

Straightforward parallelization of polynomial multiplication using parallel collections in Scala

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## Parallelization of symbolic computations

- \* Numeric computations

- Several arithmetic operations executed in parallel
  - Linear algebra
  - CPU-intensive

- \* Symbolic computations : polynomials

- Same as above, and:

- Arithmetic operation itself is parallelized
  - Multiplication, division, gcd
  - Reduction, Gröbner bases (multivariate)
  - CPU and memory-intensive (cache issues)

Polynomial multiplication

Multivariate polynomials

Distributive representation

$$X = X_0 + X_1 + \dots + X_n$$

$$X_i = c_i m_i$$

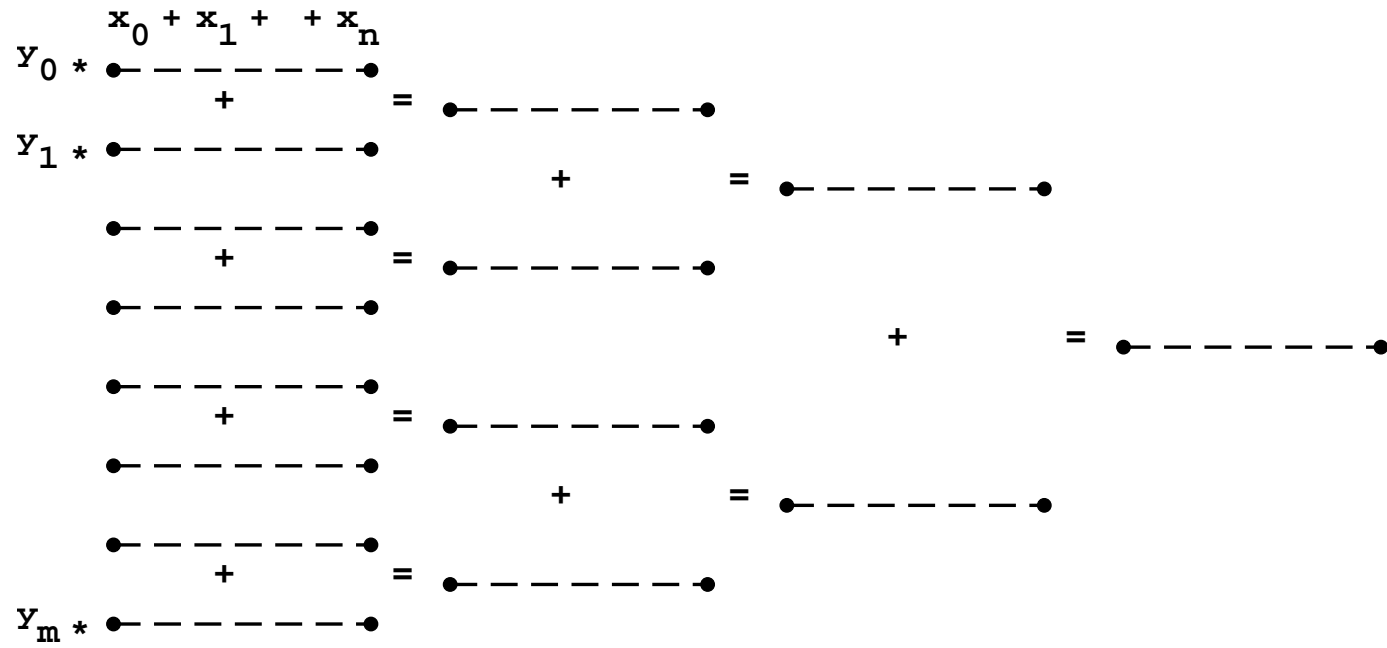
Product

$$xy$$

## Polynomial multiplication : sequential

$$\begin{array}{r}
 \begin{array}{c} x_0 + x_1 + \dots + x_n \\ y_0 * \bullet \text{---} \text{---} \text{---} \text{---} \bullet \\ \quad \quad \quad + \\ y_1 * \bullet \text{---} \text{---} \text{---} \text{---} \bullet \\ \quad \quad \quad + \\ \bullet \text{---} \text{---} \text{---} \text{---} \bullet \\ \quad \quad \quad + \\ \bullet \text{---} \text{---} \text{---} \text{---} \bullet \\ \quad \quad \quad + \\ \bullet \text{---} \text{---} \text{---} \text{---} \bullet \\ \quad \quad \quad + \\ \bullet \text{---} \text{---} \text{---} \text{---} \bullet \\ \quad \quad \quad + \\ \bullet \text{---} \text{---} \text{---} \text{---} \bullet \\ y_m * \bullet \text{---} \text{---} \text{---} \text{---} \bullet \end{array}
 \end{array}$$

