ for all } a \in \sigma \\ F \geq_{\sigma \Rightarrow \tau} G &\text{ if } F(a) \geq_\tau G(a) \text{ for all } a \in \sigma \end{aligned}$$

Higher-order monotonic algebras: definition

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

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

Choose: a function $[f]$ in σ for every f of type σ

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

- $x = \alpha(x)$
- $f = [f]$
- $s \cdot t = s(t)$

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

Prove: $\ell > r$ 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: $s > t$ whenever $s \rightarrow_R t$.

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

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.

3. Complexity notions

24

Derivational and runtime complexity (first-order)

Derivational complexity:

$n \rightarrow$ "maximum derivation height for a term of size n "

Downside: can easily get large; e.g.: $\text{mul}(\text{mul}(\text{mul}(\text{mul}(s(s(s(s(0))))), s(s(0)))), s(s(0))), s(s(0)))$

Runtime complexity:

$n \rightarrow$ "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(0)))))))`

Connection with computational complexity: depends

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

1. **APROVE** (UP:742, LOW:914, TIME:5d 14:51:28)
2. tct-trs_v3.2.0_2020-06-28 (UP:845, LOW:0, TIME:3d 23:25:46)

Derivational_Complexity: TRS Innermost 41500

1. **APROVE** (UP:1530, LOW:2070, TIME:8d 10:19:16)
2. tct-trs_v3.2.0_2020-06-28 (UP:636, LOW:0, TIME:6d 01:31:44)

Runtime_Complexity: TRS 41508

1. **APROVE** (UP:655, LOW:1782, TIME:1d 07:43:25)
2. tct-trs_v3.2.0_2020-06-28 (UP:380, LOW:0, TIME:2d 00:28:55)

Runtime_Complexity: TRS Innermost 41507

1. **APROVE** (UP:672, LOW:1288, TIME:1d 03:51:23)
2. tct-trs_v3.2.0_2020-06-28 (UP:444, LOW:777, TIME:1d 08:04:34)

Runtime_Complexity: TRS Innermost Certified 41508

1. **tct-trs_v3.2.0_2020-06-28** (UP:419, LOW:0, TIME:1d 01:02:42, Certification:0:00:39)
2. **APROVE** (UP:400, LOW:0, TIME:1d 18:40:07, Certification:0:00:57)

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

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

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

9

Example:

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

Choose: $A_i = \mathbb{N}$ for all i

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

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

10

Example

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

Goal 1:

$$\text{map}(F, \text{list}) > \text{list}$$

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:

$$\frac{((x + l + 1) + 1) * F(x + l + 1) + 1}{F(x) + ((l + 1) * F(l) + 1) + 1} >$$

Simplifying the arithmetic, this is:

$$\frac{x * F(x + l + 1) + l * F(x + l + 1) + F(x + l + 1) + F(x + l + 1) + 1}{F(x) + l * F(l) + F(l) + 1} >$$

Let's reorganise that a bit!

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

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

11 Exercise

Given:

$$\begin{aligned} \emptyset &\rightarrow \text{list} \\ \text{cons} &\rightarrow \text{nat} \Rightarrow \text{list} \Rightarrow \text{list} \\ \text{filter} &\rightarrow (\text{nat} \Rightarrow \text{bool}) \Rightarrow \text{list} \Rightarrow \text{list} \\ \text{helper} &\rightarrow \text{bool} \Rightarrow \text{nat} \Rightarrow \text{list} \Rightarrow \text{list} \\ \text{filter}(\textcolor{blue}{F}, \emptyset) &\rightarrow \emptyset \\ \text{filter}(\textcolor{blue}{F}, \text{cons}(x, l)) &\rightarrow \text{helper}(\textcolor{blue}{F} \cdot x, x, \text{filter}(\textcolor{blue}{F}, l)) \\ \text{helper}(\text{true}, x, l) &\rightarrow \text{cons}(x, l) \\ \text{helper}(\text{false}, x, l) &\rightarrow \textcolor{blue}{l} \end{aligned}$$

