

Programming Obesity: A Code Health Epidemic

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Imagine/Dream

Direct
Fast
Transmissive

Empower vs. Control

More domain knowledge,
Less systems rumination

The Main Thing

Obsolete? Hardly.

Essential
vs.
Accidental

We Are
Consumer and Producer

We are in this
together

Tech Stack



Programming Obesity

Cascading Systemic
Failure to Simplify
and Tendency Towards
Unsustainable Waste

Structural Complexity
Performance
Economy

Change

Change is
Uncomfortable

Break Points

Simple:

Faster:

Economical:

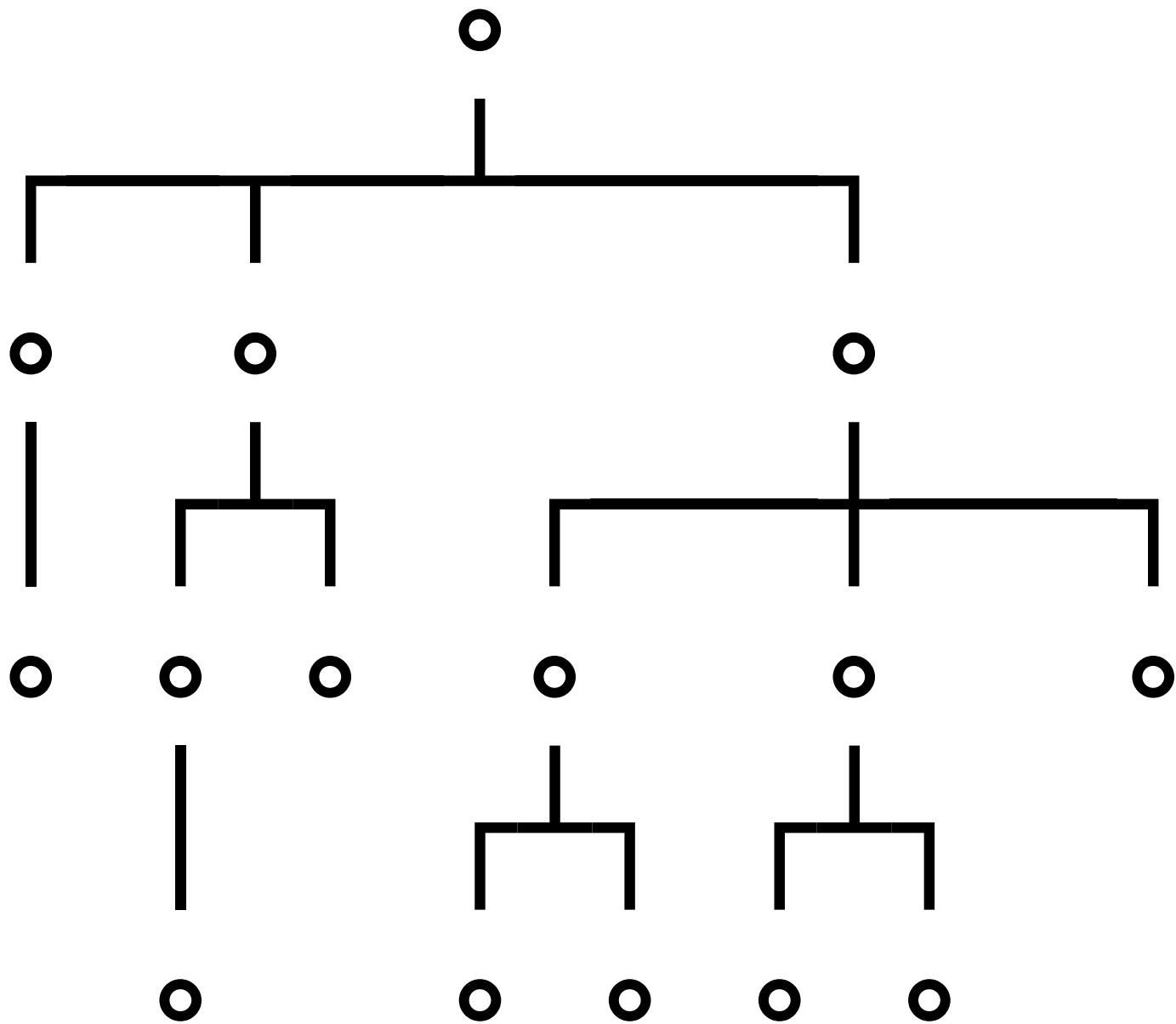
Impact

Don't wait

Given Up

Where do we start?