Polynomial multiplication : parallel



Polynomial multiplication : sequential

```
type T = List[(Array[N], C)]
```

```
def times(x: T, y: T) = (zero /: y) { (l, r) =>  
  val (a, b) = r  
  l + multiply(x, a, b)  
}
```

Polynomial multiplication : sequential

```
type T = List[(Array[N], C)]
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```
def times(x: T, y: T) = y.foldLeft(zero)({ (l, r) =>
  val (a, b) = r
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Polynomial multiplication : sequential

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})
```

```
def multiply(x: T, m: Array[N], c: C) = x.map { r =>
  val (s, a) = r
  (s * m, a * c)
} filter { r =>
  val (_, a) = r
  !a.isZero
}
```



Polynomial multiplication : parallel

```
type T = List[(Array[N], C)]
```

```
def times(x: T, y: T) = y.par.aggregate(zero)({ (l, r) =>
  val (a, b) = r
  l + multiply(x, a, b)
}, _ + _)
```

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def multiply(x: T, m: Array[N], c: C) = x.map { r =>
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```

## Experimental setup

Intel Atom D410 at 1.66Ghz with ((32K, 24K), 512K) cache

Single core

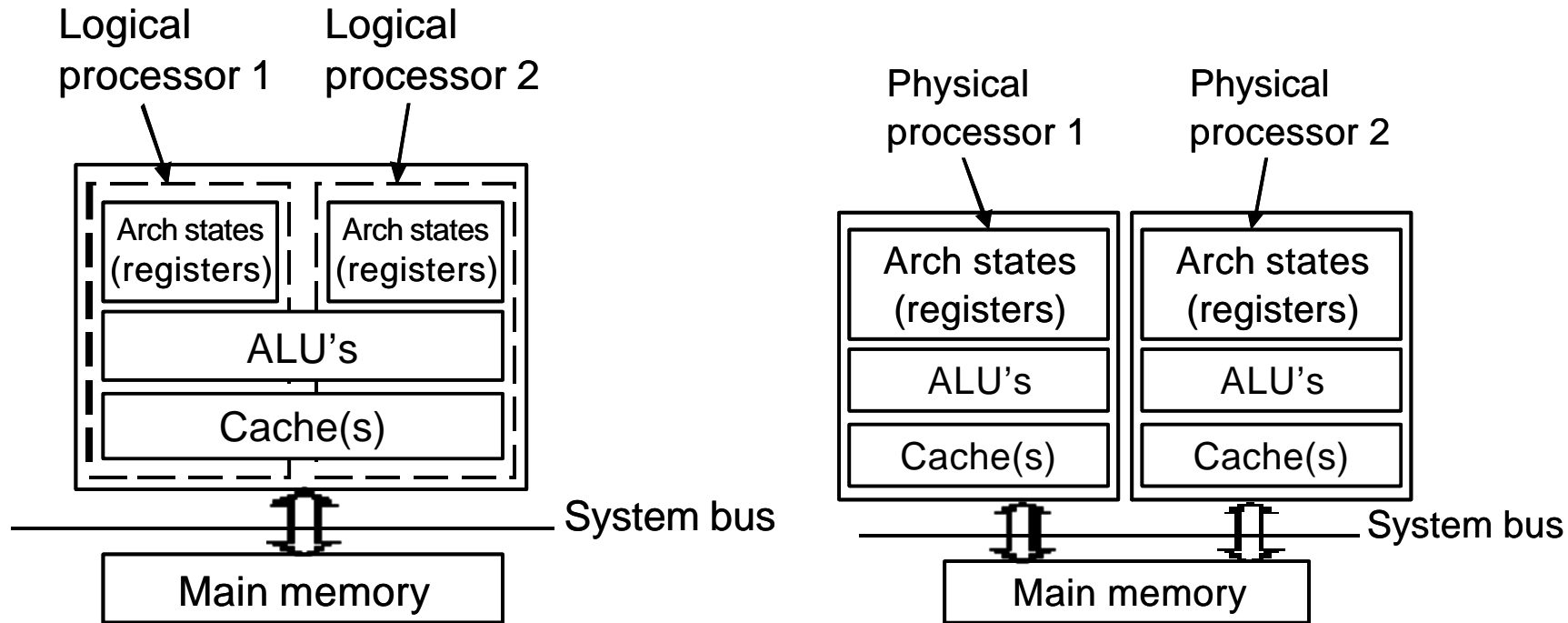
Hyper-threading

Parallel timings should not be worse than sequential

Could be eventually better (20 %)