Task: show that the following interpretation suffices:

$$\begin{aligned} [\text{cons}](x, y) &= \emptyset \\ [\text{helper}](b, x, y) &= x + y + 1 \\ [\text{filter}](\textcolor{blue}{F}, x) &= (x + 1) * (\textcolor{blue}{F}(x) + 1) \end{aligned}$$

12 Bonus exercise

Given:

$$\begin{aligned} \emptyset &\rightarrow \text{list} \\ \text{cons} &\rightarrow \text{nat} \Rightarrow \text{list} \Rightarrow \text{list} \\ \text{zip} &\rightarrow (\text{nat} \Rightarrow \text{nat}) \Rightarrow \text{list} \Rightarrow \text{list} \\ \text{zip}(\textcolor{blue}{F}, \emptyset) &= \textcolor{blue}{l} \\ \text{zip}(\textcolor{blue}{F}, \text{cons}(x, l)) &= \text{cons}(\textcolor{blue}{F} \cdot x, y, \text{zip}(\textcolor{blue}{F}, l, q)) \end{aligned}$$

Task: find an interpretation that orients these rules!

13

Abstraction

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

Naive choice: $x \rightarrow s$

Problem: the naive interpretation for $\lambda x.s$ is not monotonic if x does not occur in s ! For example, this choice would let $\lambda x.0$ 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 $\text{makes}_{\sigma, \tau}$:

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

21

Exercise

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

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

Warning: do not take $\text{x_size} = \text{y_size}$ for the size of $\text{minus}(x, y)$! Doing this would break the monotonicity requirement: we must have $\text{minus}(a, b) > \text{minus}(a, c)$ if $b > c$, which implies $\text{minus}(a, b) \cdot \text{size} \geq \text{minus}(a, c) \geq \text{minus}(a, c)$ 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 $\text{minus}(\textcolor{blue}{x}, \textcolor{blue}{y}) \geq \textcolor{blue}{y}$. 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 $\textcolor{blue}{x}$ in the last rule. Yet, here we do not only prove its termination, but also find a bound to its complexity.

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

$$\begin{aligned} \text{zip}(\textcolor{blue}{F}, \emptyset) &= \textcolor{blue}{l} \\ \text{zip}(\textcolor{blue}{F}, \text{cons}(y, q)) &= \text{cons}(\textcolor{blue}{F} \cdot y, \text{zip}(\textcolor{blue}{F}, q, \textcolor{blue}{l})) \end{aligned}$$

22

A more challenging higher-order tuple interpretation

$$\text{fold}(\textcolor{blue}{F}, x, \text{cons}(y, l)) \rightarrow \text{fold}(\textcolor{blue}{F}, (\textcolor{blue}{F} \cdot x \cdot y), l)$$

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

Interpretation:

Where:

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

23

A more challenging higher-order tuple interpretation

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

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

- $\text{makesm}_{\sigma,\tau}$ should itself be monotonic!

- we need to have $(\lambda x.s) \cdot t > s[x := t]$

19 Example: interpreting list functions

$$\begin{array}{rcl}
 \text{append}(\boxed{}, l) & \rightarrow & \text{green} \\
 \text{append}(\text{cons}(x, l), q) & \rightarrow & \text{cons}(x, \text{append}(l, q)) \\
 \text{sum}(\boxed{}) & \rightarrow & \text{blue} \\
 \text{sum}(\text{cons}(x, \text{green})) & \rightarrow & \text{add}(x, \text{sum}(\text{green}))
 \end{array}$$

Interpretations:

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

20 Higher-order tuple interpretations: an example

$$\begin{array}{rcl}
 \boxed{} & :: & \text{list} \\
 \text{cons} & :: & \mathbb{N} \Rightarrow \text{list} \Rightarrow \text{list} \\
 \text{map} & :: & (\mathbb{N} \Rightarrow \mathbb{N}) \Rightarrow \text{list} \Rightarrow \text{list} \\
 \text{map}(F, \boxed{}) & \rightarrow & \boxed{} \\
 \text{map}(F, \text{cons}(x, l)) & \rightarrow & \text{cons}(F \cdot x, \text{map}(F, l))
 \end{array}$$