Generalized pointers
are the **Refined Sugar**
of Programming



(Expr (e)

v

(B.0 ([name ns]) e)

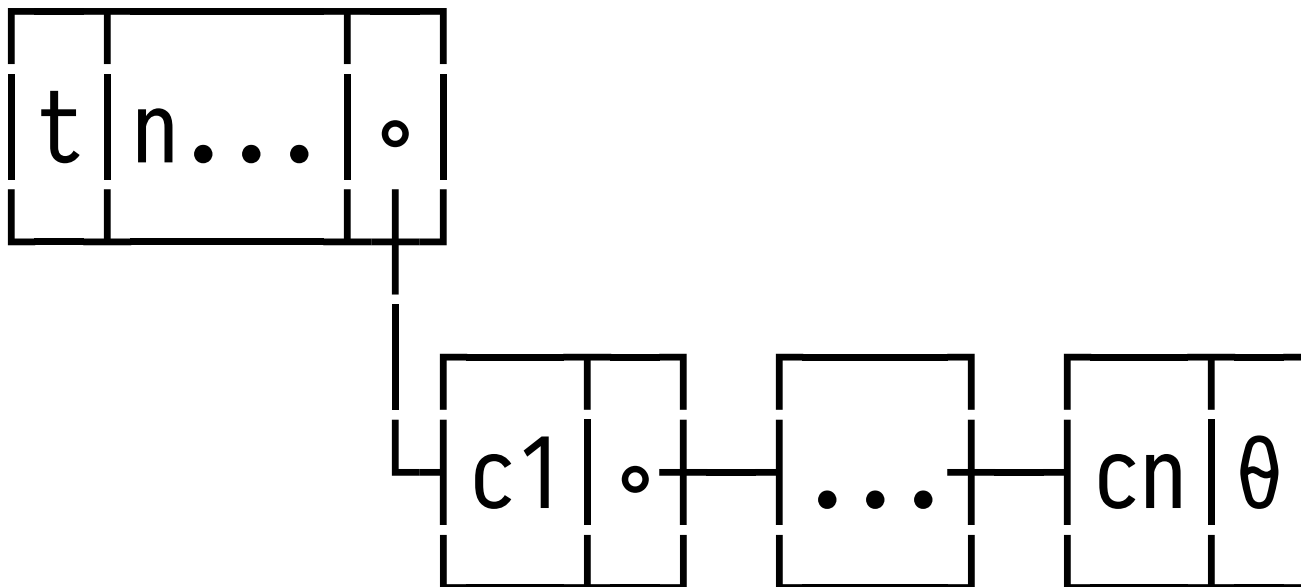
(A.0 ([name ns]) num* ...)