Further experiments need to be done on multicore hardware

## Experimental setup



Hyper-threading

Dual-processor

(Chen et al. Media Applications on Hyper-Threading Technology - Intel Technology Journal, Q1, 2002)

Test case

Squaring a sparse polynomial  $p^n$  with

$$p = 1 + x + y + z \in \mathbb{Z}[x, y, z] \text{ and}$$

$$n \in \mathbb{N} \text{ sufficiently large : } n \geq 20$$

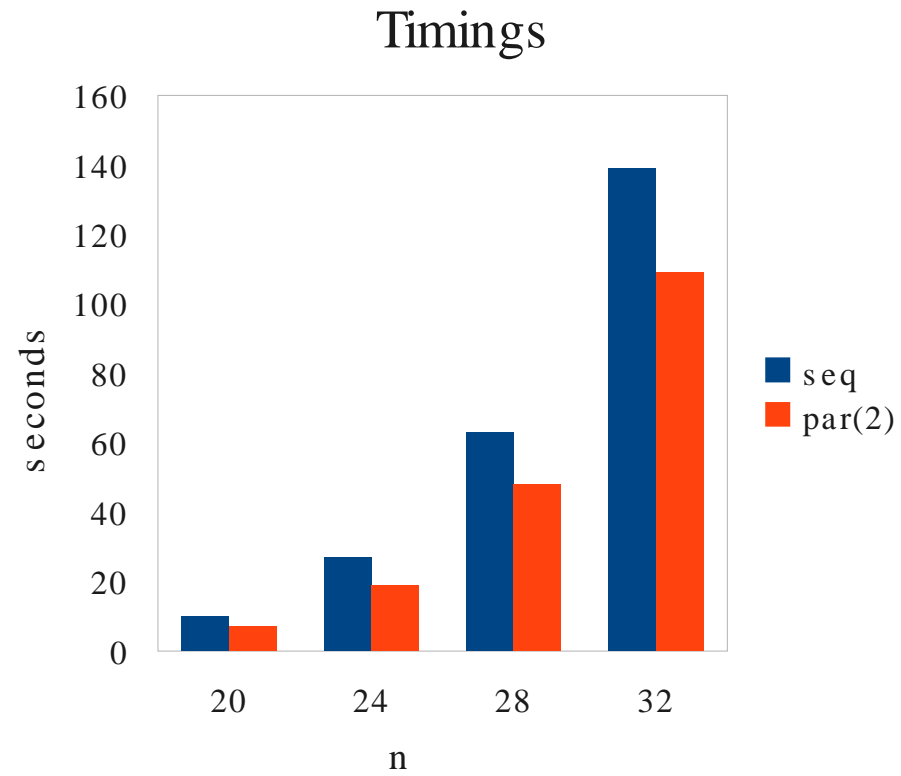
(Fateman, R. J. DRAFT: Comparing the speed of programs for sparse polynomial multiplication, 2002)