Let:

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

2. Tuple interpretations

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

14

An observation

Consider:

$$\mathbf{add}(\mathbf{s}^n(\mathbf{0}), \mathbf{s}^m(\mathbf{0})) = 1 + m + 2 * n$$

- 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 *tonat*($\langle x, y \rangle$) = x . That is, if $a > b$ in \mathbb{N}^2 then *tonat*(a) > *tonat*(b) – so if we can express s 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.

Separating cost and size

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

Let:

$$\begin{array}{rcl} \mathbf{0} & = & \langle \mathbf{cost} & \mathbf{size} \\ & = & \langle & \rangle \\ \mathbf{s}(\mathbf{x}) & = & \langle \mathbf{cost} & \mathbf{size} \\ & = & \langle & \rangle \\ \mathbf{add}(\mathbf{x}, \mathbf{y}) & = & \langle \mathbf{cost} + \mathbf{y}.cost & \mathbf{size} + 1 \\ & = & \langle & \rangle \end{array}$$

Then:

$$\begin{array}{rcl} \mathbf{add}(\mathbf{0}, \mathbf{y}) & = & \langle 1 + \mathbf{y}.cost, \mathbf{y}.size \rangle \\ & > & \langle \mathbf{y}.cost, \mathbf{y}.size \rangle \\ \mathbf{add}(\mathbf{s}(\mathbf{x}), \mathbf{y}) & = & \langle 2 + \mathbf{x}.cost + \mathbf{y}.cost, \mathbf{x}.size + \mathbf{y}.size \rangle \\ & > & \langle 1 + \mathbf{x}.cost + \mathbf{y}.cost, 1 + \mathbf{x}.size + \mathbf{y}.size \rangle = \mathbf{s}(\mathbf{add}(\mathbf{x}, \mathbf{y})) \end{array}$$

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

15
16
17
18

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

Separating cost and size

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

Then:

$$\begin{array}{rcl} \mathbf{a}(\mathbf{b}(\mathbf{x})) & = & 2 + 2 * \mathbf{x} > 1 + 2 * \mathbf{x} = \mathbf{b}(\mathbf{a}(\mathbf{x})) \\ \text{Hence: } \mathbf{a}^n(\mathbf{b}^m(\epsilon)) & = & 2^n * m: \text{exponential!} \end{array}$$

17

$$\begin{array}{rcl} \mathbf{Let:} & & \mathbf{cost} & \mathbf{size} \\ \mathbf{a}(\mathbf{x}) & = & \langle & \rangle \\ \mathbf{b}(\mathbf{x}) & = & \langle & \rangle \\ \epsilon & = & \langle & \rangle \\ & = & \langle & \rangle \\ & = & \langle & \rangle \\ & = & \langle & \rangle \end{array}$$

Then:

$$\mathbf{a}(\mathbf{b}(\mathbf{x})) = \langle \mathbf{x}_1 + \mathbf{x}_2 + 1, \mathbf{x}_2 + 1 \rangle > \langle \mathbf{x}_1 + \mathbf{x}_2, \mathbf{x}_2 + 1 \rangle = \mathbf{b}(\mathbf{a}(\mathbf{x}))$$

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

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

Tuple interpretations

Definition: monotonic algebras with $A_k = \mathbb{N}^{K[k]}$ for all k (where $K[\ell]$ is a positive integer for all ℓ).

⇒ 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 [f].

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 $\text{list}(\mathbf{N} \Rightarrow \mathbf{N})$ to be mapped to a function space such as $\mathbf{N} \Rightarrow \mathbf{N}$. However, for this lecture, we will limit interest to tuples of the form \mathbb{N}^k .

Example **sort interpretations**:

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

Let:

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