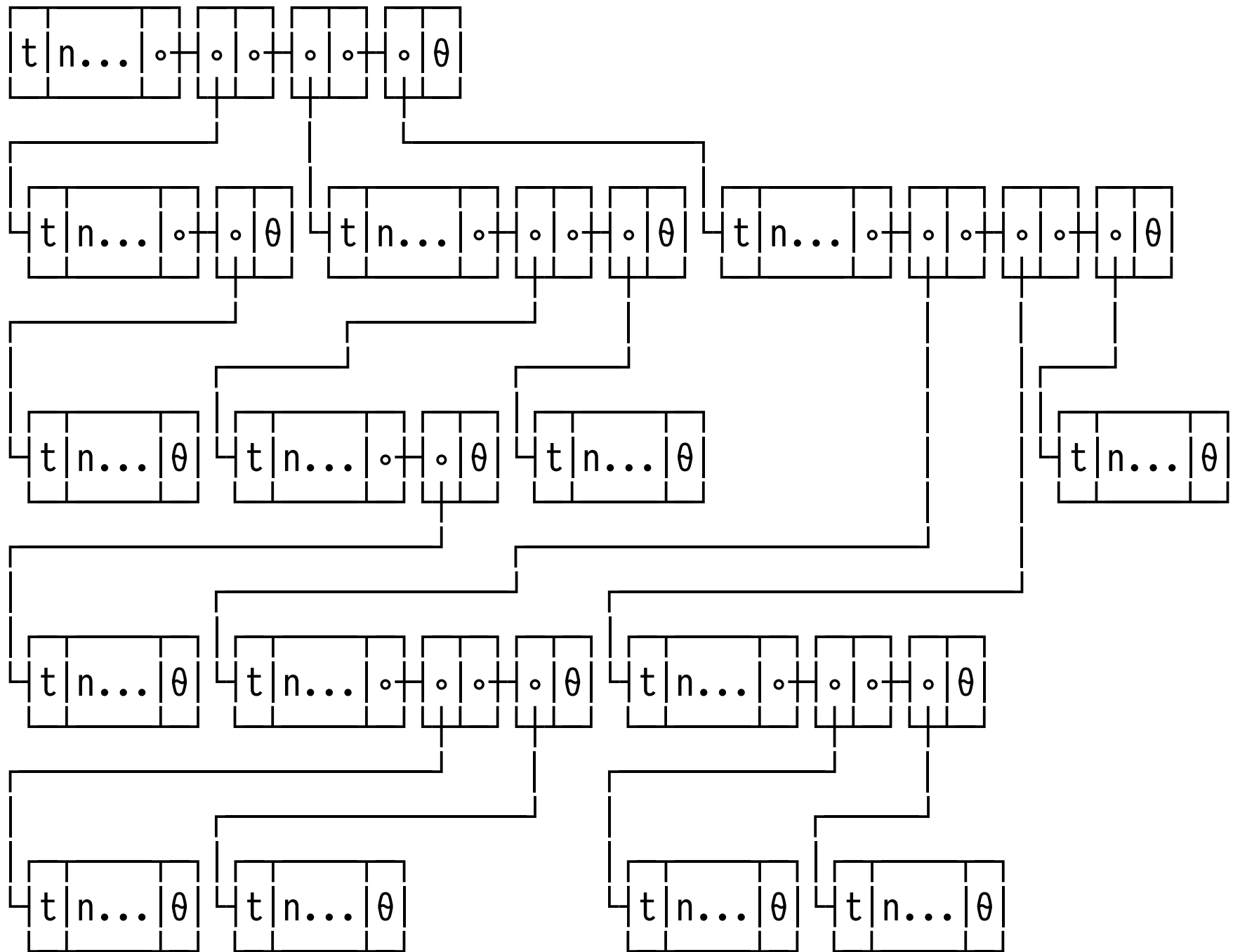
(A.3 ([name ns]) v* ...)

(E.1 ([name ns]) f e)

(E.2 ([name ns]) e f ebrk)

(E.4 ([name ns]) v ebrk e))





Eliminate
expedient complexity
and waste

Achieve
real simplicity

Holistic Economy

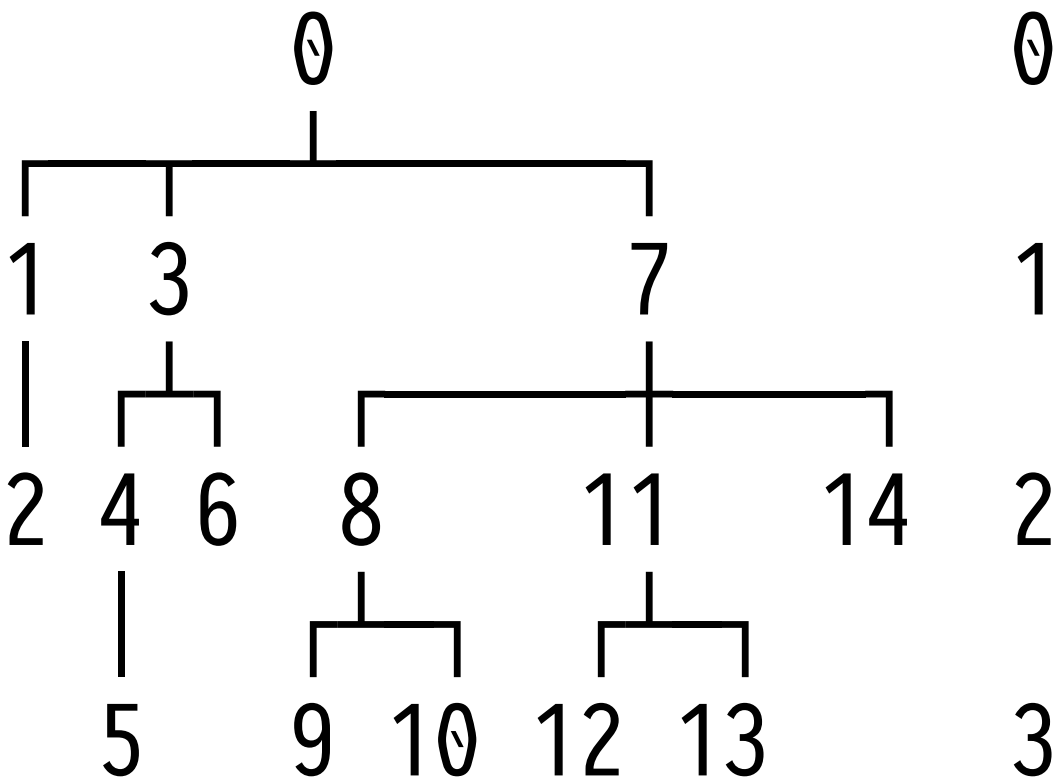
Can a compiler be:
GPU hosted, fast,
portable, and simple?