Test case : implementation

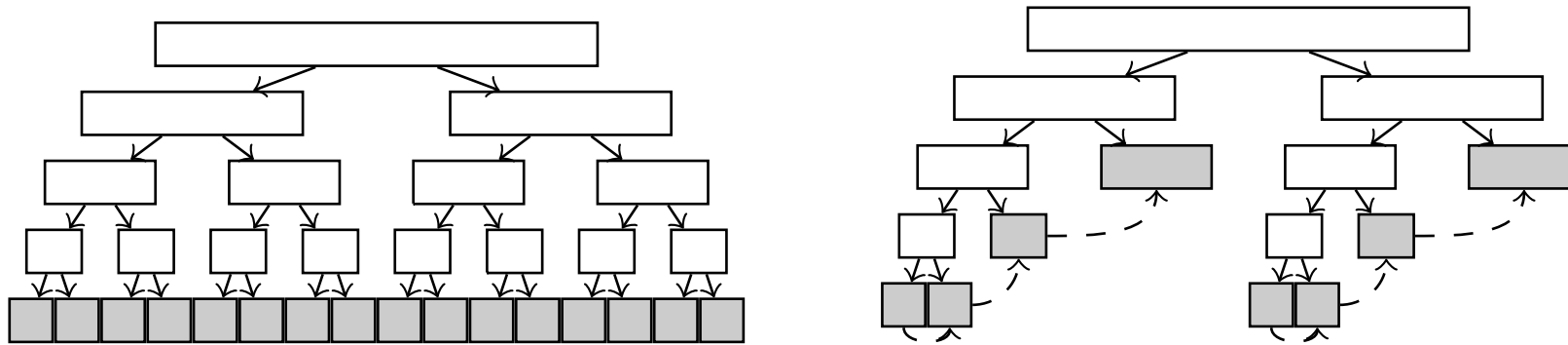
```
import scas._
import Implicits.ZZ
implicit val r = Polynomial(ZZ, 'x, 'y, 'z)
val Array(x, y, z) = r.generators
val p = 1 + x + y + z
val q = pow(p, 20)
val q1 = 1 + q
val q2 = q * q1
```

## Timings

n	seq	par(2)	speedup
20	10	7	1.38
24	27	19	1.37
28	63	48	1.32
32	139	109	1.27



## Fine-grained and exponential task splitting

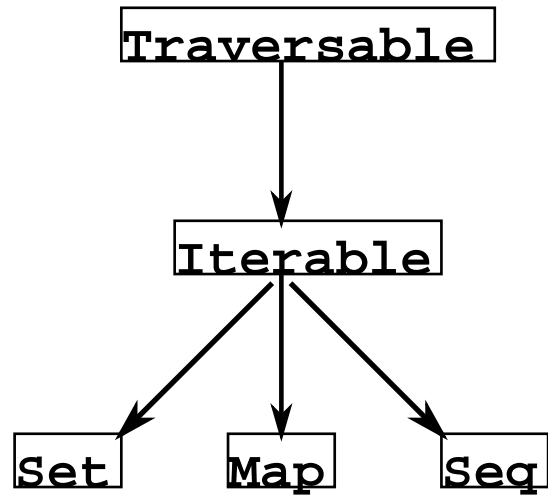


$$\text{threshold} = \max \left\{ n, \frac{1}{8P} \right\}$$

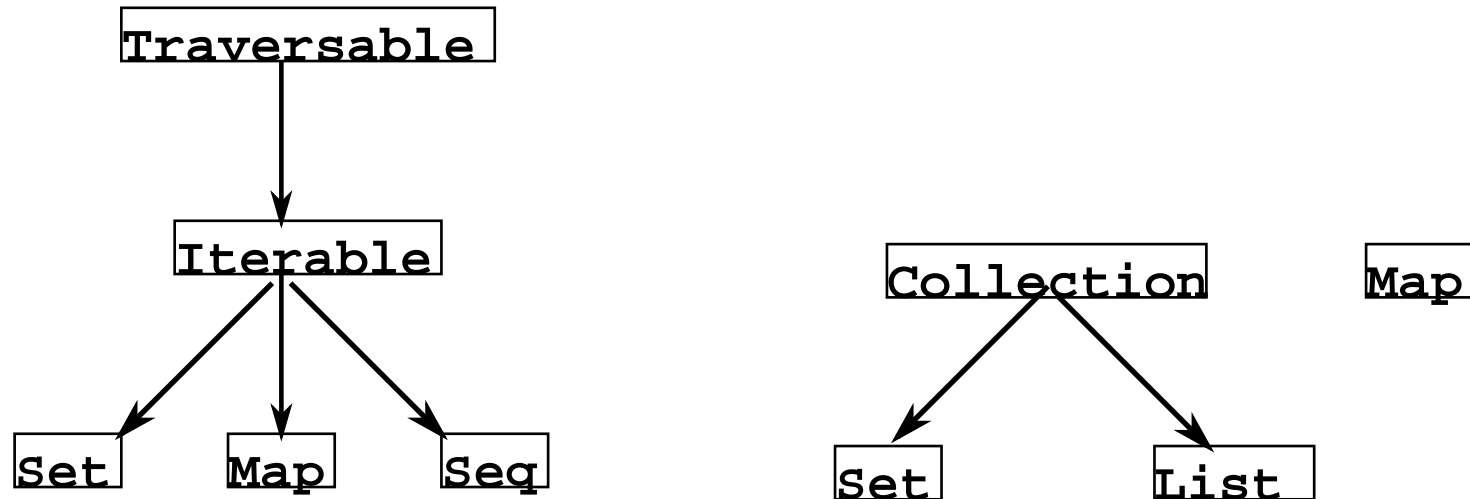
"stolen tasks are divided into exponentially smaller tasks until a threshold is reached and then handled sequentially starting from the smallest one, while tasks that came from the processor's own queue are handled sequentially straight away" (Prokopec, A.; Bawgell, P.; Rompf, T. & Odersky, M. On a Generic Parallel Collection Framework, 2011)



## Collection base classes hierarchy



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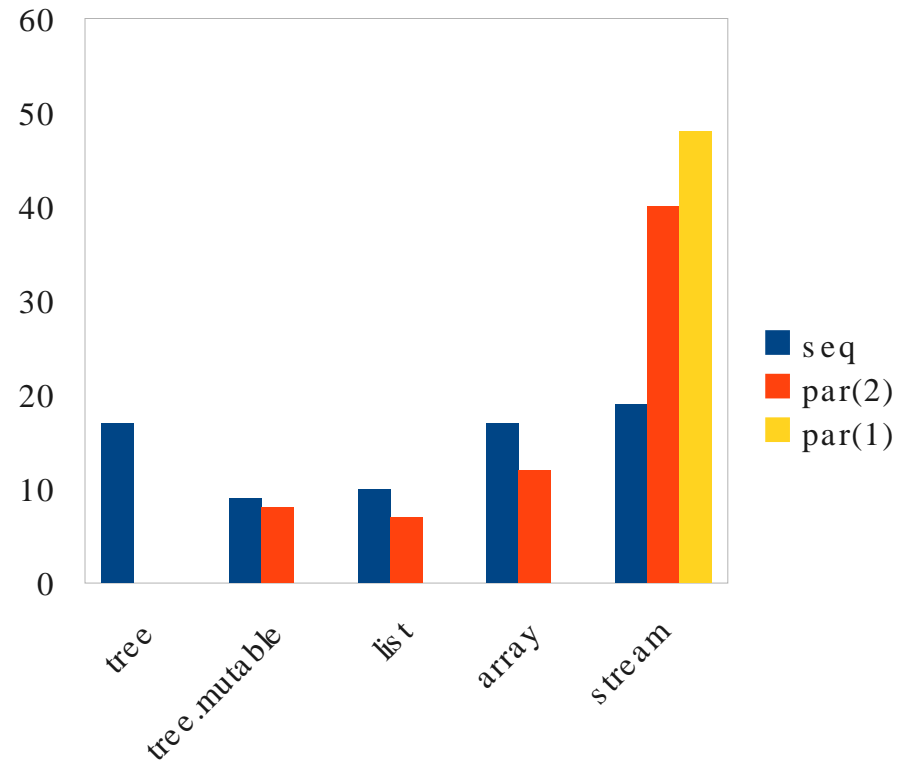


Traversable[A]