Impossible

Eliminate Everything

A New (Old) Hope

APL



$\phi(z \neq d), \bar{d}$

0	1	2	3	4	5	6	7	8	9	10	11	12	13	14
0	1	2	1	2	3	2	1	2	3	3	2	3	3	2

Simple:		94%	15×
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Faster:	CPU	89%	9×
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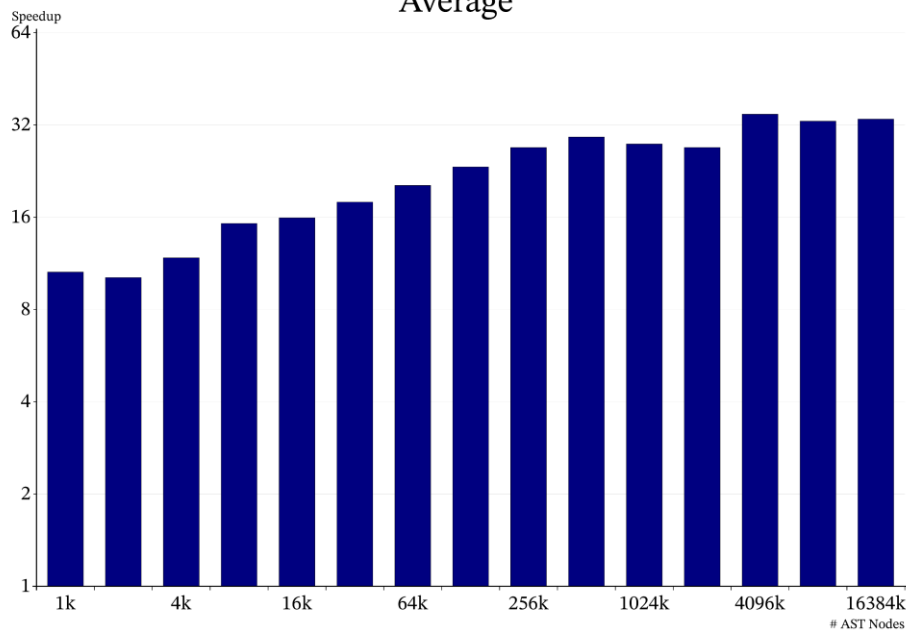
	GPU	98%	56×
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Economy:	Raw	70%	3×
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	Ratio	88%	8×
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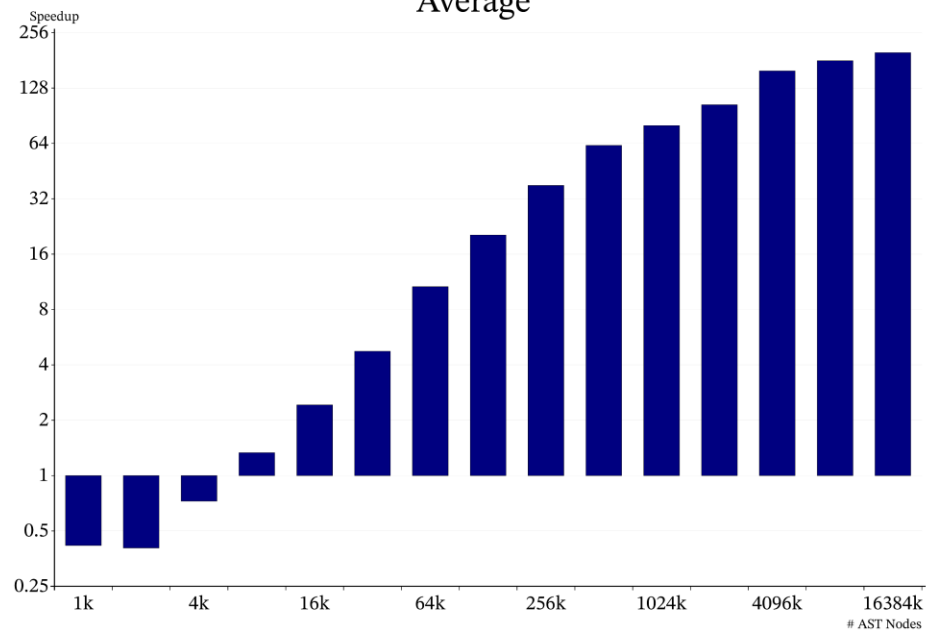
	LoC	Tokens	Names	Nodes
<i>Nanopass</i>	1012	20947	248	14680
<i>Co-dfns</i>	17	760	74	948