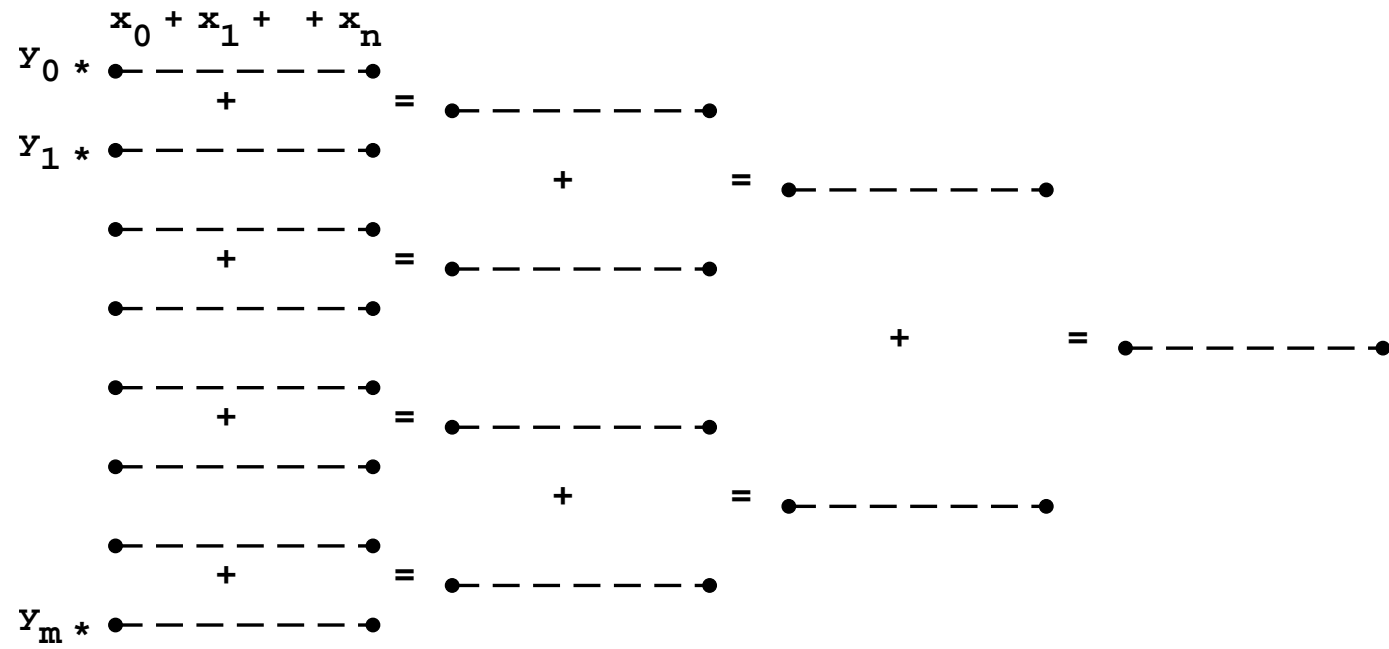
```
def map[B, That](f: A => B): That
def flatMap[B, That](f: A => GenTraversableOnce[B]): That
def filter(p: A => Boolean): Traversable[A]
def foreach[U](f: A => U): Unit
def forall(p: A => Boolean): Boolean
def exists(p: A => Boolean): Boolean
def count(p: A => Boolean): Int
def reduce[A1 >: A](op: (A1, A1) => A1): A1
def aggregate[B](z: B)(seqop: (B, A) => B, combop: (B, B)
=> B): B
def sum[B >: A](implicit num: Numeric[B]): B
def product[B >: A](implicit num: Numeric[B]): B
def min[B >: A](implicit cmp: Ordering[B]): A
def max[B >: A](implicit cmp: Ordering[B]): A
```

## Other data structures (n = 20)

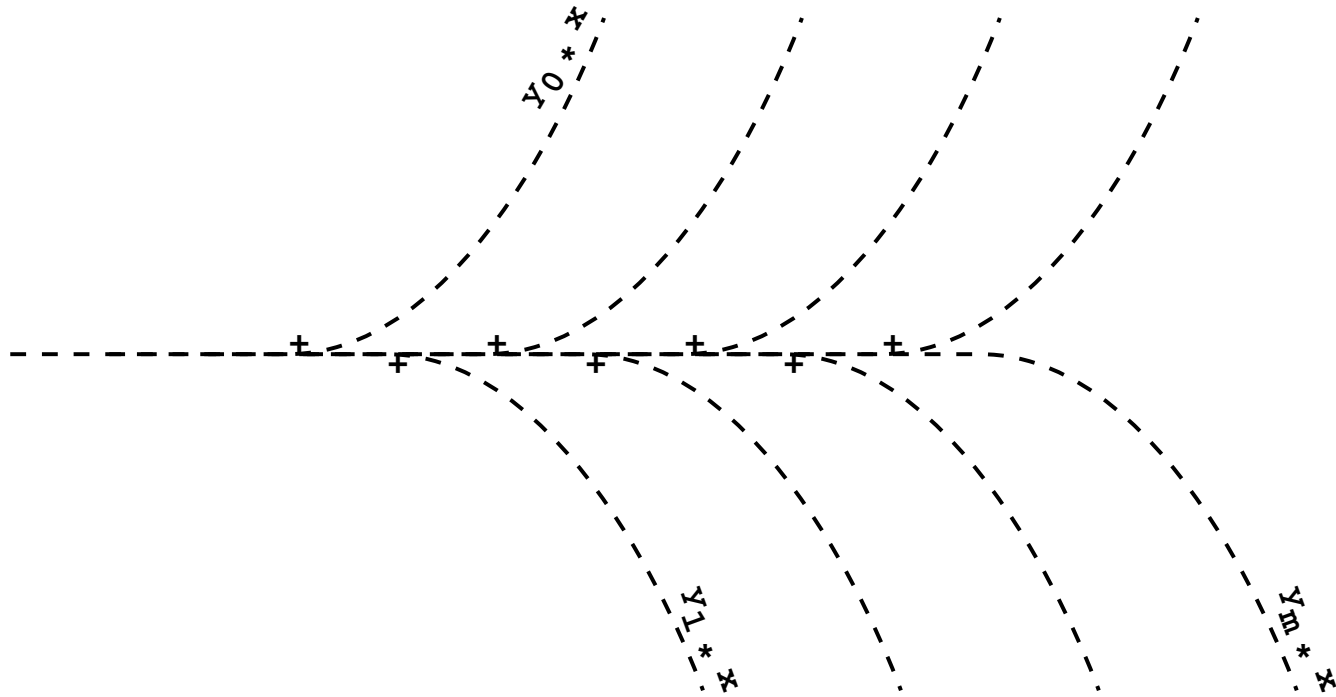
structure	seq	par(2)	par(1)
tree	17		
tree.mutable	9	8	
list	10	7	
array	17	12	
stream	19	40	48