Average



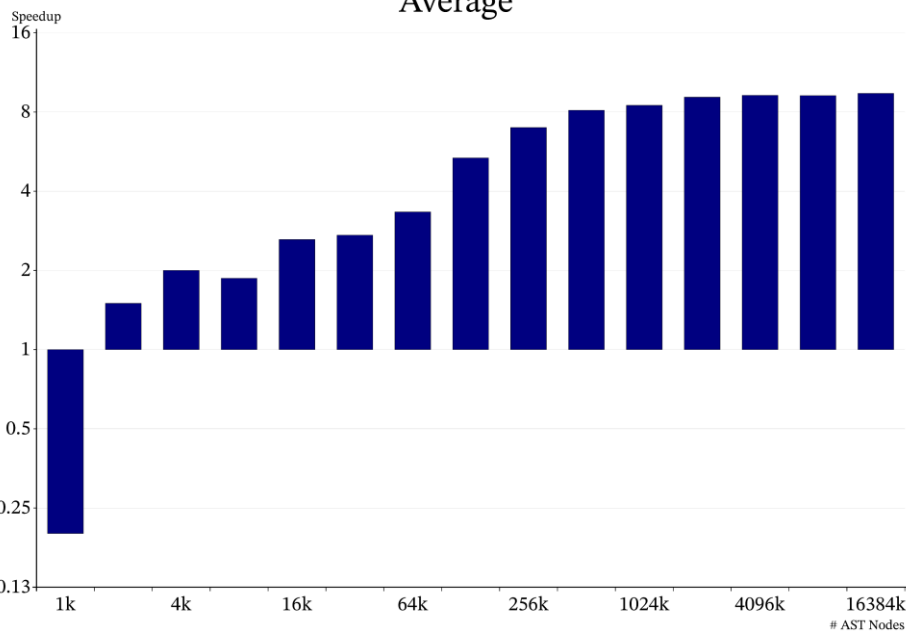
Racket vs. APL-CPU (Racket time divided by APL-CPU time)