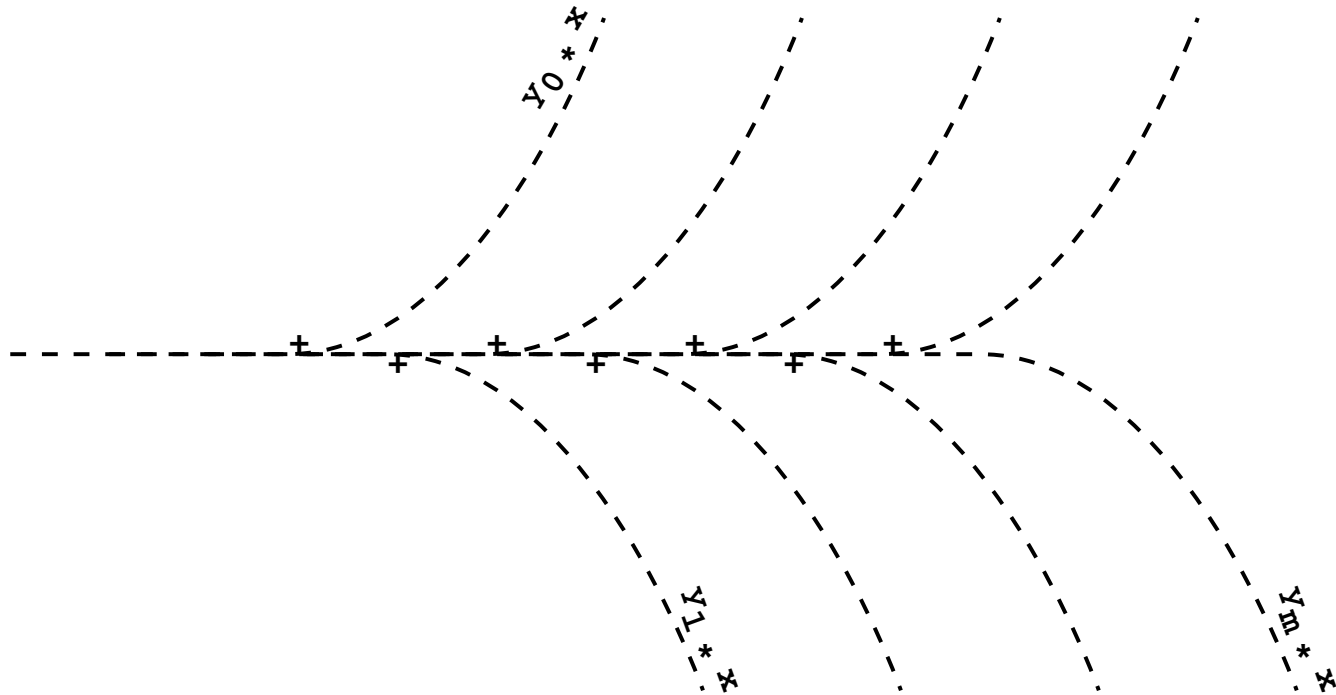
## Data parallelism



## Task parallelism



## Task parallelism



Thank you !

<http://github.com/rjolly/scas>