Average



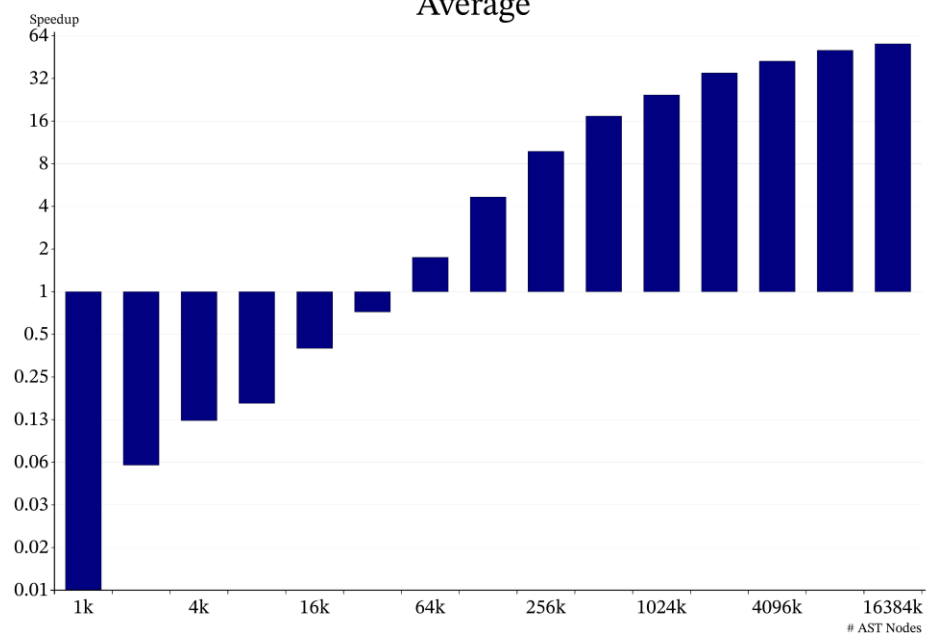
Racket vs. APL-GPU (Racket time divided by APL-GPU time)

Average



Chez vs. APL-CPU (Chez time divided by APL-CPU time)

Average



Chez vs. APL-GPU (Chez time divided by APL-GPU time)

Memory Usage

Size	Dya	Rack	Chez	Size	Dya	Rack	Chez	Size	Dya	Rac	Chez
0	0.009	0.065	2.027	5	0.129	2.056	4.055	10	3.985	65.721	93.258
1	0.013	0.13	2.027	6	0.254	4.11	6.082	11	7.966	131.44	186.516
2	0.021	0.259	2.027	7	0.503	8.217	10.137	12	15.927	262.877	371.508
3	0.036	0.516	2.027	8	1	16.432	22.301	13	31.849	525.752	739.980
4	0.067	1.029	2.027	9	1.995	32.862	44.602	14	63.692	1051.502	1485.023

Feeling is Believing

Positive Cascading Effects

Conclusion

Start with APL

Thank You.

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A data parallel compiler hosted on the GPU

<http://www.dyalog.com>

$$i \leftarrow (\underline{1}(\sim t \in 3 \ 4) \wedge t[p]=3), \{w \neq 2 \mid \underline{1} \neq w\} \underline{1} t[p]=4 \diamond p \ t \ k \ n \ r \neq \leftarrow c m \leftarrow 2 @ i \vdash 1 \rho \neq p$$

$$p \ r \ i \ I \neq \leftarrow c j \leftarrow (+ \backslash m) - 1 \diamond n \leftarrow j \ I @ (0 \leq \vdash) n \diamond p[i] \leftarrow j \leftarrow i - 1$$

$$k[j] \leftarrow - (k[r[j]] = 0) \vee 0 @ (\{ \supset \phi w \} \exists p[j]) \vdash t[j] = 1 \diamond t[j] \leftarrow 2$$

```
(define i
  (catenate
    (where
      (and
        (not (members-of t `(3 4)))
        (pequal? 3 (indexing t p))))
    ((lambda (w)
      (replicate
        (modulo
          (index-gen (tally w))
          2)
        w))
      (where (pequal? 4 (indexing t p))))))
(define m ((at 2 i) (reshape (tally p) 1)))
(define-values (p t k n r)
  (let ([m^ (enclose m)])
    (apply values
      (map (lambda (x) (replicate m^ x))
           p t k n r))))
(define j (pminus (scan-first + m) 1))
(define-values (p r i)
  (let ([j^ (enclose j)])
    (apply values
      (map (lambda (x) (indexing j^ x)) p r i))))
```

```
(define n
  ((at indexing (lambda (x)
    (pless-equal? 0 x)))
    j n))
(define j (pminus i 1))
(array-set! p i j)
(array-set! k j
  (negate
    (or
      (pequal? 0
        (indexing k
          (indexing r j)))
      ((at 0
        ((key (lambda (x)
          (disclose (reverse x)))
            (indexing p j)))
          (pequal 1 (indexing t j)))))))
(array-set! t j 2)
```